

## EU Regional Competitiveness Index 2010

Paola Annoni and Kornelia Kozovska


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# EU REGIONAL COMPETITIVENESS INDEX RCI 2010 

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## ExECUTIVE Summary

The concept of competitiveness has in the last decades extended from the micro-level of firms to the macro-level of countries. Between the two levels stands the concept of regional competitiveness which is the focus of the "EU Regional Competitiveness Index", RCI hereafter, a joint project between DG Joint Research Centre and DG Regional Policy.

The final goal is measuring the competitiveness of European regions at the NUTS2 level by developing a composite index. But, why measuring regional competitiveness is so important? Because "if you can not measure it, you can not improve if" (Lord Kelvin). A quantitative score of competitiveness will facilitate Member States in identifying possible regional weaknesses together with factors mainly driving these weaknesses. This in turn will assist regions in the catching up process.

The study starts from the review of the latest literature contributions to the concept of 'regional competitiveness' and of some well-known existing competitiveness indices at country and regional level (NUTS1 and NUTS2). At the country level, the Global Competitiveness Index by the World Economic Forum, and the World Competitiveness Yearbook by the Institute for Management Development (IMD) are presented. At the regional NUTS1 level, the European Competitiveness Index by the University of Wales Institute is discussed. A simpler but more detailed geographical description of competitiveness is offered by the 'Altas of Regional Competitiveness' (Eurochambers)," reflecting the international recognition of the importance of analysis at the regional NUTS2 level. Specific examples of competitiveness measures at the regional level in some European countries are also discussed.

The WEF Global Competitiveness Index - GCI - has been the main reference framework for the construction of the RCI. This choice has been driven by the fact that GCI is the most internationally recognized and acclaimed index in the field of competitiveness and its framework covers a very comprehensive set of aspects relevant to competitiveness. There are, however, some key differences that distinguish the RCI from GCI due to the RCI European and regional dimension.

Eleven pillars are included in the RCI with the objective of describing different dimensions of the level of competitiveness. The pillars are designed to capture short- as well as longterm capabilities of the region. They are classified into three major groups: the pillars Institutions, Macro-economic stability, Infrastructure, Health and Quality of Primary \& Secondary Education are included in the first group and represent the key basic drivers of all types of economies. As the regional economy develops, other factors enter into play for its advancement in competitiveness and are grouped in the second group of pillars - Higher Education/ Training and Lifelong Learning, Labor Market Efficiency and Market Size. At the most advanced stage


Basic pillars

1. Institutions
2. Macrocconomic stability
3. Infrastructure
4. Health
5. Quality of Primary and

Sccondary Education
of development of a regional economy, key drivers for regional improvement are factors related to Technological Readiness, Business Sophistication and Innovation, included in the third group.

The set of indicators which populate each pillar

## RCI general framework

 is carefully chosen according to the literature review, experts' opinion and data availability. The major data source is Eurostat with some additional official sources - OECD-PISA, OECD Regional Patent database, European Cluster Observatory, World Bank Governance Indicators and Ease of Doing Business Index - where appropriate data was not directly available from Eurostat.Most recent data have been used for all indicators, with a temporal range for most indicators between 2007 and 2009.

A detailed statistical analysis is carried out separately for each pillar with the aim of assessing the consistency of the proposed framework both at the level of indicators and of pillars. The analysis is twofold: a univariate analysis indicator by indicator and a multivariate analysis on each pillar as a whole. The former allows for detecting possible problems with: i) missing
data; ii) distribution asymmetry and outliers and iii) different measurement scales. These problems are addressed by adopting: i) specific imputation methods; ii) power-type transformations to correct for skeweness; iii) standardization. The multivariate analysis is carried out at the pillar level on the set of indicators as a whole. The aim is to assess their contribution in describing the latent dimension behind each pillar. 'Anomalous' indicators are in some cases detected and excluded from further analysis.

The final RCI is composed of a total number of 69 indicators, chosen by a starting set of 81 candidate indicators. The statistical analysis showed as most consistent pillars Institutions, Quality of Primary and Secondary Education, Labor Market Efficiency, Market Size and Innovation.

The key driver for the computation of the RCI has been to keep it simple, to be easily understood by non-statisticians, and at the same time robust and consistent. For each pillar, RCI sub-scores are computed as a simple average of the transformed/normalized indicators. Scores at the pillar group level (sub-indexes) are computed as an average of the corresponding sub-scores. The overall RCI score is the result of a weighted aggregation of the three sub-indexes. For the final aggregation we follow the approach that the World Economic Forum adopts for the GCI with the aim of taking into account the level of


Geographical distribution of RCI score heterogeneity of European regions, especially after the 2004 and 2007 enlargements. The set of weights adopted for aggregating the sub-indexes depend on the level of development of the regions, classified into medium, intermediate and high stage on the basis of their GDP value. Regions in the medium stage are assigned more weight to the basic and efficiency pillars in comparison to the innovation pillars. The level of competitiveness of more developed economies, on the
other hand, takes into account to a larger extent their innovation capability as a key driver for their advancement. The weighting scheme of pillar groups has the effect of not penalizing regions on factors where they lay too far behind. The RCI message is then more constructive: the index provides a measure of competitiveness which allows for fair comparison of European regions and highlights realistic areas of improvement. The final RCI shows a heterogeneous situation across EU regions with Eastern and Southern European regions showing lower performance while more competitive regions are observed in Northern Europe and parts of Continental Europe.

As for almost every composite indicator, the procedure followed for the setting up of the RCI is affected by a certain degree of subjectivity. A full robustness analysis is then performed to check the sensitivity of the index with respect to these choices. The variation in score and ranks of the regional RCI is assessed on the basis of the following scenarios:

- Different sets of weights chosen by random selection within a selected range of variation plus different GDP levels for the classification of the region's development stage;
- Different composition of the index by discarding one dimension (pillar) at a time to verify whether the pillar contribution to the RCI framework is well balanced;
- Different types of aggregation based on fully or non-compensatory operators (Ordered Weighted Operators).


RCI robustness analysis

A Monte-Carlo type analysis is carried out for a total number of 1200 different simulations. Overall, the distribution of the shift in rank for all the simulations and all the regions clearly shows a pick around zero. A closer look at the distribution highlights
that in more than $80 \%$ of the cases the shift in rank is at most of 5 positions. The RCI index proves to be rather robust with only a very small fraction of regions with 'volatile' rankings. The analysis of the impact of each pillar on the final score shows that the most influential pillars are Higher Education/Training and Lifelong Learning, Labor Market Efficiency and Market Size. This is in line with the fact that these three pillars are assigned, on average across the three development stages, the highest weights.

RCI represents the first measure of the level of competitiveness at the regional level covering all EU countries. It takes into account both social and economic aspects, including the factors which describe the short and long term potential of the economy. A statistical analysis has been used to support and, in some cases, to correct the ideal framework of the index, which is characterized by a simple and, at the same time, multifaceted structure. A series of tests have been used to 'stress' the index, which proved to be rather consistent with respect to a set of key (at least to our judgment) sources of subjectivity and uncertainty. The RCI provides a synthetic picture of the level of competitiveness of Europe at the NUTS2 level representing, at the same time, a well balanced plurality of different fundamental aspects.

## 1 Defining regional competitiveness

The concept of 'competitiveness' has been largely discussed over the last decades. A broad notion of competitiveness refers to the inclination and skills to compete, to win and retain position in the market, increasing market share and profitability, thus, being commercially successful (Filó, 2007).

An important aspect is the level at which the concept of competitiveness is defined; in most cases the micro and macroeconomic level are considered, which are strictly interrelated. The former is relatively clearly defined and is based on the capacity of firms to compete, grow and be profitable (Martin et al., 2006). The latter is, instead, subject to debate and is generally viewed and measured at the country level. One of the most important definitions of macroeconomic competitiveness is given by the World Economic Forum which states that competitiveness is the "set of institutions, policies and factors that determine the level of productivity of a country" (Schwab and Porter, 2007). The link between the two levels is straightforward: a stable context at the macro level improves the opportunity to produce wealth but does not create wealth by itself. Wealth is created by utilizing at best human, capital and natural resources to produce goods and services, i.e. 'productivity'. But productivity depends on the microeconomic capability of the economy which ultimately resides in the quality and efficiency of the firms (Schwab and Porter, 2007).

Despite the strict linkage between micro (firm) and macro (country) competitiveness, much criticism to the notion of national competitiveness has been raised, mainly due to the existence of an analogy between firms and nations. This is in contrast to the fact that: a) an unsuccessful firm will be expunged from the business whilst this cannot be the case for an underperforming nation; b) the competition among firms is a zero-sum game where the success of one firm destroys opportunities of the others whilst the success of one country may be of benefit for the others (Krugman, 1996). Many authors, with Krugman (1996) and Porter (Porter and Ketels, 2003) among others, agree on the definition of competitiveness as productivity, which is measured by the value of goods and services produced by a nation per unit of human, capital and natural resources. They see as the main goal of a nation the production of high and raising standard of living for its citizens which depends essentially on the productivity with which a nation's resources are employed.

Between the two levels of competitiveness stands the concept of regional competitiveness which has gained more and more attention in recent years, mostly due to the increased attention given to regions as key in the organization and governance of economic growth and the creation of wealth. An important example is the special issue of Regional Studies 38(9), published in 2004, fully devoted to the concept of competitiveness of regions. Regional competitiveness is not only an issue of academic interest but of increasing policy deliberation and action. This is reflected in the interest devoted in the recent years by the European Commission to define and evaluate competitiveness of European regions, an objective closely related to the realization of the Lisbon Strategy on Growth and Jobs.

Regional competitiveness cannot be regarded as neither macroeconomic nor microeconomic concept. A region is neither a simple aggregation of firms nor a scaled version of nations (Gardiner et al., 2004) and the meso-level it characterizes is to de duly described. Hence, competitiveness is not simply resulting from a stable macroeconomic framework or entrepreneurship on the micro-level. New patterns of competition are recognizable, especially at regional level: for example, geographical concentrations of linked industries, like clusters, are of increasing importance and the availability of knowledge and technology based tools show high variability within countries. An interesting broad definition of regional competitiveness is the one reported by Meyer-Stamer (2008, pg. 7):
"We can define (systemic) competitiveness of a territory as the ability of a locality or region to generate high and rising incomes and improve livelihoods of the people living there."

This definition focuses on the close link between regional competitiveness and regional prosperity, characterizing competitive regions not only by output-related terms such as productivity but also by overall economic performance such as sustained or improved level of comparative prosperity (Bristow, 2005). Huggins (2003) underlines, in fact, that "true local and regional competitiveness occurs only when sustainable growth is achieved at labour rates that enhance overall standards of living."

The complexity of competitiveness was interestingly decomposed by Esser et al. (1995) into four analytical levels as shown in Fig. 1.1 where different types of determinants drive competitiveness. Apart from the meta level, which regards basic orientations of a society and other 'slow' variables that are not of primary interest here, the micro- meso- and macrolevels of competitiveness are clearly described. The meso-level is between the macro- and
micro-level and aims at designing specific environment for enterprises. At this level it is highly important that physical infrastructure (such as transport, communication and power distribution systems) and sector policies (such as those regarding education and R\&D policies) are oriented towards competitiveness.


Figure 1-1: Determinants of competitiveness at different levels (from Meyer-Stamer, 2008; pag. 3)

As stated in the Sixth Periodic Report on the Region (DG Regional Policy, 1999), the challenge is to capture into a competitiveness index the notion that every region has common features which affect and drive the competitiveness of all the firms located there, even if the variability of competitiveness level of the firms within the region may be very high. These features should describe physical and social infrastructure, the skills of the work force and the efficiency and fairness of the institutions.

The final goal of the present contribution is to develop a competitiveness index for EU NUTS 2 regions which captures all these aspects and describes in synergy the complex nature of economic and social development.

In the following section a review of recent competitiveness indices both at national and regional level is due.

## 2 Literature review

As discussed in the previous section, the complexity in defining competitiveness leads to difficulties in its measurement. Nevertheless, there are examples of well-established studies which apply specific methods for the measurement of the level of competitiveness at national and, more recently, at regional level.

In the following section a brief discussion of selected studies on the theme is provided.
At the country level, the Global Competitiveness Index, prepared by the World Economic Forum (Schwab and Porter, 2007), and the World Competitiveness Yearbook by the Institute for Management Development (IMD, 2008) are by far the most influential and best known indices.

With regards to regional competitiveness, the European Competitiveness Index, computed by the University of Wales Institute, for European regions at the NUTS1 level is discussed (Huggins and Davies, 2006). A simpler but more detailed geographical description of competitiveness is addressed in the very recent 'Altas of Regional Competitiveness' presented in 2007 by the Association of European Chambers of Commerce and Industry (EUROCHAMBERS, 2007), which reflects the international recognition of the importance of analysis at the regional NUTS 2 level. Finally, specific examples of measurement of regional competitiveness in some European countries are given.

### 2.1 The Global Competitiveness Index - World Economic Forum

One of the most known competitiveness indices is the Global Competitiveness Index (GCI), published yearly by the World Economic Forum - WEF (Schwab and Porter, 2007). It covers a large amount of countries, a total of 131 economies in 2007, and is based on over 100 indicators which describe 12 major pillars of competitiveness.

The GCI is intended to measure competitiveness at the national level, taking into account both micro- and macroeconomic foundations of competitiveness. The following definition of competitiveness is the starting point of the WEF index:
"Competitiveness (is) the set of institutions, policies and factors that determine the level of productivity of a country. The level of productivity, in turn, sets sustainable level of prosperity that can be earned by an economy".

The notion of competitiveness implicit in the GCI is, therefore, a mixture of static and dynamic factors including the concept of a country's potential: high levels of current productivity lead to high levels of income and high levels of returns to investment which, in turn, are one of the major determinants of growth potential. This is why a more competitive economy is likely to grow faster over the medium-long run.

## Different dimensions described

To describe the complex notion of competitiveness, the World Economic Forum analyses twelve major pillars (dimensions in statistical terminology) briefly described here.

## 1. Institution

Private individuals, firms and governments interact with each other in an environment created by both private and public institutions. The Institution pillar aims at describing the legal framework, level of bureaucracy, regulation, corruption, fairness in handling public contracts, transparency, political (in)dependence of the judiciary system. The private sector is also represented as private counterpart of the health of an economy.

## 2. Infrastructure

High quality infrastructure is obviously critical for efficient functioning of the economy. The pillar describes roads, railroads, ports and air transport as well as the quality of power supply and telecommunications.
3. Macro-economy

It describes the macroeconomic stability with variables such as government surplus/deficit and debt, saving rate, inflation and interest rate spread.

## 4. Health and primary education

Health of workforce and basic education received by the population are clearly key aspects of a productive and efficient economy. This pillar aims to measure the incidence
of major invalidating illnesses, infant mortality, life expectancy and the quality of primary education.

## 5. Higher education and training

If basic education is the starting point of a ductile and efficient workforce, higher education and continuous training are crucial for economies not restricted to basic process and products. This pillar describes secondary and tertiary education together with the extent of staff training.

## 6. Goods market efficiency

The ideal environment for the exchange of goods is the one which features the minimum of impediments to business activity through government intervention. The three main aspects described by the pillar are: distortions, competition and market efficiency.

## 7. Labour market efficiency

This pillar measures efficiency and flexibility of the labour market, as well as the equity in the business environment between women and men.

## 8. Financial market sophistication

A well-functioning financial sector provides the right framework for business growth and private sector investments. It mainly describes the sophistication of financial market, the easiness for accessing loans, the strength of investor protection and other similar variables.

## 9. Technological readiness

A regulatory framework which is friendly to Information and Communication Technology (ICT) together with ICT penetration rates are of key importance for the overall competitiveness of a nation. Representative variables describing this dimension are for instance internet and mobile telephone subscribers, personal computers, availability of latest technologies and laws relating to ICT.

## 10. Market size

The size of the market determines at which level firms may exploit economies of scale. Firms which operate in large markets have more possibility of exploiting scale economies. Both domestic and foreign markets are taken into account in order to avoid discrimination against geographic areas.

## 11. Business sophistication

This pillar concerns the quality of the business networks of the country and the quality of individual firms' operations and strategies. These aspects are measured using variables on the quality and quantity of local suppliers, the marketing extent and the production of sophisticated unique products.

## 12. Innovation

The pillar refers to technological innovation which, similar to the technological readiness pillar, is a dynamic factor of competitiveness. This pillar is particularly important for more advanced countries which have already reached a higher stage of development. Such countries cannot improve their productivity by 'simply' adopting existing technologies but must invent innovative products and processes to maintain and improve their productivity level.

The 12 pillars taken into account are described by a variety of observable qualitative and/or quantitative variables (indicators). Each pillar is described from a minimum of 2 variables (Market size) to a maximum of 18 variables (Institutions). See Table A. 1 in Appendix A for the complete list.

## Data sources

Indicators used for GCI come from two basic data sources called survey data and hard data. The survey data are drawn from a survey, specifically designed by the World Economic Forum, called Executive Opinion Survey. The survey is completed yearly by over 11,000 top management business executives and gathers qualitative data in order to capture information on a wide range of variables for which sources are scarce or inexistent. With this survey the WEF aims at collecting information not covered by quantitative data provided by official public sources.

Hard data are composed of (quantitative) indicators, such as GDP, number of personal computers or life expectancy, coming from a variety of sources. Examples of data sources are international organizations, such as the International Monetary Fund, the World Bank, United Nations agencies, the International Telecommunication Union, and, when necessary, other sources at national level.

## The role of a country's stage of development

The first step of the aggregating technique for the development of the GCI consists in the definition of the development stage of a country. In fact, different pillars affect different countries in different ways. Three major stages of development are defined.

## 1. Factor-driven economy

At the lower stage of development the economy is called factor-driven and is mainly driven by unskilled labour and natural resources. The first four pillars (Institutions, Infrastructure, Macroeconomic stability, and Health and Primary Education) are the ones which can affect the productivity level at this stage and are thus, included in the factor group.

## 2. Efficiency-driven economy

As countries move along the development path, wages tend to increase and countries can be classified as efficiency-driven. Aspects related to higher education, well-functioning labour markets, large domestic and foreign markets come into play. Pillars from $5^{\text {th }}$ to $10^{\text {th }}$ are included in the efficiency group (Higher education and Training, Goods market efficiency, Labor market efficiency, Financial market sophistication, Technological readiness, Market Size).

## 3. Innovation-driven economy

At the highest level of development countries are defined as innovation-driven. They are able to sustain higher wages only if their businesses are able to exploit the innovation capability of the workforce, developing new products using sophisticated processes. The last two pillars belong to the innovation group (Business sophistication and Innovation).

To take into account the different role various pillars play in the competitiveness definition, GCI developers introduce a weighting scheme for the three sub-indices critical to a particular stage of development.

The stage of development of a country is defined on the basis of two criteria: 1. the level of GDP per capita at market exchange rates; 2. the share of exports of primary goods with respect to total exports of goods and services. The first criterion aims at approximating the wage level of a country, which is not always available worldwide. The second criterion is used to define a threshold: countries which export more than $70 \%$ of primary products are defined to be factor-driven.

Table 1 reports the different weights which are assigned to the three pillar groups (factor, efficiency and innovation groups) and consequently to the countries belonging to each of the different stages of development. Reading the table column by column it is evident that in factor-driven economies basic pillars are assigned the highest weight ( $60 \%$ ), while weights decrease for intermediate and innovation pillars. In countries with efficiency-driven economy, basic and intermediate pillars weight almost equally ( $40 \%$ and $50 \%$, respectively) with innovation pillars weighting $10 \%$. Finally, more innovative economies are assigned the lowest weight to basic pillars ( $20 \%$ ) and weights of $50 \%$ and $30 \%$ to intermediate and innovative pillars.

Table 1: Different weights given to the three pillar groups in countries at different development stages

| Pillar group <br> (sub-index) | Pillars included in <br> the group | Weight for <br> $1^{\text {st }}$ stage $\%$ | Weight for <br> $2^{\text {nd }}$ stage $\%$ | Weight for <br> $3^{\text {rd }}$ stage $\%$ |
| :--- | :--- | :--- | :--- | :--- |
| Factor-driven (basic) | $1-4$ | 60 | 40 | 20 |
| Efficiency-driven <br> (intermediate) | $5-10$ | 35 | 50 | 50 |
| Innovation-driven <br> (innovative) | $11-12$ | 5 | 10 | 30 |

The final index is also tested for sensitivity to different weighting schemes. In short, for each country $i$ the GCI is firstly computed using the weighting scheme of Table $1\left(\mathrm{GCI}_{i}\right)$, then it is computed using more than one million different weighting schemes with weights $\alpha_{1}, \alpha_{2}$ and $\alpha_{3}=1-\alpha_{1}-\alpha_{2} ; \alpha_{1}, \alpha_{2} \in(0,1)$. The steps of the sensitivity analysis are:

1. randomly choose $\alpha_{1}, \alpha_{2} \in(0,1)$;
2. for each country $i$, compute the GCI for the particular (random) weighting scheme in step $1, \operatorname{GCI}_{\alpha, i}=\operatorname{GCI}_{i}\left(\alpha_{1}, \alpha_{2}, 1-\alpha_{1}-\alpha_{2}\right)$;
3. regress $\mathrm{GCI}_{i}$ on $\mathrm{GCI}_{\alpha, i}$ and store the regression goodness of fit $\mathrm{R}^{2}$;
4. repeat steps 1-3 (in the specific case over one million of regressions are computed).

For the 2007 GCI, the analysis shows that the index is not very sensitive to the actual numbers used for weighting the three super-pillars.

In addition to the differential weighting procedure, GCI authors adopt a moving average technique with the aim of improving robustness of the data. For each indicator the weighted average of the country average response in 2007 and 2006 is computed. This should improve the stability of responses and reduce the impact of random variations in the sample. For details, see the following section.

The definition of different development stages is a very interesting approach which will also be adopted for the setting-up of the EU Regional Competitiveness Index, as will be illustrated in Sect. 3.12.

## Computation of GCI

Each indicator $q_{i}$ is rescaled on a 1-7 scale ${ }^{1}$. Let $c$ denote the country, while $T_{1}$ and $T_{2}$ denote the two years of interest $\left(T_{1}=2006, T_{2}=2007\right)$. Then for country $c$ each indicator is computed as:
$q_{i, c}^{T_{1} T_{2}}=w_{c}^{T_{1}} \times \bar{q}_{i, c}^{T_{1}}+w_{c}^{T_{2}} \times \bar{q}_{i, c}^{T_{2}}$
where
$\stackrel{-\bar{q}_{i, c}}{T_{j}}=\frac{1}{N_{c}^{T_{j}}} \sum_{k=1}^{N_{c}^{T_{j}}} q_{k, i, c}^{T_{j}} \quad j=1,2$
$N_{c}^{T_{j}}=$ sample size in country $c$ at time $T_{j}$
$q_{k, i, c}^{T_{j}}=$ response of unit $k$ for indicator $i$ in country $c$

[^0]If the indicator value is the same for the whole country (as for indicators from hard data) there is no need to compute the country average $\bar{q}_{i, c}$.

Weights $w_{c}^{T_{2}}$ and $w_{c}^{T_{2}}$ are defined according to a certain criterion which will not be detailed here (for further details see Schwab and Porter, 2007, pg. 96).

Let $\bar{q}_{c}^{m}$ be the average value for $q_{c}^{T_{1} T_{2}}$ computed for all the indicators describing pillar $m$ ( $m=1 . . .12$ ). ${ }^{2}$ Each pillar is then grouped into macro-pillars according to the development stage of the country as previously described. Macro-indicators for basic-, efficiency and innovation-driven economy are then computed as:
$Q_{c}^{\text {basic }}=\frac{1}{4} \sum_{m=1}^{4} \bar{q}_{c}^{m}$
$Q_{c}^{\text {efficiency }}=\frac{1}{6} \sum_{m=5}^{10} \bar{q}_{c}^{m}$
$Q_{c}^{\text {innovation }}=\frac{1}{2} \sum_{m=11}^{12} \bar{q}_{c}^{m}$

The final score is computed as the weighted average of $Q_{c}^{\text {basic }}, Q_{c}^{\text {efficiency }}$ and $Q_{c}^{\text {innovation }}$ with weights depending on the development stage of the country according to Table 1.

### 2.2 World Competitiveness Yearbook - Institute for Management Development

The World Competitiveness Yearbook (WCY) is an annual report on the competitiveness of countries, published since 1989 by the Institute for Management Development (IMD), a not-for-profit foundation located in Switzerland (IMD, 2008). It analyses and ranks the ability of countries to create and maintain an environment which sustains the

[^1]competitiveness of enterprises. The 2008 report covers 55 countries, chosen on the basis of their impact on the global economy and the availability of comparable international statistics. The WCY identifies four main competitiveness pillars (factors): economic performance, government efficiency, business efficiency and infrastructure. Each of these pillars is broken down into five sub-pillars (sub-factors) which describe different facets of competitiveness, for a total of 20 sub-pillars.

In the following section each pillar is discussed.

## Different dimensions described

The four competitiveness pillars identified by the WCY are:

## 1. Economic performance

## 2. Government efficiency

3. Business efficiency

## 4. Infrastructure

The Economic Performance pillar is comprised of 80 variables (criteria) and describes the macroeconomic evaluation of the domestic economy. In particular, it focuses on the following sub-pillars: domestic economy, international trade, international investment, employment, prices.

The Government Efficiency pillar is comprised of 73 variables and describes the extent to which government polices are conducive to competitiveness. Its sub-pillars are public finance, fiscal policy, institutional framework, business legislation, societal framework.

The Business Efficiency competitiveness pillar is comprised of 70 variables and describes the extent to which the national environment encourages enterprises to perform in an innovative, profitable and responsible manner. Its sub-pillars are productivity, labor market, finance, management practices, attitudes and values.

The Infrastructure competitiveness pillar is comprised of 108 variables and describes the extent to which basic, technological, scientific and human resources meet the needs of business. Its sub-pillars are basic infrastructure, technological infrastructure, scientific infrastructure, health and environment and education.

A detailed list of all variables included in each of the pillars is found in Table A. 2 of the Appendix A.

## Data sources

The data used for the construction of the WCY is a combination of quantitative (hard) and qualitative data (survey). Hard data consist of statistical indicators acquired from international, national and regional organizations, private institutions and the WCY network made of 55 partner institutions. Survey data are drawn from the WCY annual Executive Opinion Survey data sent to executives in top and middle management in all of the economies covered by WCY. The survey is compiled by a panel of 4000 executives from a representative cross-section of the business community in each country. The hard data represents $2 / 3$ of the overall weight in the final rankings while survey data are assigned a weight of $1 / 3$.

## Computation of WCY

There are a total of 331 variables in the WCY of which 254 are used to calculate the Overall Competitiveness rankings. The Standard Deviation Method (SDM) is used in order to obtain a comparable standard scale for computing the overall, pillar and sub-pillar results.

To this aim, for each of the 254 variable the standardized value (STD) is computed:
$\operatorname{STD}(x)=\frac{x-\bar{x}}{S}$
where:
$x=$ original value
$x=$ average value of the 55 countries
$S=\operatorname{standard}$ deviation of $x$

The sub-pillar rankings are obtained by computing the weighted average of the STD values for all variables which make up the given sub-pillar. The survey data variables, coming from the Executive Opinion Survey, are weighted so that they account for one-third in the determination of the overall ranking.

In case of missing data for a particular country, the missing values are replaced by a STD value equal to 0 .

The sub-scores of each sub-pillar are then aggregated in order to obtain the pillar score. Each sub-pillar, independently of the number of variables it contains, is assigned an equal weight of $5 \%$ on the overall score. ( 20 sub-pillars x $5=100$ )

The STD values of each of the four pillars are aggregated to determine the overall score as the average of the four pillars' scores. The number is then converted into an index with the leading economy given a value of 100 .

One of the major differences between the WCY by IMD and the GCI by WEF, described in Section 2.1, is that, first, a higher number of variables are comprised in the WCY and, second, the latter puts more emphasis on survey data while the WCY focuses more on hard statistics. Hard data availability is, in fact, the reason why WCY can cover a lower number of countries (55) with respect to those covered by the GCI (131). On the other hand, survey data are considered by IMD less reliable since they are entirely based on subjective opinion (IMD, 2008).

### 2.3 The European Competitiveness Index - University of Wales Institute, Cardiff-UWIC

Currently two editions of the Robert Huggins Associates' European Competitiveness Index (ECI) are available, issued in 2004 and 2006. The index' main purpose is to measure, compare and examine the competitiveness of regions and nations.

The 2004 edition of the ECI comprised EU-15 member states as well as Norway and Switzerland, and their regions at the NUTS-1 level The 2006 ECI has been expanded to include EU- 25 countries and their respective NUTS-1 regions, in total 116 regions plus Norway and Switzerland (Huggins and Davies, 2006).

The focus on regions reflects and confirms the growing consensus on the relevance of regions as key territorial units for economic analysis. It is well-established that the geographic concentration of specialized inputs, employees, information and institutions favors firms and industries especially in the most advanced economies. This process feeds off itself: the localized productivity advantages of agglomeration push firms to cluster and reinforce these
clusters over time. Thus, as globalization tends to nullify traditional forms of advantages, the business environment where firms are located becomes more and more important. In this sense "globalization is reinforcing localization" (Huggins and Davies, 2006 pg. 4).

The ECI takes into account three major pillars: creativity, economic performance and infrastructure/accessibility. Two additional pillars, education and knowledge employment, are separately analyzed at regional level in order to ascertain their correlation with the ECI. They are in fact considered as respectively cause and effect of competitiveness rather than its direct measure. The underlying assumption is twofold: i) highly educated population is a key ingredient for business performances; ii) regions which are competitive in terms of creativity, economic performance and accessibility also tend to host high value-added and knowledgeintensive employment. Correlating education expenditure/enrolments with ECI gives an insight into which regions are most effective in converting human capital resources into economic outcomes. Correlation of knowledge employment with ECI gives an insight into which areas are effective in turning their potential into actual high level employment.

In the next Section the dimensions used in the ECI report are detailed.

## Different dimensions described

Five different groups of variables are included in the ECI report, but only the first three are included in the computation of the composite ECI:

1. Creativity
2. Economic Performance
3. Infrastructure and Accessibility
4. Knowledge Employment
5. Education

The Creativity dimension is described by 8 quantitative variables mainly related to $\mathrm{R} \& \mathrm{D}$ employment and expenditure by sector. The list of variables is shown in Box 1.

```
    Box 1: Creativity variables
    (source: Huggins and Davies, 2006, pg. 2)
Creativity
R8D Employment in the Business Sector per 1,000 Inhabitants
R&D Employment In the Government Sector per 1,000 inhabitants
R8D Employment In the Higher Education Sector per 1,000
Inhablants
R8D Expenditure In the Business Sector per capita
R&D Expenditure In the Govemment Sector per caplta
R8D Expenditure In the Higher Education Sector per caplta
Number of Patent Applications per 1 million Inhabitants
Employment In ICT Services per 1,000 Inhabitants
```

Economic performance is described by GDP, monthly earnings, rates of productivity, unemployment and economic activity (Box 2).


Quantitative data related to motorways, railways and air transportation of both passengers and freight are considered to describe the transport and infrastructure density. Two variables related to ICT usage, Broadband lines and Secure Servers, are only available at national level (Box 3).

## Box 3: Infrastructure and Accessibility variables (source: Huggins and Davies, 2006 pg. 2)

Infrastructure and Accessibility
Motorway Length per kmp
Motorway Length per venicle
Length of Rallway per km ?
Air Freight Disembarked per 1,000 inhabitants
Alr Passengers Disembarked per 1,000 inhabitants
Number of Vehlcies per 1,000 inhabitants
Broadband Access Lless per 1,000 Inhabitants (national data only)
Secure Servers per 100,000 inhabtiants (national data only)

These three groups of variables form the core for the composite index computation. The methodological approach is detailed in later on in this section.

After the ECI computation, further analysis is provided in the report to get an insight into the level of knowledge economy that can be observed in regions. To this purpose the proportion of knowledge-based employment and the level of education of the population are related to regional ECI.

Knowledge-based employment is described by employment (per 1000 inhabitants) and number of business units (per 1 million inhabitants) by nine sectors, as indicated in Box 4.

## Box 4: Knowledge employment sectors

Biotechnology and Chemical
ICT Services
Research and Development
IT and Computer Manufacturing
Telecommunications
Machinery and Equipment Manufacturing
Instrumentation and Electrical Machinery
Automotive and Mechanical Engineering
High-Technology Services

The correlation between ECI and Education is based on aggregate data for the number of students per 1000 employees enrolled in secondary and tertiary education, as well as data for secondary and tertiary education at national level (the authors consider data on education expenditure not reliable at the regional level). The choice of aggregating different types of education is driven by the difficulty in comparing data across specific categories of education since the method for students' classification is not homogeneous across countries. Variables for this pillar are listed in Box 5 .

## Box 5: Education

Number of Students in Upper Secondary Education per employed person
Number of Students in Academic Tertiary Education per employed
person
Secondary Education Expenditure per Capita (national data only)
Tertiary Education Expenditure per Capita (national data only)

## Data sources

Data comes from different European Institutions, such as Eurostat and DG Regional Policy, as well as country specific organizations. The complete list of data sources is shown in Table A. 2 of Appendix A.

## Computation of ECI

For the computation of the composite index, data is first standardized. Afterwards, a Factor Analysis (FA) is performed on the whole set of variables in order to extract communalities which represent the common part of variation of the dataset. The "image factoring" is employed as extraction method and the varimax is used to obtain optimally rotated factors. The scores of each region for the common dimensions are interpreted as sub-composite indices. Finally, a single composite is derived from FA sub-indices using Data Envelopment Analysis - DEA (Cherchye, 2001). DEA is a linear programming tool which estimates an efficiency frontier used as a benchmark to measure the relative performance of countries. DEA computes a benchmark (the frontier) and measures the distance between units (regions in this case) and the frontier. The benchmark can be obtained as the solution of a maximization problem or by external definition. In a DEA solution each unit (region) is assigned a set of weights which depend on the distance of the unit from the frontier. Note that both weights and the frontier are country specific and in general there would be no unique frontier (OECD, 2008).

By DEA each region receives a score between 0 and 1 for each sub-composite index. For each region, a composite score is then computed as the geometric mean of all the DEA scores for that region. These scores are finally indexed round the European average giving the ECI.

## Further analysis

To explore the assumption of a positive relation between the competitiveness level of a region and its level of knowledge-intensive employment, a correlation analysis between ECI and employment indicators is performed. The strength of this relation is computed with
respect to an index of total knowledge employment ${ }^{3}$ and to knowledge employment indices separated by sectors. Of the knowledge employment sectors only ICT services are included in the composite ECI so as only a small endogenous correlation effect is expected.

Similarly, the correlation between ECI and education expenditure and enrolments is computed. The ECI versus expenditure analysis is performed at national level whilst ECI versus enrolment analysis is performed at regional level.

### 2.4 The Atlas of Regional Competitiveness - Eurochambers

The Association of European Chambers of Commerce and Industry has recently published a study which measures and compares regional competitiveness of the 268 EU regions at NUTS2 level (EUROCHAMBERS, 2007). Competitiveness is measured in terms of seven main pillars described by reference indicators. For each Member State and indicator the best performing region is singled out. The result is a comparison of the best performing regions of the 27 Member States.

No composite indicator is computed; instead comparison of regions is discussed separately for each indicator. In this sense the analysis can be seen as a partial view of EU competitiveness as it describes only excellence within each EU country with respect to each dimension.

Despite its simplicity the Atlas of Regional Competitiveness provides a relevant example of competitiveness measurement at a very detailed geographical level giving valuable suggestions for the selection of indicators in the analysis at the NUTS2 level.

## Different dimensions described

Seven dimensions (pillars) have been selected for analysis:

1. Economic Performance
2. Employment and Labour Market
3. Training and Lifelong learning
4. Research and Development/Innovation

[^2]5. Telecommunication Networks
6. Transport

## 7. Internationalization

For each dimension a reference indicator is chosen and used for separate comparison of EU regions. Descriptive analysis of other related indicators is provided as well.

The Economic Performance is described by means of GDP per capita in Purchasing Power Standard (PPS). A closer look into the economic background is provided by separate analysis of GDP growth rate in 2004 and average annual growth rate between 2000 and 2004. A description of total regional GDP by three sectors (Agriculture, Forestry and Fishing / Industry / Services) is also discussed.

The reference indicator for Employment and Labour Market is the employment rate, taken as the rate of number of individuals aged $15-64$ in employment and the total population of the same age group. The indicator is based on the Eurostat Labour Force Survey. Related descriptive analysis is based on unemployment rate (percentage of unemployed persons in the active population), long-term unemployment (people unemployed for not less than twelve months) and average of hours worked per week. Employment is also analyzed by sector (same three sectors as for Economic Performance).

For the third dimension on Training and Lifelong learning, the reference indicator is the education attainment, classified as the percentage of the population with a higher degree ${ }^{4}$. Further analysis is carried out considering the proportion of students in higher education compared to the entire student population and the rate of $25-64$ years age group having received training in the past twelve months, as an indicator of lifelong learning.

The Innovation dimension is described by the number of patent applications to the EPO per million inhabitants. The indicator is supposed to reveal the dynamism of the R\&D sector of a region and can be regarded as an output indicator. It is worthwhile to note that when new Member States are considered, the indicator can give rather distorted results since new Members have no tradition of applying for patents with the EPO. Such a comparison could then disadvantage those countries.

[^3]In case of data availability, $R \& D$ expenditure as percentage of GDP and R\&D staff as percentage of active population are analyzed both totally and by three sectors (Enterprises, Public Sector and Higher Education).

The reference indicator for Telecommunications Networks is the percentage of households and enterprises which have access to internet. Additionally, the analysis of patent applications in the field of telecommunications is provided as an indicator of regional dynamism in the field.

The Transport pillar is the only one described by multiple indicators. Specifically:
a. Motorway length and density, in terms of length per million inhabitants;
b. Airfreight transport, in terms of total goods loaded and unloaded;
c. Maritime freight, in terms of total goods loaded and unloaded.

Finally, the last pillar Internationalization lacks data at the regional level. This theme has been described only at country level in terms of the following indicators:
a. Exports and Imports by product type and with respect to population size;
b. Average annual growth rate of exports/imports between 2000 and 2004;
c. Incoming Foreign Direct Investment - FDI stocks both in absolute value and as a percentage of GDP;
d. Average of incoming and outgoing flow of FDI in relation to GDP.

## Data sources

Data has been extracted from Eurostat and refers to the last available year in September 2007. Figures related to the use of internet by households and enterprises have been taken from the European Spatial Planning Observation Network - ESPON (http://www.espon.eu/).

### 2.5 Country specific regional indices

Besides international studies on regional competitiveness, in the past years several country specific analyses on the topic were published. Three cases have been selected for discussion: the United Kingdom, Croatia and Finland. They represent valid attempts to describe regional competitiveness with an overall perspective and sound methodology.

## United Kingdom

The United Kingdom has a long tradition in competitiveness studies which is testified by the UK Competitiveness Index reports, first introduced and published in 2000. The 2008 edition represents a benchmark of the competitiveness of the UK's regions and localities (Huggins and Izushi, 2008). The concept of competitiveness adopted regards the development/sustainability of businesses and the economic welfare of individuals. Competitiveness is in fact defined as "the capability of an economy to attract and maintain firms with stable or rising market shares in an activity, while maintaining stable or increasing standards of living for those who participate in if' (Huggins and Izushi, 2008; pg. 7).

Competitiveness of a region is viewed as the result of a complex interaction between input, output and outcome factors. To this aim, the UK Competitiveness index comprises a series of indicators incorporating data that are available and comparable at the regional level (NUTS1) and at a very detailed local area level.

The conceptual framework underlying the index for regional competitiveness is a 3 -factor model (Huggins, 2003) as shown in Box 6. Three major dimensions (factors here) are described with the indicators listed in Box 6 and are assigned different meanings. The input variables, such as firms per 1000 inhabitants and proportion of knowledge-based businesses, are assumed as contributing to the output productivity of a region, which is described in the output dimension. The impact of the input and output factors is given by the level of average earnings and the unemployment rate, which are considered as the only tangible outcomes.

Each of the three dimensions is assigned equal weight in the composite computation, i.e. each dimension has a weight of 0.333 . Further, within each dimension this weight is equally distributed among the indicators. This means that, for instance, the two indicators describing the Outcome dimension are assigned a weight of $0.333 / 2$ each. Three sub-indices are then computed.

Before computing the overall composite, each sub-index is transformed into its logarithmic form to dampen out extremes which may distort the final composite score. Afterwards the
composite score is finally anti-logged through exponential transformation in order to reflect as far as possible the scale of difference in competitiveness between regions.

| Box 6: UK Regional Competitiveness Index <br> Framework <br> (source: Huggins and Izushi, 2008, pg. 9) |
| :---: |
| Input factors |
| R\&D expenditure <br> Economic Activity Rates <br> Business Start-up Rates per 1,000 Inhabitants <br> Number of Business per 1,000 inhabitants <br> GCSE Results - 5 or more grades $\mathrm{A}^{*}$ to $C$ <br> Proportion of Working Age Population with NVQ Level 4 or <br> Higher <br> Proportion of Knowledge-Based Business |
| Output factors |
| Gross Value Added per head at current basic prices <br> Exports per Head of Population <br> Imports per Head of Population <br> Proportion of Exporting Companies <br> Productivity- Output per Hour Worked Employment Rates |
| Outcome factors |
| Gross weekly pay unemployment rates |

The analysis is carried out for the 12 UK regions at NUTS1 level and for 408 local areas.

## Croatia

The Croatian National Competitiveness Council and the Croatian Chamber of Economy recently published the first edition of "The Regional Competitiveness Index of Croatia, 2007" (UNDP, 2008). The definition of competitiveness adopted is the one by the World Economic Forum which defines competitiveness as "a range of factors, policies and institutions which determine the level of productivity" (Schwab and Porter, 2007).

The report is based on the methodologies of the World Economic Forum and the National Institute for Management Development. It provides an insight into the competitiveness of Croatia's regions by evaluating the quality of the business sector and business environment. The focus is, thus, specifically on the measurement of the business aspect of competitiveness. The underlying assumption is that wealth is primarily generated at the enterprise level and that the environment in which the enterprise operates can either support or disturb its ability to compete.

The analysis is carried out for the three NUTS 2 regions, newly defined in Croatia, in accordance with the principles of Eurostat, as well as for Croatian counties at the NUTS 3 level.

Two main economic areas are described - the business environment and the quality of the business sector - and are the result of 135 indicators structured into eight sub-groups, as indicated in Box 7. For the complete list of selected indicators, see the entire report which is freely available on-line at www.undp.hr. Most of the indicators are expressed as numbers per person, as an activity trend (index) over several years, or as a percentage.

Indicator values derive from numerous statistical as well as survey data, with a proportion of survey to statistical data of about one third. Statistical data is of quantitative type whilst survey data is of qualitative type. Survey data is analyzed on the basis of the Business Competitiveness Index by the World Economic Forum. Statistical indicators are, instead, analyzed using the International Institute for Management Development - IMD - approach.

```
Box 7: Major pillars of the Croatian Regional Competitiveness Index 2007 (source: UNDP, 2008)
```

```
A. Business environment
```

A. Business environment
Demographics, health and culture
Demographics, health and culture
Education
Education
Basic infrastructure and public sector
Basic infrastructure and public sector
Business infrastructure
Business infrastructure
B. BusIness sector
B. BusIness sector
Investments and entrepreneurial trends
Investments and entrepreneurial trends
Level of entrepreneurship development
Level of entrepreneurship development
Economic results - level
Economic results - level
Economic results - trends

```
    Economic results - trends
```

Qualitative data obtained from surveying entrepreneurs' opinions is used to set up two subindices for the business sector and the business environment following the WEF approach. Analogously, statistical data is used to set up two quantitative sub-indices following the IMD methodology.

For survey data, a seven categories measurement scale is adopted. The calculation of subindices for the business sector and the business environment is carried out using exact weights for individual questions as recommended in the WEF methodology.

The quantitative analysis was based on the IMD methodology using more than one hundred indicators to calculate sub-indices adopting an equal weight scheme. These sub-indices were subsequently used in the calculation of the two main indices, weighting equally the subindices.

In the end, each region receives four scores: two survey and two statistical scores for the two business areas. Then two basic indices, survey and statistical, are computed as weighted averages of the two sub-indices. Different weights are given to the business environment and to the quality of the business sector: a greater weight to the former - 0.844 - and a smaller to the latter -0.166 . The weights are computed based on the WEF method.

Finally, the overall regional competitiveness index is computed as the average of the survey and statistical indices, after standardization.

## Finland

The Finnish case (Huovari et al., 2001) represents a relevant example of competitiveness measurement at a very detailed geographical level (NUTS4). The definition of regional competitiveness adopted in the study is "the ability of regions to foster, attract and support economic activity so that its citizens enjoy relatively good economic welfare". Authors recognize that, despite the existence of well established international studies on competitiveness, they cannot be applied as such to a regional framework since some of the indicators used at country level are either unavailable or meaningless at the regional level. For example indicators which represent the efficiency of public sector or barriers to foreign trade do not vary within a country and are then considered inadequate for regional comparing, especially when a single country is investigated.

In the Finnish case the index is set-up using available indicators at the labour market level as well as indicators which measure the innovativeness and agglomeration of regions. Specifically, four dimensions of competitiveness are defined:

1. Human Capital
2. Innovativeness
3. Agglomeration
4. Accessibility

These four major dimensions are described by 16 variables (indicators) at the NUTS 4 level for a total of 85 Finnish sub-regions.

It is interesting to note that no indicator related to economic performance has been included in the index. In fact, indicators of economic performance and well-being, such as per capita GDP and personal income, have been included afterwards via a study of correlation between them and the competitiveness index. The association between the index and short-term outcome indicators, i.e. change in production, employment and population, has been assessed as well. In this sense the measure of competitiveness given here is related to a larger extent to the potential and innovativeness of the region than to its actual economic productivity. The study represents a peculiar view of regional competitiveness which greatly differs from the more common perception of business competitiveness.

Human capital is measured by means of 5 variables: number of highly educated residents; total number of students; number of technical students; size of the working age population (15 64); participation rate in the labour market.

Innovativeness is captured by 4 variables: average of the number of patents between 1995 and $1999^{5}$; R\&D expenditures; proportion of establishments which have been innovative during the years 1985 and $1998^{6}$; proportion of value added produced in high technology sectors. Agglomeration of firms and economic activity is described with 4 indicators: population density; proportion of workers in sectors where external economies are large (manufacturing,

[^4]wholesale, retail trade and private services); proportion of workers in business services; size of the largest sector within the sub-region.

Three variables measure Accessibility: road distance of each sub-region to every other, weighted by the size of the sub-region; distance from airports, weighted by the size of airports; proportion of firms in a sub-region engaged in foreign trade. It should be noted that rail accessibility has not been taken into account because of data availability at subregional level and also because of the dominant role of road and air accessibility for the trade of goods.

To set-up the index all variables are firstly weighted with the relative size of the sub-region with respect to the population. Selected variables are of two types: one comprises variables expressed as absolute numbers, such as number of students; the other comprises variables expressed as proportions, such as proportion of workers in a sector. The weighting method differs for the two types of variables:
$V_{i}=100 \frac{x_{i} / X}{p_{i} / P} \quad$ for type I variables
$V_{i}=100 x_{i} / X \quad$ for type II variables
where $x_{i}$ is the value of variable $x$ for sub-region $i, X$ the value of variable $x$ for the whole country, $p_{i}$ is the number of inhabitants of sub-region $i, P$ is the number of inhabitants of the whole country.

Standardization is then applied to indicators which generally show high differences in standard deviations.

For each dimension the average sub-index is computed, with equal weights, and the overall competitiveness index is the simple average of the four sub-indices, each with weight 0.25 .

## 3 Developing the RCI: theoretical framework

The main goal of the EU Regional Competitiveness Index (RCI) is to map economic performance and competitiveness at the NUTS 2 regional level for all EU Member States. The expected results are of great variation within each country, with regions with low levels of competitiveness located among strongly competitive regions. Furthermore, a higher degree of heterogeneity is foreseen due to the accession of the 12 new Member States.

The aim of the project is to develop a rigorous method to benchmark regional competitiveness and to identify the key factors which drive the low competitiveness performance of some regions. To this purpose RCI should present an overall but synthetic picture of regional competitiveness.

On the basis of existing competitiveness studies discussed in Section 2, an ideal framework for RCI is proposed which includes eleven major pillars. The reference is the well-established GCI by the WEF (Section 2.1) but some variations and adaptations have been considered necessary in order to address the regional dimension of RCI. The main differences between RCI and WEF-GCI are: a) the application of a regional as supposed to country level analysis; b) the exclusion of two pillars (Goods market efficiency and Financial market sophistication); c) the division in two separate pillars of the GCI Health and Primary education pillar; and d) the preference towards hard (quantitative) data with respect to survey data.

The reason for the exclusion of the Goods market efficiency pillar is related to the fact that EU regions are subject to the single market and the customs union. The pillar is then expected to show little if any variation across the EU. Moreover, some of the indicators selected by WEF to describe this pillar have been included in the RCI Institutions pillar (ex. World Bank Ease of Doing Business Index).

Little variation across EU is also expected for the Financial market sophistication pillar. In addition, only few hard data are available to describe this aspect for the EU. These have been the reasons behind the choice of excluding the pillar from the RCI framework as well. The pillars included in the RCI framework are listed in Box 8 .

Box 8: The RCI-2010 framework
RCI pillars

1. Institutions
2. Macroeconomic Stability

## 3. Infrastructure

4. Health

## 5. Quality of Primary and Secondary Education

## 6. Higher Education/Training and Lifelong Learning

7. Labour Market Efficiency
8. Market Size

## 9. Technological Readiness

## 10. Business Sophistication

## 11. Innovation

With respect to the WEF framework, the pillar Health and Primary Education has been slightly modified and split into two different pillars to better distinguish between two distinct aspects of regional competitiveness across the EU. Health - pillar 4 - is described at the regional level while Quality of Primary and Secondary Education - pillar 5 - is described at the country level in terms of achievements and skills of pupils of age 15. In fact, the compulsory education system in force in the EU fixes to either 15 or 16 the ending age of compulsory education for most countries, with the exception of Hungary and the Netherlands where the minimum age is 18 .

Pillars may be grouped according to the different dimensions (input versus output aspects) of regional competitiveness they describe. Figure 3-1 shows the classification chosen for the RCI. The terms 'inputs' and 'output' are meant to classify pillars into those which describe driving forces of competitiveness, also in terms of long-term potentiality, and those which are direct or indirect outcomes of a competitive society and economy.


Figure 3-1: Interpretation of the pillars included in the ideal framework for RCI. ${ }^{7}$

As already mentioned, the indicators selected for the RCI framework are all of quantitative type (hard data) and the preferred source has been Eurostat. Whenever information has been unavailable or inappropriate at the required territorial level, other data sources have been explored such as the World Bank, Eurobarometer, OECD, the European Cluster Observatory.

[^5]Candidate indicators for each pillar are discussed in the current section. The following basic criteria for the initial selection of candidate indicators within each pillar have been applied:

1. experts' opinion and literature review;
2. elimination of overlapping information across pillars;
3. balanced number of indicators across pillars.

The complete list of candidate indicators is listed in Appendix C. The final list of indicators included in the RCI is a subset of the candidate indicators. As it will be detailed in Chapter 4 and 5 , two additional criteria have been used to refine the candidate list and arrive at the final choice of the suite of included indicators from those belonging to the ideal framework. :
4. data availability (in terms of missing data - Section 4.2);
5. statistical consistency (multivariate analysis - Section 4.4).

In some cases, applying all criteria has not been possible due to the complex structure of the index and that is why, for example, not all pillars are populated with roughly the 'same' number of indicators.

The following sections provide an overview of each pillar, its relevance in terms of regional competitiveness, the specific aspects to be measured within it and the set of indicators selected to this aim. In the discussion below, we will limit ourselves to outlining only the candidate indicators and their source. Appendix C provides detailed information on the geographical level, unit of measurement and periodicity of all potential indicators.

### 3.1 Institutions

## Why does it matter?

The importance of institutions for economic growth has gained increasing attention in the last decades in search of additional factors impinging on economic development beyond traditional growth theories (Rodriguez-Pose, 2010 and Storper, 2005). Rodrik et al. (2004) go as far as claiming that the quality of institutions is more important than traditional development factors such as geography in determining levels of income and growth prospects. Effective institutions have a number of positive impacts on the competitiveness of a country/region. In an overview of the academic literature on the subject, Rodriguez-

Pose (2010) points out that they improve the provision of public goods, address market failures, improve efficiency (Streeck, 1991), reduce transaction costs (North, 1990), foster transparency (Storper, 2005), promote entrepreneurship and facilitate the functioning of labour markets. Effective local institutions provide the adequate conditions for investment, economic interaction and trade, while reducing the risk of social and political instability (Jütting, 2003). Putnam (2000) points out that solid institutions are the key enables of innovation, mutual learning and productivity growth and puts them as the core of the factors driving economic growth.

The pillar Institutions aims at measuring the quality and efficiency of institutions, the level of perceived corruption and the general regulatory framework within countries. It tries to give an insight into how favorable is the institutional climate for enterprises, how easy it is to open a new business, how much trust people have in their national legislative and regulatory systems and its effectiveness.

There is not much agreement in the academic literature as to the best way of including indicators of institutional quality within competitiveness indices in general and even more so within regional competitiveness indicators. The GCI includes in its institutional pillar private and public institutions with a focus on both firm-level and public implications. The ECI puts as important factors of the institutional structure social capital and the efficiency and effectiveness of the public administration. All of these aspects, however, are not easily measured quantitatively so that to allow for a cross-country comparison. Their variability on regional level is also somewhat problematic as they describe national contexts which hardly present significant differences on the regional level.

Given the fact that regional indicators describing these aspects for EU regions have not been identified, we have opted for using country level data. Even though it does not carry any message as to the variability in the quality of institutions at the regional level, we have chosen to still include this pillar as any description of competitiveness, regardless of the level, needs to take into account the quality and efficiency of institutions as an essential determinant of economic growth.

Given the intrinsic features of the pillar, we propose some indicators which measure citizens' perception of the quality of the institutions. To this aim we considered two recent Eurobarometer studies which offer information on EU 27 citizens' perception of corruption
and fraud in their home countries (European Commission, 2009b and 2008). The former is a Special Eurobarometer issue and refers to fieldwork carried out in September-October 2009; the latter is a Flash Eurobarometer and refers to a survey carried out in June 2008.

Further, we have taken into account the Worldwide Governance Indicators (WGI) project (http://info.worldbank.org/governance/wgi/index.asp), which is one of the most wellknown databases describing the quality of institutions. It reports aggregate and individual governance indicators for 212 countries and territories over the period 1996-2007, for six dimensions of governance: a. Voice and Accountability; b. Political Stability and Absence of Violence; c. Government Effectiveness; d. Regulatory Quality; e. Rule of Law and f. Control of Corruption. The aggregate indicators combine the views of a large number of enterprises, citizens and expert survey respondents in industrial and developing countries. The individual data sources underlying the aggregate indicators are drawn from a variety of survey institutes, think-tanks, non-governmental and international organizations. It is important to note that these are composite indicators whose raw data variables in most cases are not readily accessible. For the RCI we have considered the aggregate indicators which are measured in units ranging from -2.5 to 2.5 , with higher values corresponding to better governance outcomes. Data have been extracted from the official website: www.govindicators.org. More details on the World Bank indicators may be found in Kaufmann et al. (2009).

We also propose to include one indicator from the Doing Business 2010 report by the World Bank (www.doingbusiness.org). The Doing Business project, launched 8 years ago, looks at domestic small and medium-size companies and measures the regulations applying to them through their life cycle. It provides a quantitative measure of regulations for starting a business, dealing with construction permits, employing workers, registering property, getting credit, protecting investors, paying taxes, trading across borders, enforcing contracts and closing a business-as they apply to domestic small and medium-size enterprises. A fundamental premise of Doing Business is that economic activity requires good rules. These include rules that establish and clarify property rights and reduce the costs of resolving disputes, rules that increase the predictability of economic interactions and rules that provide contractual partners with core protections against abuse. The Doing Business 2010 covers the period June 2008 through May 2009. Economies are ranked on their ease of doing business, from 1 - 183, with a high ranking on the ease of doing business index meaning that the
regulatory environment is conducive to the operation of business. This index averages the country's percentile rankings on 10 topics, made up of a variety of indicators, giving equal weight to each topic.

Box 9 shows the set of eleven candidate indicators proposed to describe the Istitutions pillar. The six governance indicators (from $5^{\text {th }}$ to $10^{\text {th }}$ ) belong to the set of World Bank Worldwide Governance Indicators. They are measured in units ranging from -2.5 to 2.5 , with higher values corresponding to better governance outcomes.

The last indicator, from Doing Business 2010, has been reversed to be positively related to the level of competitiveness of the country.

| Box 9: Indicators for Institution |  |  |
| :---: | :---: | :---: |
| Data Source |  | Indicator description |
| 1. | Special Eurobarometer 325 | Corruption as a major problem at the national level |
| 2. |  | Corruption as a major problem at the regional level |
| 3. | Flash Eurobarometer 236 | Perceived extent to which the state budget is defrauded (customs fraud, VAT fraud, fraud with subsidies, etc.) |
| 4. |  | Perceived extent of corruption or other wrongdoing in the national government institutions |
| 5. | World Bank Worldwide Governance Indicators | Voice and accountability |
| 6. |  | Political stability |
| 7. |  | Government effectiveness |
| 8. |  | Regulatory quality |
| 9. |  | Rule of law |
| 10. |  | Control of corruption |
| 11. | Doing Business 2010 | Ease of doing business |

### 3.2 Macroeconomic stability

Why does it matter?

Macroeconomic stability measures the quality of the general economic climate. Economic stability is essential for guaranteeing trust in the markets both for consumers and producers of goods and services. Stable macroeconomic conditions lead to higher rate of long-term investments and are essential ingredients for maintaining competitiveness.

We propose a set of indicators similar to the ones chosen by WEF for the GCI, with the exception of the 'interest rate spread' that is included in the GCI but is not available for EU countries. On the basis of experts' opinion we have replaced this indicator with the government long term bond yields which measures the trust of the market in the country.

The candidate indicators for this pillar are listed in Box 10. They are all measured at the country level as the aspects captured by the pillar are intrinsically national.

Box 10: Indicators for Macroeconomic Stability

| Data source |  | Indicator description |
| :---: | :---: | :---: |
| 1. | Eurostat | General government deficit (-) and surplus (+) |
| 2. |  | Income, saving and net lending / net borrowing |
| 3. |  | Annual average inflation rate |
| 4. |  | Long term bond yields |
| 5. |  | General government gross debt |

### 3.3 Infrastructure

## Why does it matter?

The quality of infrastructure is essential for the efficient functioning of an economy. Modern and efficient infrastructure endowment contributes to both economic efficiency and territorial equity as it allows for the maximization of the local economic potential and the efficient exploitation of resources (Crescenzi and Rodriguez-Pose, 2008). As pointed out by Schwab and Porter (2007), it is an important factor determining the location of economic activity and the kinds of activities and sectors that can develop in an economy. High-quality infrastructure guarantees easy access to other regions and countries, contributes to better integration of peripheral and lagging regions, and facilitates the transport for goods, people and services. This has a strong impact on competitiveness as it increases the efficiency of
regional economies. The pillar describes different dimensions of infrastructural quality such as infrastructure density, connectivity and accessibility.

The list of candidate indicators, all available at the regional level, is shown in Box 11.

## Box 11: Indicators for Infrastructure

| Data source |  | Indicator description |
| :--- | :--- | :--- |
| 1. | Eurostat/DG <br> TREN/EuroGeographics/National | Motorway index |
| 2. | Railway index |  |
| 3. | Eutistical Institutes | Surat/EuroGeographics/National |
| Statistical Institutes |  |  | Number of flights accessible with 90' drive $\quad$.

### 3.4 Health

## Why does it matter?

This pillar is devoted to the description of human capital in terms of health condition and well-being, with special focus on the workforce. The 2006 Community Strategic Guidelines on Cohesion (Official Journal of the European Union, 2006) underline that a healthy workforce is a key factor in increasing labor market participation and productivity and enhancing competitiveness at national and regional level. They point out to major differences in health status and access to health care across European regions. Good health conditions of the population lead to greater participation in the labor force, longer working life, higher productivity and lower healthcare and social costs. Box 12 shows possible indicators to measure some of these aspects, available from Eurostat at the NUTS 2 regional level.

Box 12: Indicators for Health

| Data source |  | Indicator |
| :--- | :--- | :--- |
| 1. | Eurostat Regional Health Statistics | Hospital beds |
| 2. | Eurostat, CARE, ITF, National Statistical <br> Institutes, DG Regional Policy | Road fatalities |
| 3. | Eurostat, DG Regional Policy | Healthy life expectancy |
| 4. | Eurostat Regional Health Statistics | Infant mortality |
| 5. | DG Regional Policy | Cancer disease death rate |
| 6. | Eurostat, DG Regional Policy | Heart disease death rate |


| 7. | Eurostat, DG Regional Policy | Suicide death rate |
| :--- | :--- | :--- |

Among the candidate indicators, Hospital beds is the only one which gives an indication of an 'input' factor from the health system. The remaining indicators are related either to outcomes - infant mortality, cancer and heart disease death rates - or to the social welfare in more general terms - road fatalities and suicide rate. Our intent is, in fact, to measure some aspects of the population well-being from not only strictly health but also more social point of view.

### 3.5 Quality of Primary and Secondary Education

## Why does it matter?

High levels of basic skills and competences increase the ability of individuals to subsequently perform well in their work and to continue to tertiary education. To capture this dimension we focus on compulsory education outcomes as an indication of effectiveness and quality of the educational system across EU Member States. To this aim, we have taken into account the performance of students in the OECD Programme for International Student Assessment (PISA) 2006 wave. PISA indicators make it possible to identify the share of pupils, 15 year old, who have a low level of basic skills in reading, math and science. Pupils who fail to reach higher levels can be considered to be inadequately prepared for the challenges of the knowledge society and for lifelong learning, thus indicating a lower potential in terms of human capital.

In order to describe educational input factors, we also consider indicators related to teacher to pupil ration, public expenditure on compulsory education and financial aid available for students. Investment in education can be considered as an essential element in guaranteeing good quality of the educational system.

Participation in early childhood education has become one of the new EU benchmarks in the field of education and training. Several studies have pointed out to the positive effects of early childhood education from an educational and social perspective as it can counter potential educational disadvantages of children, coming from unfavorable family situations (NESSE, 2009; European Commission, 2009a). We have, thus, included a potential indicator measuring this aspect.

The following box presents the set of proposed indicator describing the Quality of Primary and Secondary education.

| Box 13: Indicators for Quality of Primary and Secondary Education |  |  |
| :---: | :---: | :---: |
| Data source |  | Indicator description |
| 1. | OECD - PISA | Low achievers in Reading of 15-year-olds |
| 2. |  | Low achievers in Math of 15-year-olds |
| 3. |  | Low achievers in Science of 15-year-olds |
| 4. | Eurostat Educational Statistics | Teacher/pupil ratio |
| 5. |  | Financial aid to students ISCED 1-4 |
| 6. |  | Public expenditure ISCED 1 |
| 7. |  | Public expenditure ISCED 2-4 |
| 8. |  | Participation in early childhood education |

### 3.6 Higher Education/Training and Lifelong Learning

## Why does it matter?

The contribution of education to productivity and economic growth has been widely researched in the last decades. Knowledge-driven economies based on innovation require well-educated human capital, capable to adapt, and education systems which successfully transmit key skills and competences. A clear picture of the economic benefits of education can be found in the most current release of the OECD publication Education at a Glance 2009 (OECD, 2009). As also underlined by the Lisbon Council president (Hofheinz, 2009), the main findings of the OECD report are straightforward: investment in educations pays always, for the individual and for society at large. Further, a stream of research literature in the past two decades has shown that the quality of human resources is not only directly involved in knowledge generation but plays a crucial role for applying and imitatatin technologies developed somewhere else (ex. Azariadis and Drazen, 1990).

It is clear that this pillar plays a key role in describing competitiveness.
Variables traditionally used for measuring educational quality are levels of educational attainment of the population, number of years of schooling of the labour force or literacy
rates (Psacharopoulous, 1984). Participation in education throughout one's life has also been deemed essential for the continuous upgrade of the skills and competences of workers in order to assist them in handling the challenges of continuously evolving technologies. In this pillar, these aspects are captured by proposing to include indicators on levels of tertiary educational attainment, participation in lifelong learning among the population as well as percentage of young people who have left the educational system at an earlier stage. Furthermore, an indicator of geographical accessibility to higher education institutions is proposed as a relevant factor, especially at the regional level. All these indicators are available at the required NUTS2 regional level. The analysis has been complemented by adding a fifth indicator to take into account the expenditure on tertiary education.

Box 14 presents the indicators proposed to describe the pillar.

| Box 14: Indicators for Higher Education/Training and Lifelong Learning |  |  |
| :--- | :--- | :--- |
| Data source | Indicator |  |
| 1. | Eurostat - LFS | Higher educational attainment (ISCED 5-6) |
| 2. | Eurostat Regional Education Statistics | Lifelong learning |
| 3. | Eurostat Structural Indicators | Early school leavers |
| 4. | Nordregio, EuroGeographics, <br> GISCO, EEA ETC-TE | Accessibility to universities |
| 5. | Eurostat Educational Statistics | Total public expenditure on tertiary <br> education (ISCED 5-6) |

### 3.7 Labor Market Efficiency

## Why does it matter?

The efficiency of the labor market gives an important indication as to the economic development or a region. Efficient and flexible labor markets contribute to efficient allocation of resources (Schwab and Porter, 2007).

We have used nine indicators to describe this pillar. Three of them are directly related to the level of employment/unemployment. Employment and unemployment rates indicate the level of activity of the regional economy while long-term unemployment can give indication as to the presence of structural problems in the economy. Furthermore, high employment rates do not necessarily correspond to high labor productivity which is one of the main
factors in a region's competitiveness. High labor productivity attracts economic activity and increases competitiveness. Thus, we have included data on regional labor productivity.

An interesting indicator on job mobility has been added to the suite of candidate indicators. It is officially defined by Eurostat as people who started to work for the current employer or as selfemployed in the last two years (as percentage of total employment). Our aim is to describe, following the most recent trends in employment policy, a labor market which promotes job creation and flexibility while maintaining quality of employment. Clearly job mobility includes temporary workers, but the intention is here to value temporary work as it may represent a way for the worker to acquire valuable experience while not having to commit himself to a single employer.

According to Schwab and Porter (2007), efficient labor markets ensure equity in the business environment between men and women. We have, thus, analyzed three indicators describing the equity aspect of the labor market - female unemployment, and differences in unemployment and employment rates between females and males in order to account for any gender bias in labor market participation.

Labor market policies (LMP) contribute to the more efficient match between labor market demand and supply. Data on LMP provides information on labor market interventions defined as "Public interventions in the labor market aimed at reaching its efficient functioning and correcting disequilibria and which can be distinguished from other general employment policy interventions in that they act selectively to favor particular groups in the labor market." The scope of LMP statistics is limited to public interventions which are explicitly targeted at groups of persons with difficulties in the labor market: the unemployed, persons employed but at risk of involuntary job loss and inactive persons who would like to enter the labor market. ${ }^{8}$

Box 15 reports the list of candidate indicators selected for the pillar.

| Box 15: Indicators for Labour Market Efficiency |  |  |
| :--- | :--- | :--- |
| Data source |  | Indicator |
| 1. | Eurostat Regional Labour Market Statistics | Employment rate |

[^6]| 2. | (LFS) | Long-term unemployment |
| :---: | :--- | :--- |
|  |  | Unemployment rate |
| 4. |  | Job mobility |
| 5. | Eurostat Economic Statistics | Labour productivity |
| 6. | Eurostat, DG Regional Policy | Difference between female and male <br> unemployment rates |
| 7. | Eurostat, DG Regional Policy | Difference between male and female <br> employment rates |
| 8. | Eurostat Regional Labour Market Statistics <br> (LFS) | Female unemployment |
| 9. | Eurostat Regional Labour Market Policy <br> Statistics (LFS) | Public expenditure on Labour Market <br> Policies |

### 3.8 Market Size

## Why does it matter?

The pillar Market Size aims at describing the size of the market available to firms which directly influences their competitiveness. In fact, larger markets allow firms to develop and benefit from economies of scale and could potentially give incentive to entrepreneurship and innovation. We capture not only the regional market, proxied by GDP, but also the potential market, which is not confined to the administrative borders of a region, by using an indicator on potential GDP within a pre-defined distance matrix (for more information, see Appendix E). Thus, we take into account the fact that the EU common market allows for easy access to neighboring regions, regardless of whether they are situated within the same or another country.

Candidate indicators describing this theme are listed in Box 16.

Box 16: Indicators for Market Size

| Data source |  | Indicator |
| :---: | :--- | :--- |
| 1. | Eurostat Regional Economic Accounts | GDP |
| 2. |  | Compensation of employees |
| 3. | Eurostat, DG Regional Policy | Disposable income |
| 4. |  | Potential market size in GDP |


| 5. | Potential market size in population |
| :--- | :--- | :--- |

### 3.9 Technological Readiness

## Why does it matter?

The pillar Technological Readiness aims at measuring the level at which households and enterprises are using and adopting existing technologies. It is largely recognized that technological infrastructures are a fundamental ingredient for country development. The last two decades have seen a steady increase of the importance of new information and communication technologies - ICT - both in business and every-day life. ICT has profoundly changed the organizational structure of firms, facilitating the adoption of new and more efficient technologies, improving productivity and speeding-up commercial processes. Hence, the use of ICT has become an essential element of competitiveness. ICT have also changed the way people do things in their private life. In fact, the way employees within firms are able to use efficiently new technologies is to a large degree dependent upon the ways in which technologies have penetrated their everyday life. We, thus, measure this aspect of technological readiness by concentrating also on the use of ICT by households as a proxy for the level of penetration of technologies in the population.

We propose to divide the pillar into two sub-pillars which describing access and use of technology by individuals/families, on the one hand, and enterprises, on the other. The subpillar related to personal use ('households') is described by three indicators collected at the NUTS2 level, whilst the sub-pillar related to technological readiness of enterprises ('enterprises') is described by some indicators at the NUTS2 level and by others at the country level. However, as it will be detailed later in Section 5.9, indicators available at the regional level are affected by a high percentage of missing values.

Box 17 and Box 18 show the candidate indicators for the two sub-pillars.

Box 17: Indicators for Technological Readiness - Sub-pillar HOUSEHOLDS

|  | Data source | Indicator |
| :--- | :--- | :--- |
| 1. |  | Households with access to broadband |
| 2. | Regional Information <br> Society Statistics | Individuals who ordered goods or services over the <br> Internet for private use |
| 3. |  | Households with access to Internet |


| Box 18: Indicators for Technological Readiness - Sub-pillar ENTERPRISES |  |  |
| :---: | :---: | :---: |
| Data source |  | Indicator |
| 1. | Community Survey on ICT usage and e-Commerce | Enterprises use of computers |
| 2. |  | Enterprises having access to Internet |
| 3. |  | Enterprises having a website or a homepage |
| 4. |  | Enterprises using Intranet |
| 5. |  | Enterprises using internal networks (e.g. <br> LAN) <br> Per |
| 6. |  | Persons employed by enterprises which use Extranet |
| 7. |  | Persons employed by enterprises which have access to the Internet |

### 3.10 Business Sophistication

## Why does it matter?

The level of business sophistication within an economy gives a sign as to the level of its productivity and its potential for responding to competitive pressures. Specialization in sectors with high value added contributes positively to the competitiveness of regions. We have, thus, included indicators on employment and GVA specifically in the NACE sectors J (information and communication) and K (Financial and insurance activities).

Furthermore, it is widely accepted that Foreign Direct Investments (FDI) are beneficial for the economic performance of countries and regions as they contribute to enhancing the capital and technological endowment of the host country or region (e.g. Barba Navaretti and Venables, 2004). We have included an indicator of FDI intensity, proxied by the number of new foreign firms, in order to capture this aspect of competitiveness.

Geographical proximity and interconnectedness among firms and suppliers leads to different types of spillovers, productivity and efficiency, but most importantly knowledge spillovers due to the higher concentration of specialized human capital. We, hence, propose to include a measure of the state of cluster development, similar to the practice used by the GCI, which describes the state of cluster development and gives an indication of the level of regional specialization and business sophistication (Schwab and Porter, 2007). As Porter also points out (Porter, 1998), regional clusters could lead to higher competitiveness for firms that are part of them due to the increasing productivity, higher innovation rate and availability of specialized resources. A variable on the strength of regional clusters is included which not only evaluates the level to which a region has been able to specialize in a given sector(s) but especially so in knowledge and technology-intensive sectors.

We have also considered indicators describing the availability of venture capital as it can give information as to the financial sophistication of the region and the potential of access to captal.

Proposed indicators to be included in the pillar are shown in Box 19 .
Box 19: Indicators for Business Sophistication

| Data source |  | Indicator |
| :--- | :--- | :--- |
| 1. | Eurostat Regional Labour Market <br> Statistics | Employment in 'sophisticated' sectors <br> (NACE sectors J-K) |
| 2. | Eurostat Regional Economic <br> Accounts | Gross Value Added (GVA) in 'sophisticated' <br> sectors (NACE sectors J-K) |
| 3. | ISLA-Bocconi | FDI intensity |
| 4. | European Cluster Observatory | Aggregate indicator for strength of regional <br> clusters (for details on the computation, see <br> Appendix B) |
| 5. | Eurostat, European Private <br> Equity and Venture Capital <br> Association (EVCA) | Venture capital (investments early stage) |
| 6. | Venture capital (expansion-replacement) |  |
|  | Venture capital (buy outs) |  |
| 7. |  |  |

### 3.11 Innovation

Why does it matter?

As pointed out by Schwab and Porter (2007), innovation is especially relevant for developed economies. They need to be at the forefront of new technologies, produce cutting-edge products and processes in order to maintain their competitive advantage. This requires an environment which is conducive, as Cantwell (2006) underlines, to creating relationships between firms and the science infrastructure, producers and users of innovation and the inter-firm level and between firms and the wider institutional environment. Furthermore, he stresses that such mechanisms are strongly influenced by spatial proximity. The level of innovative capability of a region influences directly the ways in which technology is diffused within the region. Research has shown that knowledge production is highly geographically concentrated. Feldman (1993) suggests that firms producing innovations tend to locate in areas with resources and that resources accumulate due to a region's success with innovations.

We have included both input or innovative potential indicators, such as employment in science and technology, knowledge workers, core creativity class, R\&D expenditure, and outcome indicators (patent applications). Our objective is to capture as much as possible both the regional potential to innovate as well its actual performance in innovative activities. Potential indicators are listed in Box 20.

Box 20: Indicators for Innovation

| Data source |  | Indicator |
| :---: | :---: | :---: |
| 1. | OECD REGPAT | Innovation patent applications |
| 2. |  | Total patent applications |
| 3. | Eurostat - LFS | Core Creative class employment |
| 4. |  | Knowledge workers |
| 5. | Thomson Reuters Web of Science \& CWTS database (Leiden University) | Scientific publications |
| 6. | Eurostat Regional Science and Technology Statistics | Total intramural R\&D expenditure |
| 7. |  | Human resources in Science and Technology (HRST) |
| 8. |  | Employment in technology and knowledge-intensive sectors |


| 9. | OECD - REGPAT | High-tech inventors |
| :---: | :---: | :---: |
| 10. |  | ICT inventors |
| 11. |  | Biotechnology inventors |

### 3.12 Stages of development of the EU NUTS2 regions

As mentioned in Section 2.1, the GCI by WEF takes into account the development stage of a country and accordingly assigns a different weighting scheme to groups of pillars (Schwab and Porter, 2007). Given that some variability across the development stages of NUTS 2 regions of the 27 EU members is expected, a similar approach is adopted for the RCI.

The first criterion proposed by WEF is considered ${ }^{9}$, that is the development stage of a region is defined according to its GDP level per capita at current market prices. We have taken GDP per capita measured as PPP per inhabitants and expressed as percentage of the EU average (\% GDP) as a defining variable. The year of reference is 2007. We have classified EU regions in three categories - low, medium and high according to the \%GDP.

Table 2: GDP thresholds for RCI computation

| Stage of development | GDP per capital (PPP per inhabitant as <br> $\%$ of EU average) |
| :--- | :--- |
| Medium | $<75 \%$ |
| Intermediate | $\geq 75 \%$ and $<100 \%$ |
| High | $\geq 100 \%$ |

The threshold defining the 'low' level (GDP below $75 \%$ of EU average) has been taken as a reference as it is the criterion for identifying regions eligible for funding under the Convergence criteria of the EU Regional Policy 2007-2013 framework. The second threshold ( $100 \%$ of EU average) has been a more arbitrary choice and has been examined by uncertainty analysis in Chapter 6.

[^7]Medium stage of development is associated with regional economies primarily driven by factors such as lower skilled labor and basic infrastructures. Aspects related to good governance and quality of public health are considered basic inputs in this framework.

Intermediate stage of development is characterized by labor market efficiency, quality of higher education and market size, factors which contribute to a more sophisticated regional economies and greater potential for competitiveness.

In the high stage of development, factors related to innovation, business sophistication and technological readiness are necessary inputs for innovation-driven regional economies.

On the basis of these thresholds, EU NUTS 2 regions are classified into different development stages. Appendix F shows the relative development stage assigned to each of the EU NUTS 2 regions.

## 4 Statistical assessment

The project aims at measuring the level of competitiveness by providing the RCI for European regions at the NUTS2 level.

The phenomenon is a multi-faceted concept that cannot be directly measured. The underlying hypothesis of this kind of analysis is that the phenomenon to be measured may be indirectly observed by several variables (indicators), which describe different features/aspects of the latent dimension. Choosing different aspects and indicators is equivalent to choosing the 'framework' of the index. This framework may be seen as the 'measurement instrument' of the latent phenomenon.

In the case of the RCI, the framework has been constructed on the basis of literature review, general reasoning, experts' opinion and practitioners' advice, as outlined in Chapter 3.

The RCI is structured according to the framework illustrated in Figure 3-1 and comprises eleven pillars which aim at measuring the economic strength of a region and its potentialities in the short and log run. For each pillar a number of candidate indicators have been selected and then screened by a set of univariate and multivariate statistical analyses to assess the validity of the theoretical framework and the internal consistency of each pillar. The statistical analysis is useful to support the framework and identify possible pitfalls which require further refinements. As a result, the final set of indicators may be, and in general is, a subset of the initial candidates.

The present chapter is devoted to the process of data analysis from the methodological point of view and the various statistical techniques employed at different steps of the analysis. Chapter 5 provides a discussion of results separately for each pillar.

The guidelines which drove this preliminary data analysis are major references of applied statistical analysis (Zani, 2000; Helsel and Hirsch, 2002; Knoke et al., 2002; Morrison, 2005) and the OECD (2008) Handbook on composite indicators.

### 4.1 Distortion due to commuting patterns

The geographical level is that of the NUTS2 as defined officially by the EU. However, in a few cases some regions have been combined to correct for the bias due to commuting patterns. Commuting patterns can indeed distort some of the data points for certain NUTS2 regions. In particular a very high share of jobs in Inner London (UKI1) is taken up by residents of Outer London (UKI2). The same is true for jobs in Brussels. Almost half the jobs in Bruxelles Capital (BE10) are taken by residents of Vlaams Brabant (BE24) or Brabant Wallon (BE31). This is why these NUTS-2 regions are integral parts of the metro-region as defined by the OECD-EU. In addition, from a competitiveness point of view, it is not operational to calculate a score for a half of a functional labour market area. The competitiveness of a region relies on the quality of the available skills. In the case of Brussels and Inner London, the skills of the residents of the surrounding NUTS-2 region(s) are also relevant. To solve this problem we chose to combine Inner and Outer London to obtain a 'new' London region (coded as UKI00 for the purpose of the report) and the three regions around Bruxelles (BE10, BE 24, and BE31) to obtain BE00 (coded for the purpose of the report). The value of all the indicators has been accordingly combined in these cases, taking into account population size.

For the RCI computation, 268 total regions are then considered (the official number of NUTS 2 regions being 271).

Appendix D shows the NUTS2 classification used for RCI and the population sizes used for computing the weighted combined value of the indicators for the merged regions.

### 4.2 Missing data

When analyzing real data, the problem of missing data is always present at various degrees. In the case of RCI the preliminary selection of possible indicators has been driven also by the availability of a sufficient number of observations at least for the most recent surveys. For the purpose of this study a limit rate of $10 \%-15 \%$ of missing data has been considered as threshold for including an indicator in the RCI computation. In this way we could limit the issue to few cases. For some indicators we still face the following two situations:
$\checkmark$ for some NUTS2 regions NUTS2 values are not observed while NUTS1 values are available;
$\checkmark$ for all NUTS2 regions data are available at the country level only.
In the former case we assign NUTS1 values to the corresponding NUTS2 regions, thus imposing no variation at the NUTS2 level. Whenever only the country level is available no imputation is performed and data are marked as missing as we consider that imputing country values to the NUTS2 level would not give any information as to the regional variation within a country and would give a distorted message in the construction of the RCI.

In the latter case, if within the pillar most indicators are available at the NUTS2 level and only a minority of them at the country level, we adopted an empirical imputation method as described in the following subsection.

### 4.2.1 Imputation method

Whenever one or more indicators, within one pillar, are observed at the country level only, an imputation method is adopted which imputes missing data by statistical estimates using available data, which has been recently employed for the methodological assessment of the Regional Innovation Scoreboard (Hollanders et al, 2009). The method is detailed in the following.

Let $Y$ be an indicator observed only at the country level $-Y^{\text {national }}-$ for which it is necessary to estimate values at the regional level $-Y_{j}^{\text {regional }}$, with $j$ being the region index. Select a subset of indicators $\left\{X_{1}, X_{2}, \ldots . X_{k}\right\}$, where both national $-X^{\text {national }}-$ and regional values $-X_{j}^{\text {regional }}-$ have been observed, which are in direct relation to $Y$, according to either the analyst's judgment or some quantitative analysis. For a certain country $C^{10}$, the procedure calculates, for region $j$ and indicator $X_{i}$, the ratio:

$$
r_{i}^{j}=\frac{X_{i}^{\text {national }}}{X_{i j}^{\text {regional }}}
$$

where $X_{i}^{\text {national }}$ is the value of indicator $X_{i}$ at the country level and $X_{i j}^{\text {regional }}$ is the value of $X_{i}$ for region $j$ in country $C$. The arithmetic mean over of $r_{i}^{j}$ over $i$, the subset of indicators $\left\{X_{1}\right.$, $\left.X_{2}, \ldots . X_{k}\right\}$, is then computed to obtain an average ratio for each region of country $C$ :

[^8]$\bar{r}^{j}=\frac{1}{k} \sum_{i=1}^{n} r_{i}^{j}$
Eventually the missing value of indicator $Y$ for region $j$ in country $C$ is imputed by assuming that the average ratio $\bar{r}^{j}$ between region $j$ and $C$ is valid for $Y$, that is the missing value for region $j$ of indicator $Y$ is imputed as:
$Y_{j}^{\text {regional }}=\frac{Y^{\text {national }}}{\bar{r}^{j}}$
Given that all national values are available for indicator $Y$, all missing values at the regional level can be imputed.

The procedure stems from the idea to 'spread' national values of the indicator $Y$ across the regions according to the average performance of that region with respect to its country. The average performance is computed as a mean ratio of country and regional values for all the indicators, observed at the regional level, which show a significant correlation with $Y$, and are thus, considered as 'reference indicators' for $Y$.

### 4.3 Univariate analysis

In the first part of the statistical analysis, we focus on evaluating the quality of the candidate indicators and the extent to which they are sufficient and appropriate to describe their respective pillar. To this aim, indicators are first analyzed separately by a univariate analysis to
$\checkmark$ check for the presence of missing values and evaluate the feasibility of including the indicator;
$\checkmark$ compute basic descriptive statistics - mean, standard deviation, coefficient of variation, percentiles, minimum and maximum values;
$\checkmark$ check for skewness, that implies the presence of outliers, and adopt appropriate transformations;
$\checkmark$ normalize indicators.

For each indicator in each pillar a summary table with descriptive statistics and maps with the $10 \%$ best performing (market in blue) and $10 \%$ worst performing (marked in $\operatorname{red}^{11}$ ) regions are shown. These maps are effective in visualizing the performance of a region along the various aspects, described by the single indicators, contributing to competitiveness.

Histograms are also provided for depicting large differences in shape or symmetry across indicators. They give helpful information as to the need of scale transformation for indicators which demonstrate highly skewed or asymmetric distributions. The choice of transforming the indicator is based on the value of the distribution skewness. Each indicator is, hence, checked for skewness, transformed if necessary and then normalized. Note that the term 'transformation' is a general term which includes linear and non linear transformations. In this context we understand 'transformation' as a non linear one to symmetrize indicators in order to reduce the influence of outliers, and normalization as a linear transformation to get comparability across indicators and homoscedasticity.

## Data transformation

In data analysis transformations are done in order to make data more symmetric, more linear, and more constant in variance. Transformation are monotonic, to preserve order relation, and have in general the effect of either expanding or contracting the distances to extreme observations on one side of the median, making distributions more symmetric around their central location. The classical measure to detect asymmetry in a distribution of an indicator is the skewness, which is defined as the adjusted third moment divided by the cube of the standard deviation (Helsel and Hirsch, 2002):

$$
\kappa=\frac{n}{(n-1)(n-2)} \sum_{i=1}^{n} \frac{\left(x_{i}-\bar{x}\right)^{3}}{s^{3}}
$$

where $n$ is the number observed values for the indicator, $\bar{x}$ is the arithmetic mean and $s$ is its standard deviation. A right-skewed distribution has positive $\kappa$ and outliers on the right hand side of the histogram; a left-skewed distribution has negative $\kappa$ and a tail on the left. According to the skewness value the analyst chooses to proceed or not with data

[^9]transformation. A comment is in due. Despite the fact that transformations are often employed in the setting-up of composite indicators, it is worth noting that every transformation alters the original data. It is in general advisable to ponder the choice of transformation and to employ it only if really considered not avoidable. As for RCI, given the variety of indicators and their initial distribution used in the construction of RCI pillars, we adopted an approach which addresses data diversity while at the same time limits the use of transformations as much as possible. According to this, we chose a relative high threshold for $\kappa$, that is $|\kappa|=1$, to limit the number of transformations.

Indicators are then transformed if $|\kappa|>1$. In these cases we used a transformation belonging to the Box-Cox family.

The Box-Cox transformations are a set of power transformations for skewed data, which include the logarithmic transformation as particular case. They depend on parameter $\lambda$ and take the following form (Zani, 2000):

$$
\begin{array}{ll}
\Phi_{\lambda}(x)=\frac{x^{\lambda}-1}{\lambda} & \text { if } \lambda \neq 0 \\
\Phi_{\lambda}(x)=\log (x) & \text { if } \lambda=0
\end{array}
$$

Box-Cox transformations are continuous, monotonously increasing, concave if $\lambda<1$ or convex if $\lambda>1$. Due to these properties, the Box-Cox transformations generate a contraction of higher values when $\lambda<1$ and a stretching of the higher values when $\lambda>1$. Figure 4-1 shows some Box-Cox transformations corresponding to different values of the parameter $\lambda$. The choice of the value of $\lambda$ depends on whether the distribution has a positive or negative asymmetry; hence it depends on the value of the skewness $\kappa$. In the RCI case we set:
$\lambda=2 \quad$ if $\kappa \leq-1$ (left or negative skewness)
$\lambda=-0.05 \quad$ if $\kappa \geq+1$ (right or positive skewness)

We then adopted $\lambda=2$ to correct for negative skewness and $\lambda=-0.05$ to correct for negative skewness. This choice is the result of a series of experiments carried out on the RCI

EU.
data-set. This is in line with literature recommendation of avoiding the tendency to search for the 'best' transformation tailor-made on each indicator. When dealing with several similar data-sets, it is in fact suggested to find one single transformation which fits reasonably well for all, rather than using slightly different ones for each (Helsel and Hirsch, 2002). Nevertheless, for two (out of 57) RCI indicators a slight adaptation of parameter $\lambda$ was necessary to decrease the skewness value below the selected threshold.

It is worth noting that, given the low value chosen to correct for negative skewness, ( $\lambda=-$ $0.05)$, the transformation to correct for right skewness is very close to the logarithmic one, which corresponds to $\lambda=0$ (see 4-2).


Figure 4-1: Box-Cox transformations for some values of $\lambda$ of particular interest (Zani 2000)

If a negative value of $\lambda$ is necessary, as is the case with highly negatively skewed distributions, the Box-Cox transformation is inappropriate if some observations are null. In
these cases a logarithmic transformation corrected for zero values is adopted (Longman et al., 1995):
$\Phi_{\lambda}(x)=\log (x+1)$
After transformation, the indicator distribution is checked again to verify that the skewness of the transformed indicator falls below the threshold. With this regard, for highly asymmetric distributions, which are generally associated to the massive presence of null values, a robust measure of skewness is adopted instead of , namely the quartile skew coefficient (Helsel and Hirsch, 2002):
$\kappa_{\text {quartile }}=\frac{\left(P_{0.75}-P_{0.50}\right)-\left(P_{0.50}-P_{0.25}\right)}{\left(P_{0.75}-P_{0.25}\right)}$
where $P_{0 . \mathrm{m}}$ is the $m$-th percentile. By definition $\mathcal{K}_{\text {quartile }}$ is based on the difference between distances of the upper and lower quartiles from the median divided by the interquartile range. As for $\kappa$, a right-skewed distribution has positive $\kappa_{\text {quartile }}$ and a left-skewed distribution has a negative $\kappa_{\text {quartile }}$.

In all cases where these transformations have been undertaken, the histograms include both the distribution of the original indicator and the one of the transformed indicator as well as the description of the type of transformation adopted.

## Normalization

Normalization is a kind of linear transformation. Normalization is necessary for any data aggregation as the indicators in a dataset have very frequently different measurement units and aggregation is meaningful only when indicators are comparable. There are a variety of normalization methods and the most frequently used in composite indicators are $z$-scores and min_max transformations (OECD, 2008).

For RCI weighted $z$-scores are adopted. As known, the $z$-scores transformation converts indicators to a common scale with a mean of zero and unitary standard deviation putting all indicator scores onto the same scale, one where the unit of measurement is the standard deviation (Knoke et al., 2002). In the RCI case, weighted averages and weighted standard deviation are chosen for the standardization with weights being the average population size
of the region in the period 2004-2008 (see Table in Appendix D), which is the period covered by the indicators in the RCI data-set. The value of each indicator is then transformed as

$$
\begin{align*}
& x_{\text {std }}=\frac{x-\bar{x}_{w}}{\sigma_{w}} \\
& \bar{x}_{w}=\frac{1}{P_{\text {tot }}} \sum_{i=1}^{n} x_{i} p_{i} \quad P_{\text {tot }}=\sum_{i=1}^{n} p_{i} \\
& \sigma_{w}=\sqrt{\frac{1}{P_{\text {tot }}} \sum_{i=1}^{n}\left(x_{i}-\bar{x}_{w}\right)^{2} p_{i}}
\end{align*}
$$

where $n$ is the total number of NUTS2 region $p_{i}$ is the average population size in region $i$ in the period 2004-2008.

For RCI computation, indicators are firstly transformed by a Box-Cox or logarithmic transformation, if necessary, and then they are all $z$-standardized.

### 4.4 Multivariate analysis

Multivariate analysis is carried out to verify internal data consistency within each pillar. Some general considerations are due at this point. In the setting-up of a composite each pillar is designed to describe a particular aspect of the latent phenomenon which is viewed as a 'combination' of related still different aspects. This implies that a desired feature of the composite framework is to have a high level of correlation within each pillar that would imply, in turn, that a unique single aspect is underlying each pillar. To assess, ex ante, that the selected indicators fulfill this requirement, a dimensionality reduction method is applied. To this aim Classical Principal Component Analysis (PCA) is employed separately for each pillar, as all the RCI indicators are numerical, quantitative variables. PCA is a classical multivariate exploratory technique that does not assume any statistical underlying model (Morrison, 2005).

Standard practice in PCA is to choose relevant dimensions if they (OECD, 2008):
$\checkmark$ are associated to eigenvalues above one (Kaiser's rule);
$\checkmark$ individually account to total variance by more than $10 \%$;
$\checkmark$ cumulatively contribute to total variance by more than $60 \%$.
For each pillar an overall PCA is carried out with all the indicators included in the pillar to assess/confirm the number of relevant dimensions 'behind' the pillar itself. Prior PCA, indicators are checked for the right orientation with respect to the level of competitiveness. As a rule, we chose to have a positive orientation, that is the higher the score the higher the competitiveness level. Accordingly, some indicators have been reversed.

The main goal of PCA for RCI is to statistically detect the number of underlying dimensions within each pillar. In the ideal situation, every sub-pillar should show a single most relevant dimension accounting for a large amount of variance, evenly described by all indicators included in the sub-pillar, with all concordant ${ }^{12}$ correlations with the main PCA component, that is the component loadings. This would also allow for completely avoiding compensability when aggregating indicators to get sub-scores at the pillar level, where 'compensability' is intended as the undesirable offsetting of low performing indicators with high performing ones. As it will be shortly discussed (Chapter 5), overall the framework chosen for RCI has been confirmed by the multivariate statistical analysis. Only few cases present anomalous indicators which may be due either to the choice of the indicators or their actual observed values. In these cases better alternatives have been looked for.

Various outcomes from PCA are reported and discussed for each of the ten pillars (Chapter 5) :
$\checkmark$ the correlation matrix between indicators;
$\checkmark$ the plot of eigenvalues with respect to their corresponding PCA dimension - scree plot, which visually indicates the presence of a major unique dimension, if any;
$\checkmark$ the component matrix, which shows the correlation coefficients between indicators and the PCA dimensions to identify indicators relevance in the composition of PCA components;
$\checkmark$ the total variance (both absolute and cumulative) explained by PCA dimensions, to determine their relevance in explaining the total indicators variance.

[^10]The PCA analysis helped to assess the validity of the underlying starting hypothesis of each pillar describing the same latent aspect of the level of competitiveness.

All the statistical analyses for RCI development are carried out using Matlab ${ }^{\circledR} 6.5$ and PASW Statistics ${ }^{\circledR} 18$.

The next Chapter presents the outcomes of the statistical assessment carried out pillar by pillar.

## 5 Pillar by pillar statistical analysis

Following the structure of the statistical assessment presented in Chapter 4, a separate discussion of each of the eleven pillars is outlined in the following sections. For each pillar, the chosen indicators are individually analyzed by univariate statistical methods and as a whole by the multivariate approach. The indicators used have a direct positive relation with competitiveness, i.e. the higher their value the higher the level of competitiveness. Whenever necessary, original indicators have been reversed. Multivariate analysis has been used to verify the existence of a single latent dimension. In few cases indicators which do not describe this common dimension, underlying the specific pillar, have been discarded (Appendix C gives information on all indicators considered and the reasons for discarding some of them). The geographical distribution of the pillar sub-score, computed as a simple average of the transformed/standardized indicators, is shown. Sub-scores are presented as min-max normalized scores (as percentage) and are divided into six classes, with high values associated with high competitiveness. Tables with corresponding sub-scores and the regions' ranks have been included at the end of each section.

### 5.1 Institutions

The candidate indicators identified to describe the pillar are detailed in Section 3.1. In the following we recall them, including the abbreviations used for the statistical analysis.

Indicators included, in brackets short names:

1. Corruption as a major national problem (reversed)
(country_corruption)
2. Presence of corruption in regional institutions (reversed) (regional_corruption)
3. Perceived level of budget defraud (reversed)
(budget_defraud)
4. Frequency of corruption and/or wrongdoing of national institutions (reversed)
(corruption_frequency)
5. Voice and accountability
(voice_accountability)
6. Political stability
7. Government effectiveness
(political_stability)
8. Regulatory quality
9. Rule of law
(govt_effectiveness)
10. Control of corruption
(regulatory_quality)
11. Ease of doing business (reversed)
(rule_of_law)
(corruption_control)
(business_ease)

## UNIVARIATE ANALYSIS

Table 3 presents some basic descriptive statistics of the eleven indicators listed above. All indicators are measured at the country level and we have no missing data for all but one indicator. Malta is not included in the ranking of the Ease of Doing Business index. Most indicators do not present high coefficients of variation with the exception of some of the World Bank Governance indicators - political stability, government effectiveness, rule of law, control of corruption and ease of doing business - which indicate a somewhat more heterogeneous situation among EU Member States.

Table 3：Descriptive statistics of Institutional indicators

|  |  |  |  |  | 악 뚱 n y |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | \％ | $\bigcirc$ |  |
|  |  |  | 号 | 앙 ${ }_{\text {d }}^{\sim}$ | กับ |
|  |  |  | 号 |  | $\underset{\sim}{\sim} \times{ }_{\sim}^{\sim}$ |
|  |  |  | 号 | ¢\％ | 示云苟 |
|  |  |  | \％ |  | $\underset{\sim}{\sim}$ N |
|  |  |  | 号 | O\％¢ | $\stackrel{\sim}{\sim}$ |
|  |  |  | $\stackrel{\circ}{\circ}$ | O앙 | $\stackrel{\circ}{\infty}$ |
|  |  |  | \％ | O- © |  |
|  |  |  | Oid | ○웅 굿 | ¢ |
|  |  |  | \％ | O- 웃 |  |
|  |  | \＃ |  |  |  |

## How do EU regions score in each of the indicators?

We can note that Scandinavian countries (Denmark, Sweden, Finland) are best performers in almost all indicators describing the Institutional pillar. Denmark is a top performer in five out the eleven indicators. We see Eastern European countries (Bulgaria, Romania, Estonia and Lithuania), and some of the Mediterranean countries (Greece, worst performer in four indicators, and Spain), having the lowest scores.



Figure 5-1: Best and worst performing regions for each indicator - Institutions

Out of all indicators, only two, national and regional corruption, have been transformed using the Box-Cox method. Histograms are shown in Table 4.

Table 4: Histograms of Institutional indicators


Budget defraud


Corruption frequency









## MULTivariate analysis

Despite the different sources of indicators which describe this pillar, the PCA analysis clearly depicts a single latent dimension almost uniformly represented by all the selected indicators. This can be easily seen in the scree plot (Figure 5-2) which reveals the presence of a clear unique aspect underlying the whole set of indicators included in the pillar. The correlation matrix (Table 5) accordingly shows that all the indicators are well correlated. The first PCA component alone explains more than $73 \%$ of total variation (Table 7). From Table 6 one can see that the contribution of each indicator to this component is approximately the same, with the exception of indicators budget_defraud, political_stability and business_ease which show a relatively lower correlation with the first dimension.

Overall, the multivariate analysis indicates the presence of a unique single latent dimension to which all the indicators contribute in a balanced way. This supports the simple choice of equal weights for the computation of the Institutional pillar sub-score as linear combination of transformed and standardized indicators. Figure 5-3 shows the geographical distribution of the Institutional sub-score at the country level, while Table 8 reports the Institutions pillar sub-score values. The distribution of sub-score values across countries is due in Figure 5-4.

Table 5: Correlation matrix between indicators included in the Institutions pillar

|  |  | country corruption reversed | regionalreversed | budget reversed | corruptionreversed | $\begin{aligned} & \text { voice } \\ & \text { accountability } \end{aligned}$ | $\begin{aligned} & \text { political } \\ & \text { stability } \end{aligned}$ | $\begin{aligned} & \text { effectiveñess } \\ & \text { ent } \end{aligned}$ | regulatory | rule_of_law | corruption_ control | $\begin{gathered} \hline \text { business_ } \\ \text { ease } \\ \text { reversed } \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Correlation | country_corruption_ reversed | 1.000 | . 958 | . 487 | . 757 | . 748 | 516 | . 769 | . 722 | . 709 | . 803 | 490 |
|  | regional_corruption_ reversed ${ }^{-}$ | . 958 | 1.000 | . 586 | . 834 | . 760 | . 584 | . 809 | . 738 | . 745 | 837 | . 510 |
|  | budget_defraud_reversed | . 487 | . 586 | 1.000 | . 800 | . 536 | . 356 | . 545 | . 598 | . 505 | . 599 | 486 |
|  | corruption_frequency_ reversed | . 757 | . 834 | . 800 | 1.000 | . 799 | . 510 | . 805 | . 750 | . 777 | . 861 | . 538 |
|  | voice_accountability | . 748 | . 760 | . 536 | . 799 | 1.000 | . 631 | . 950 | . 892 | . 962 | . 944 | . 507 |
|  | political_stability | . 516 | . 584 | . 356 | . 510 | .631 | 1.000 | 624 | . 494 | . 620 | . 594 | . 327 |
|  | govt_effectiveness | . 769 | . 809 | . 545 | . 805 | . 950 | . 624 | 1.000 | . 888 | . 964 | . 962 | . 606 |
|  | regulatory_quality | . 722 | . 738 | . 598 | . 750 | . 892 | 494 | . 888 | 1.000 | . 892 | . 878 | . 681 |
|  | rule_of_law | . 709 | . 745 | . 505 | . 777 | . 962 | . 620 | . 964 | . 892 | 1.000 | . 948 | . 552 |
|  | corruption_control | . 803 | . 837 | . 599 | . 861 | . 944 | . 594 | . 962 | . 878 | . 948 | 1.000 | . 601 |
|  | business_ease_reversed | . 490 | . 510 | . 486 | . 538 | . 507 | . 327 | . 606 | . 681 | . 552 | . 601 | 1.000 |
| Sig. (1-tailed) | country_corruption reversed |  | . 000 | . 005 | . 000 | . 000 | . 003 | . 000 | . 000 | . 000 | . 000 | . 006 |
|  | regional_corruption_ reversed ${ }^{-}$ | . 000 |  | . 001 | . 000 | . 000 | . 001 | . 000 | . 000 | . 000 | . 000 | . 004 |
|  | budget_defraud_reversed | . 005 | . 001 |  | . 000 | . 002 | . 034 | . 002 | . 000 | . 004 | . 000 | . 006 |
|  | corruption_frequency_ reversed | . 000 | . 000 | . 000 |  | . 000 | . 003 | . 000 | . 000 | . 000 | . 000 | . 002 |
|  | voice_accountability | . 000 | . 000 | . 002 | . 000 |  | . 000 | . 000 | . 000 | . 000 | . 000 | . 004 |
|  | political_stability | . 003 | . 001 | . 034 | . 003 | . 000 |  | . 000 | . 004 | . 000 | . 001 | . 052 |
|  | gov_effectiveness | . 000 | . 000 | . 002 | . 000 | . 000 | . 000 |  | . 000 | . 000 | . 000 | . 001 |
|  | regulatory_quality | . 000 | . 000 | . 000 | . 000 | . 000 | . 004 | . 000 |  | . 000 | . 000 | . 000 |
|  | rule_of_law | . 000 | . 000 | . 004 | . 000 | . 000 | . 000 | . 000 | . 000 |  | . 000 | . 002 |
|  | corruption_control | . 000 | . 000 | . 000 | . 000 | . 000 | . 001 | . 000 | . 000 | . 000 |  | . 001 |
|  | business_ease_reversed | . 006 | . 004 | . 006 | . 002 | . 004 | . 052 | . 001 | . 000 | . 002 | . 001 |  |

Scree Plot


Figure 5-2: PCA analysis of the Institutions pillar - eigenvalues

Table 6: PCA analysis Institutions pillar:
correlation coefficients between indicators and PCA components
Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 |
| country_corruption_ reversed | . 853 | -. 064 | . 238 | -. 370 | . 236 | . 077 | . 006 | . 095 | . 007 | . 025 | . 056 |
| regional_corruption_ reversed ${ }^{-}$ | . 894 | -. 018 | . 288 | -. 241 | . 207 | . 009 | . 033 | -. 101 | . 004 | . 001 | -. 074 |
| budget_defraud_reversed | . 680 | . 567 | . 303 | . 288 | -. 145 | . 104 | . 094 | . 027 | . 003 | . 017 | . 000 |
| corruption_frequency_ reversed | . 899 | . 228 | . 243 | . 022 | -. 114 | -. 196 | -. 168 | -. 027 | . 017 | -. 011 | . 022 |
| voice_accountability | . 940 | -. 192 | -. 100 | . 012 | -. 206 | . 025 | -. 043 | . 132 | . 046 | -. 011 | -. 065 |
| political_stability | . 657 | -. 440 | . 180 | . 506 | . 291 | . 030 | -. 023 | -. 001 | -. 010 | -. 006 | . 010 |
| govt_effectiveness | . 959 | -. 134 | -. 141 | -. 015 | -. 094 | -. 064 | . 116 | -. 049 | . 093 | -. 064 | . 037 |
| regulatory_quality | . 914 | . 060 | -. 252 | -. 038 | -. 092 | . 267 | -. 097 | -. 067 | -. 025 | -. 033 | . 016 |
| rule_of_law | . 934 | -. 191 | -. 184 | . 024 | -. 189 | -. 040 | . 025 | -. 046 | -. 006 | . 128 | . 011 |
| corruption_control | . 970 | -. 064 | -. 063 | -. 051 | -. 108 | -. 110 | . 068 | . 027 | -. 133 | -. 047 | . 002 |
| business_ease_reversed | .658 | . 402 | -. 472 | . 074 | . 413 | -. 076 | -. 004 | . 025 | . 008 | . 011 | -. 012 |

Extraction Method: Principal Component Analysis.
a. 11 components extracted.

Table 7: PCA analysis for the Institutions pillar: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative $\%$ |
| 1 | 8.115 | 73.770 | 73.770 |
| 2 | .831 | 7.557 | 81.327 |
| 3 | .676 | 6.144 | 87.471 |
| 4 | .544 | 4.950 | 92.421 |
| 5 | .495 | 4.502 | 96.923 |
| 6 | .151 | 1.376 | 98.299 |
| 7 | .068 | .622 | 98.921 |
| 8 | .048 | .441 | 99.362 |
| 9 | .030 | .270 | 99.632 |
| 10 | .025 | .230 | 99.861 |
| 11 | .015 | .139 | 100.000 |



Figure 5-3: Map of Institutions sub-score at the country level (min-max normalized values)

Table 8: Institutions sub-score as arithmetic mean of transformed and standardized indicators.

| country | Subscore | Min_max <br> normalized <br> subscore |
| :---: | :---: | :---: |
| BE | 0.54 | 57 |
| BG | -1.24 | 7 |
| CZ | -0.61 | 24 |
| DK | 2.05 | 100 |
| DE | 0.49 | 56 |
| EE | 0.41 | 53 |
| IE | 0.86 | 66 |
| GR | -1.47 | 0 |
| ES | -0.33 | 32 |
| FR | 0.3 | 50 |
| IT | -0.97 | 14 |
| CY | -0.24 | 35 |
| LV | -0.74 | 21 |
| LT | -0.75 | 20 |
| LU | 1.57 | 86 |
| HU | -0.75 | 20 |
| MT | 0.26 | 49 |
| NL | 1.56 | 86 |
| AT | 1 | 70 |
| PL | -0.84 | 18 |
| PT | -0.09 | 39 |
| RO | -1.37 | 3 |
| SI | -0.44 | 29 |
| SK | -0.44 | 29 |
| FI | 1.67 | 89 |
| SE | 1.46 | 83 |
| UK | 0.68 | 61 |
|  |  |  |

Figure 5-4. Histogram of Institutions sub-score


Table 9 shows the re-ordering of countries from best to worst in the quality of institutions.
Table 9: Institutions pillar sub-rank (from best to worst)

| Institutions |  |  |
| :---: | :---: | ---: |
| 1 | DK | Denmark |
| 2 | FI | Finland |
| 3 | LU | Luxembourg |
| 4 | NL | Netherlands |
| 5 | SE | Sweden |
| 6 | AT | Austria |
| 7 | IE | Ireland |
| 8 | UK | United Kingdom |
| 9 | BE | Belgium |
| 10 | DE | Germany |
| 11 | EE | Estonia |
| 12 | FR | France |
| 13 | MT | Malta |
| 14 | PT | Portugal |
| 15 | CY | Cyprus |
| 16 | ES | Spain |
| 17 | SI | Slovenia |
| 18 | SK | Slovakia |
| 19 | CZ | Czech republic |
| 20 | LV | Latvia |
| 21 | LT | Lithuania |
| 22 | HU | Hungary |
| 23 | PL | Poland |
| 24 | IT | Italy |
| 25 | BG | Bulgaria |
| 26 | RO | Romania |
| 27 | GR | Greece |

### 5.2 Macroeconomic stability

The indicators identified to describe the pillar are detailed in Section 3.2. In the following we recall them including the abbreviations used for the statistical analysis.

Indicators included, in brackets short names:

1. General government deficit (-) and surplus (+) (government_surplus/deficit)
2. Income, saving and net lending / net borrowing (national_savings)
3. Annual average inflation rate (reversed) (inflation)
4. Long term bond yields (reversed)
5. Government gross debt (reversed)
(government_bond_yields)
(government_debt)

Due to temporal fluctuations of all the indicators, we have computed the 2006-2008 average for each of them. The most recent data (2009) has not been included as, at the time of the RCI 2010 elaborations, the figures were not yet final but mostly provisional.

As for the orientation of the indicators, the first two - government surplus/deficit and national savings - are positively related to the level of competitiveness, while the remaining ones are all negatively related to competitiveness.

## Univariate analysis

Basic descriptive statistics of selected indicators are shown in Table 10. There are no missing values for three out of the five indicators. We have $7.41 \%$ of missing values for the indicator on national_savings which is below out threshold and thus, has been included. Similarly, government_bond_yields shows low percentage of missing values equal to $3.7 \%$. Greatest variation among EU Member states can be observed in the indicator on government_surplus/deficit.

Table 10: Descriptive statistics of Macroeconomic stability indicators

| Name of indicator | Government surplus/deficit | National savings | Inflation | Government bond yields | Government debt |
| :---: | :---: | :---: | :---: | :---: | :---: |
| description of indicator | \% of GDP | \% of GDP | Annual average rate of change in Harmonized Indices of Consumer Prices (HICPs) | EMU convergence criterion bond yields | \% of GDP |
| source | Eurostat | Eurostat | Eurostat | Eurostat | Eurostat |
| reference year | average 2006-2008 | average 2006-2008 | average 2006-2008 | average 2006-2008 | average 2006-2008 |
| $\%$ of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 0.00 \\ -0.95 \\ 2.62 \\ -2.76 \end{gathered}$ | $\begin{gathered} 7.41 \\ 20.17 \\ 5.59 \\ 0.28 \end{gathered}$ | $\begin{aligned} & 0.00 \\ & 3.90 \\ & 2.27 \\ & 0.58 \end{aligned}$ | $\begin{aligned} & 3.70 \\ & 4.63 \\ & 0.89 \\ & 0.19 \end{aligned}$ | $\begin{gathered} 0.00 \\ 44.98 \\ 26.81 \\ 0.60 \end{gathered}$ |
| maximum value region corresponding to maximum value minimum value region corresponding to minimum value | $\begin{gathered} 4.57 \\ \mathrm{FI} \\ -6.03 \\ \mathrm{HU} \end{gathered}$ | $\begin{gathered} 28.30 \\ \text { SE } \\ 7.87 \\ \text { GR } \end{gathered}$ | $\begin{gathered} 10.67 \\ \text { LV } \\ 1.83 \\ \mathrm{NL} \\ \hline \end{gathered}$ | $\begin{gathered} 7.37 \\ \text { HU } \\ 3.92 \\ \text { SE } \end{gathered}$ | $\begin{gathered} 105.27 \\ \text { IT } \\ 4.30 \\ \text { EE } \\ \hline \end{gathered}$ |

## How do EU regions score in each of the indicators?

The countries with highest government deficit are Hungary and Greece while the highest surplus is present in Finland and Denmark. Highest level of national savings is observed in Sweden and the Netherlands while lowest results are present in Greece and Cyprus. Highest inflation is present in Latvia and Bulgaria while the countries with lowest inflation rate are Sweden and the Netherlands. With regards to the indicator on government bond yields, highest trust by the markets is observed for Sweden and Germany while Romania and Hungary show the lowest results. Government debt is highest in Italy and Greece and lowest in Estonia and Luxembourg.



Figure 5-5: Best and worst performing regions for each indicator - Macroeconomic stability

Table 11 shows the histograms of the five indicators included in the Macroeconomic stability pillar. Two indicators have been transformed due to positive skewness - Inflation has been transformed with the Box-Cox method while Government_bond_yields has been transformed logarithmically.

Table 11. Histograms of Macroeconomic stability indicators






## MUltivariate analysis

The correlation and PCA analysis including all the indicators shows that the indicator Government_debt is not fully consistent with the others. The correlation matrix (

Table 12) already shows that Government_debt is significantly negatively correlated with the inflation indicator (reversed), with a correlation coefficient of -0.522 , while it is not correlated with National_savings and Government_bond_yields (reversed). Accordingly, the PCA scree plot (Figure 5-6) highlights the presence of two latent dimensions, the first accounting for $46 \%$ and the second for $32 \%$ of total variance (Table 13). The two dimensions have then comparable explanatory power, with the second one mostly related to Government_debt whose correlation coefficient with the second dimension is 0.94 (Table 14). This could be potentially explained by the fact that higher government debt is not necessarily related to a weak and unstable economy, especially in times of economic crisis. Moreover, there are particular countries, as Romania for instance, where the government debt is very low for political reasons (during the dictatorship the country was forced to be economically self-sufficient) but this is not positively correlated with higher competitiveness and economic stability. In fact, countries could have higher government debt, both in absolute terms and relative to GDP, but more competitive countries would have better prospects to pay it back, as partially described by the indicator Government_bond_yields. For these reasons the indicator Government_debt is more likely to have a 'bell shape'
behavior with respect to the level of competitiveness, rather than a linear one as can be captured by correlation and PCA-type analyses. This does not mean that the indicator is 'bad' in absolute terms, but that it does not fit into the simple mathematical structure desired, and needed, for the composite RCI.

Table 12: Correlation matrix between all initial indicators included in the Macroeconomic Stability pillar

Correlation Matrix

| Correlation Matrix |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Government_ surplus deficit | National savings | Inflation_ reversed | Government bond_yields_ reversed | Government_ debt reversed |
| Correlation | Government_surplus_deficit | 1.000 | . 496 | . 194 | . 544 | . 387 |
|  | National_savings | . 496 | 1.000 | . 368 | . 322 | . 187 |
|  | Inflation_reversed | . 194 | . 368 | 1.000 | . 610 | -. 522 |
|  | Government_bond_yields_ reversed | . 544 | . 322 | . 610 | 1.000 | -. 167 |
|  | Government_debt_reversed | . 387 | . 187 | -. 522 | -. 167 | 1.000 |
| Sig. <br> (1-tailed) | Government_surplus_deficit |  | . 006 | . 167 | . 002 | . 023 |
|  | National_savings | . 006 |  | . 035 | . 062 | . 186 |
|  | Inflation_reversed | . 167 | . 035 |  | . 000 | . 003 |
|  | Government_bond_yields_ reversed | . 002 | . 062 | . 000 |  | . 207 |
|  | Government_debt_reversed | . 023 | . 186 | . 003 | . 207 |  |



Figure 5-6: PCA analysis of all initial indicators included in the Macroeconomic Stability pillar - eigenvalues

Table 13: PCA analysis for the Macroeconomic Stability pillar, all initial indicators: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 2.279 | 45.586 | 45.586 |
| 2 | 1.624 | 32.475 | 78.061 |
| -3 | .643 | 12.861 | 90.922 |
| 4 | .242 | 4.850 | 95.772 |
| 5 | .211 | 4.228 | 100.000 |

Table 14: PCA analysis Macroeconomic Stability pillar, all initial indicators:
correlation coefficients between indicators and PCA components
Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 5 |  |
| Government_surplus_ <br> deficit | .712 | .556 | -.260 | -.335 | .062 |  |
| National_savings | .691 | .330 | .623 | .022 | -.158 |  |
| Inflation_reversed | .752 | -.551 | .129 | .066 | .332 |  |
| Government_bond_ <br> yields_reversed | .848 | -.135 | -.410 | .236 | -.196 |  |
| Government_debt_ <br> reversed | -.101 | .940 | -.048 | .263 | .185 |  |

Extraction Method: Principal Component Analysis.
a. 5 components extracted.

For the reasons discussed above, we decided to exclude the indicator Government_debt from further analysis. We believe that dropping this indicator will not penalize the pillar excessively. Indeed the indicator on government long-term bond yields, retained in the pillar, describes the market perception of the reliability of the country and its debt. In other words, no matter how large is a country's debt, the important thing is that investors believe that the country will be able to pay it back in the long-term.

In the following the multivariate analysis with the subset of indicators is discussed. The scree plot (Figure 5-7) shows that now only one prevalent dimension underlies the set of indicators, explaining almost $57 \%$ of total variability (Table 15). Each indicator contributes at roughly the same extent to this major dimension, as can be seen from the table of component loadings (Table 16).

The pillar without the Government_debt indicator is statistically consistent.


Figure 5-7: PCA analysis Macroeconomic Stability, without Government_debt

Table 15: PCA analysis Macroeconomic Stability pillar, without Government_debt: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 2.275 | 56.876 | 56.876 |
| 2 | .866 | 21.648 | 78.524 |
| 3 | .637 | 15.914 | 94.438 |
| 4 | .222 | 5.562 | 100.000 |

Table 16: PCA analysis Macroeconomic Stability pillar without Government_debt indicator:

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | :---: |
|  | 1 | 2 | 3 | 4 |  |
| Government_surplus_ <br> deficit | .743 | .520 | -.347 | .238 |  |
| National_savings | .708 | .389 | .571 | -.143 |  |
| Inflation_reversed | .719 | -.611 | .229 | .240 |  |
| Government_bond_ <br> yields_reversed | .840 | -.265 | -.371 | -.296 |  |

Extraction Method: Principal Component Analysis.
a. 4 components extracted.

The Macroeconomic stability sub-score is computed as a simple arithmetic mean of transformed (if necessary) and standardized values of the first four indicators listed at the beginning of this section. The geographical distribution of the sub-scores is shown in Figure 5-8 while Table 17 displays pillar sub-scores. The distribution of the sub-scores is shown in Figure 5-9.


Figure 5-8: Map of Macroeconomic Stability sub-score at the country level (min-max normalized values shown in Table 17)

Table 17: Macroeconomic Stability sub-score as arithmetic mean of transformed and standardized indicators.

| country | Subscore | Min_max <br> normalized <br> subscore |
| :--- | :--- | :---: |


| BE | 0.6 | 75 |
| :---: | :---: | :---: |
| BG | -0.83 | 35 |
| CZ | 0.08 | 60 |
| DK | 1.3 | 94 |
| DE | 0.83 | 81 |
| EE | -0.43 | 46 |
| IE | 0.18 | 63 |
| GR | -1.22 | 24 |
| ES | 0.23 | 65 |
| FR | 0.15 | 62 |
| IT | -0.05 | 57 |
| CY | -0.18 | 53 |
| LV | -1.35 | 20 |
| LT | -1 | 30 |
| LU | 0.67 | 77 |
| HU | -2.08 | 0 |
| MT | -0.23 | 52 |
| NL | 1.2 | 92 |
| AT | 0.75 | 79 |
| PL | -0.58 | 42 |
| PT | -0.55 | 43 |
| RO | -1.63 | 13 |
| SI | 0.25 | 65 |
| SK | -0.2 | 53 |
| FI | 1.47 | 99 |
| SE | 1.5 | 100 |
| UK | -0.55 | 43 |



Figure 5-9. Histogram of Macroeconomic Stability sub-score
Table 18 shows the re-ordering of countries from best to worst in terms of Macroeconomic Stability.

Table 18: Macroeconomic Stability pillar sub-rank (from best to worst)

| Macroeconomic stability |  |  |
| :---: | ---: | ---: |
| 1 | SE | Sweden |
| 2 | FI | Finland |
| 3 | DK | Denmark |
| 4 | NL | Netherlands |
| 5 | DE | Germany |
| 6 | AT | Austria |
| 7 | LU | Luxembourg |
| 8 | BE | Belgium |
| 9 | SI | Slovenia |
| 10 | ES | Spain |
| 11 | IE | Ireland |
| 12 | FR | France |
| 13 | CZ | Czech republic |
| 14 | IT | Italy |
| 15 | CY | Cyprus |
| 16 | SK | Slovakia |
| 17 | MT | Malta |
| 18 | EE | Estonia |
| 19 | PT | Portugal |
| 20 | UK | United Kingdom |
| 21 | PL | Poland |
| 22 | BG | Bulgaria |
| 23 | LT | Lithuania |
| 24 | GR | Greece |
| 25 | LV | Latvia |
| 26 | RO | Romania |
| 27 | HU | Hungary |

### 5.3 Infrastructure

Candidate indicators are described in Section 3.3 and are recalled bellow.
Indicators included, in brackets short names:

1. Motorway combined index (motorway_index_combined)
2. Railway combined index (railway_index_combined)
3. Number of passenger flights (number_of_passenger_flights)

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Table 19 presents the descriptive statistics for the three indicators included in the Infrastructure pillar. The motorway index refers to 2006 while the remaining two indicators refer to 2007. The coefficients of variation indicate diverse infrastructural condition within EU regions, especially so for the access to passenger flights. Two of the indicators do not have any missing data while the third one, number of passenger flights, presents only close to $2 \%$ of missing values.

Table 19: Descriptive statistics of Infrastructure indicators

| Indicator | Motorway density | Railway density | Number of passenger flights |
| :---: | :---: | :---: | :---: |
| description | motorway, combined index (average pop/area), EU27=100 | ```railway combined index (average pop/area), EU27=100``` | daily number of passenger flights (accessible within 90'drive) |
| source | Eurostat/DG <br> TREN/EuroGeographics/Na tional Statistical Institutes | Eurostat/DG <br> TREN/EuroGeographics/Na tional Statistical Institutes | Eurostat/EuroGeographics/Nat ional Statitical Institutes |
| reference year | 2006 | 2007 | 2007 |
| \% of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 0.00 \\ 146.65 \\ 127.01 \\ 0.87 \end{gathered}$ | $\begin{gathered} 0.00 \\ 138.40 \\ 91.56 \\ 0.66 \end{gathered}$ | $\begin{gathered} 1.87 \\ 587.42 \\ 672.53 \\ 1.14 \end{gathered}$ |
| maximum value region corresponding to maximum value minimum value region corresponding to minimum value | $\begin{gathered} 846.04 \\ \text { PT17 } \\ 0.00 \\ \text { BG32 } \end{gathered}$ | $\begin{gathered} 727.24 \\ \text { DE30 } \\ 0.00 \\ \text { GR21 } \end{gathered}$ | $\begin{gathered} 3428.67 \\ \text { UKJ1 } \\ 0.00 \\ \text { ES63 } \end{gathered}$ |

## How do EU regions score in each of the indicators?

Motorway development is underdeveloped in Eastern Europe while railway development sees Southern European regions underperforming. We can see that Mediterranean and Eastern European countries generally perform worse on the infrastructure indicators. Swedish regions score very high on the railway index. The UK region Berkshire, Buckinghamshire and Oxfordshire (UKJ1) has the highest number of daily passenger flights and generally, the southern regions of the UK have among the most developed passenger flight connections.


Figure 5-10: Best and worst performing regions for each indicator - Infrastructure ${ }^{13}$

[^11]Due to the nature of the infrastructure data and the presence of zero values, all indicators have been logarithmically transformed as described in Section 4.3. Table 20 shows the histograms of both the original and transformed values.

Table 20 Histograms of Infrastructure indicators




## Multivariate analysis

The PCA analysis highlights the presence of one prevalent dimension almost equally described by all the indicators. The analysis of both the scree plot (Figure 5-11) and the cumulative percentage of explained variance (Table 23) suggests the presence of a second minor dimension which accounts for about $23 \%$ of the total variance. This dimension is mainly represented by the indicator 'railway_index_combined' with which it has the highest correlation, 0.713 (Table 22). In any case, it can be concluded that this pillar has a unique, underlying dimension, well captured by the selected indicators.

The geographical distribution of sub-scores across NUTS2 regions is displayed in Figure 5-12 while the histogram of the Infrastructure sub-scores is shown in Figure 5-13. Negative skewness of the sub-score distribution can be noted which is due to the relevant presence of zero values in the original indicators (not eliminated by the indicator transformation). Reordered regions from best to worst are due in Table 25.

Table 21: Correlation matrix between indicators included in the Infrastructure pillar

## Correlation Matrix

|  |  | motorway_ <br> index___ <br> combined | railway_index <br> _combined | number_of_ <br> passenger_ <br> flights |
| :--- | :--- | ---: | ---: | ---: |
| Correlation | motorway_index_ <br> combined | .000 | .442 | .654 |
|  | railway_index_combined <br> number_of_passenger_ <br> flights | .642 | 1.000 | .334 |
| Sig. (1-tailed) | motorway_index_ <br> combined_ | .334 | 1.000 |  |
|  | railway_index_combined <br> number_of_passenger_ <br> flights | .000 | .000 | .000 |
|  |  | .000 | .000 |  |

Scree Plot


Figure 5-11: PCA analysis of the Infrastructure pillar - eigenvalues

Table 22: PCA analysis of the Infrastructure pillar: correlation coefficients between indicators and PCA components

## Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |
| :--- | ---: | ---: | :--- |
|  | 1 | 2 | 3 |
| motorway_index_ <br> combined | .885 | -.181 | -.430 |
| railway_index_combined <br> number_of_passenger_ <br> flights | .694 | .713 | .099 |

Extraction Method: Principal Component Analysis.
a. 3 components extracted.

Table 23: PCA analysis for the Infrastructure pillar:
explained variance
Total Variance Explained

| Component | Initial Eigenvalues |  |  | Extraction Sums of Squared Loadings |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative $\%$ | Total | \% of Variance | Cumulative \% |
| 1 | 1.966 | 65.549 | 65.549 | 1.966 | 65.549 | 65.549 |
| 2 | .701 | 23.361 | 88.910 | .701 | 23.361 | 88.910 |
| 3 | .333 | 11.090 | 100.000 | .333 | 11.090 | 100.000 |

Extraction Method: Principal Component Analysis.


Figure 5-12: Map of Infrastructure sub-score
(min-max normalized values)

Table 24: Infrastructure sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | 0.88 | 95 | Es30 | 0.52 | 88 | AT33 | 0.15 | 82 |
| BE21 | 0.73 | 92 | ES41 | 0.14 | 81 | AT34 | 0.08 | 80 |
| BE22 | 0.53 | 89 | ES42 | 0.37 | 86 | PL11 | -0.55 | 69 |
| BE23 | 0.65 | 91 | ES43 | -0.57 | 68 | PL12 | -1.05 | 60 |
| BE25 | 0.41 | 86 | ES51 | 0.18 | 82 | PL21 | -0.51 | 70 |
| ве32 | 0.67 | 91 | ES52 | -0.19 | 75 | PL22 | -0.14 | 76 |
| BE33 | 0.67 | 91 | ES53 | -1.96 | 43 | PL31 | -1.80 | 46 |
| BE34 | 0.54 | 89 | ES61 | -0.22 | 75 | PL32 | -1.72 | 47 |
| BE35 | 0.46 | 87 | ES62 | -0.28 | 74 | PL33 | -1.42 | 53 |
| BG31 | -1.17 | 57 | ES63 | -4.14 | 3 | PL34 | -2.19 | 39 |
| BG32 | -1.41 | 53 | ES64 | -4.14 | 3 | PL41 | -0.57 | 68 |
| BG33 | -0.63 | 67 | Es70 | -1.96 | 43 | PL42 | -0.57 | 68 |
| BG34 | -0.79 | 64 | FR10 | 0.72 | 92 | PL43 | -0.94 | 62 |
| BG41 | -0.56 | 69 | FR21 | 0.62 | 90 | PL51 | -0.42 | 71 |
| BG42 | -0.67 | 67 | FR22 | 0.73 | 92 | PL52 | -0.26 | 74 |
| czo1 | 0.67 | 91 | FR23 | 0.52 | 88 | PL61 | -1.40 | 53 |
| czo2 | 0.44 | 87 | FR24 | 0.48 | 88 | PL62 | -1.55 | 51 |
| czo3 | 0.23 | 83 | FR25 | -0.37 | 72 | PL63 | -1.20 | 57 |
| Cz04 | 0.36 | 85 | FR26 | 0.40 | 86 | PT11 | -0.54 | 69 |
| czos | -0.10 | 77 | FR30 | 0.54 | 89 | PT15 | -0.08 | 77 |
| czo6 | 0.22 | 83 | FR41 | 0.27 | 84 | PT16 | -0.16 | 76 |
| czo7 | -0.69 | 66 | FR42 | 0.46 | 87 | PT17 | 0.39 | 86 |
| czo8 | -1.10 | 59 | FR43 | 0.16 | 82 | PT18 | 0.21 | 83 |
| DK01 | 0.50 | 88 | FR51 | -0.14 | 76 | PT20 | -4.32 | 0 |
| DK02 | 0.51 | 88 | FR52 | -0.70 | 66 | PT30 | -3.27 | 19 |
| Dк03 | -0.01 | 79 | fr53 | -0.13 | 76 | RO11 | -1.64 | 49 |
| DK04 | -0.09 | 77 | FR61 | -0.11 | 77 | RO12 | -1.88 | 45 |
| DK05 | -0.31 | 73 | FR62 | -0.01 | 79 | RO21 | -1.96 | 43 |
| DE11 | 0.49 | 88 | FR63 | -0.06 | 78 | RO22 | -1.65 | 49 |
| DE12 | 0.76 | 93 | FR71 | 0.26 | 84 | RO31 | -0.51 | 70 |
| DE13 | 0.53 | 89 | FR72 | 0.04 | 80 | RO32 | 0.11 | 81 |
| DE14 | 0.31 | 84 | FR81 | 0.03 | 79 | RO41 | -2.28 | 37 |
| DE21 | 0.53 | 89 | FR82 | 0.02 | 79 | RO42 | -1.21 | 57 |
| DE22 | 0.45 | 87 | FR83 | -1.30 | 55 | S101 | 0.02 | 79 |
| DE23 | 0.53 | 89 | FR91 | -4.32 | 0 | S102 | 0.03 | 79 |
| DE24 | 0.31 | 84 | FR92 | -3.30 | 19 | Sk01 | 0.53 | 89 |
| DE25 | 0.61 | 90 | FR93 | -4.32 | 0 | SK02 | 0.09 | 80 |
| DE26 | 0.63 | 90 | FR94 | -4.32 | 0 | SK03 | -0.54 | 69 |
| DE27 | 0.56 | 89 | ITC1 | 0.43 | 87 | SK04 | -0.76 | 65 |
| de30 | 1.16 | 100 | ITC2 | 0.43 | 87 | F113 | -0.41 | 71 |
| DE41 | 0.59 | 90 | ITC3 | 0.38 | 86 | F118 | 0.00 | 79 |
| DE42 | 0.63 | 90 | ITC4 | 0.23 | 83 | F19 | -0.40 | 72 |
| DE50 | 1.14 | 100 | ItD1 | -0.06 | 78 | F11A | -0.48 | 70 |
| DE60 | 0.96 | 96 | ITD2 | -2.96 | 25 | F120 | -3.59 | 13 |
| DE71 | 0.84 | 94 | ItD3 | 0.16 | 82 | SE11 | 0.18 | 82 |
| DE72 | 0.67 | 91 | ItD4 | 0.06 | 80 | SE12 | 0.33 | 85 |
| DE73 | 0.59 | 90 | itD5 | 0.10 | 81 | SE21 | -0.05 | 78 |
| DE80 | 0.34 | 85 | ITE1 | 0.01 | 79 | SE22 | 0.35 | 85 |
| DE91 | 0.36 | 85 | ITE2 | -0.14 | 76 | SE23 | 0.04 | 80 |
| DE92 | 0.27 | 84 | ITE3 | -0.44 | 71 | SE31 | -0.07 | 78 |
| De93 | 0.45 | 87 | ITE4 | 0.26 | 84 | SE32 | -0.26 | 74 |
| DE94 | 0.27 | 84 | ITF1 | -0.01 | 79 | SE33 | -0.84 | 64 |
| deal | 0.93 | 96 | ITF2 | -0.08 | 77 | UKC1 | -0.09 | 77 |
| deaz | 0.82 | 94 | itf3 | -0.08 | 77 | UKC2 | -0.60 | 68 |
| deaz | 0.70 | 92 | ITF4 | -0.50 | 70 | UKD1 | 0.16 | 82 |
| dea4 | 0.44 | 87 | ITF5 | -0.53 | 69 | UKD2 | 0.70 | 92 |
| deas | 0.83 | 94 | ITF6 | -0.37 | 72 | UKD3 | 0.94 | 96 |
| Deb1 | 0.67 | 91 | ITG1 | -0.22 | 75 | UKD4 | 0.43 | 87 |
| DEB2 | 0.58 | 89 | ITG2 | -1.57 | 50 | UKD5 | 0.85 | 94 |
| deb3 | 0.78 | 93 | cyoo | -1.96 | 43 | UKE1 | 0.03 | 79 |
| deco | 0.79 | 93 | Lvoo | -1.10 | 59 | UKE2 | 0.05 | 80 |
| DED1 | 0.32 | 85 | Ltoo | -0.65 | 67 | UKE3 | 0.78 | 93 |
| DED2 | 0.26 | 84 | Lu00 | 0.38 | 86 | UKE4 | 0.67 | 91 |
| DED3 | 0.42 | 86 | HU10 | 0.07 | 80 | UKF1 | 0.32 | 85 |
| deeo | 0.46 | 87 | HU21 | 0.32 | 85 | UKF2 | 0.35 | 85 |
| defo | 0.30 | 84 | HU22 | 0.06 | 80 | UKF3 | -0.77 | 65 |
| DEGO | 0.26 | 84 | HU23 | -0.53 | 69 | UKG1 | 0.49 | 88 |
| EEOO | -0.71 | 66 | HU31 | -0.17 | 76 | UKG2 | 0.32 | 85 |
| 1E01 | -0.31 | 73 | HU32 | -0.31 | 73 | UKG3 | 0.94 | 96 |
| $1 E 02$ | -0.27 | 74 | низ3 | -0.19 | 75 | UKH1 | -0.06 | 78 |
| GR11 | -1.23 | 56 | мтоо | -3.23 | 20 | UKH2 | 0.69 | 91 |
| GR12 | -0.64 | 67 | NL11 | -0.11 | 77 | UкH3 | 0.47 | 87 |
| GR13 | -1.72 | 47 | NL12 | 0.27 | 84 | UKI | 1.15 | 100 |
| GR14 | -1.23 | 56 | NL13 | 0.03 | 79 | UKI1 | 0.67 | 91 |
| GR21 | -3.16 | 21 | NL21 | 0.48 | 88 | UK12 | 0.56 | 89 |
| GR22 | -3.44 | 16 | NL22 | 0.68 | 91 | UK13 | 0.48 | 88 |
| GR23 | -1.22 | 57 | NL23 | 0.13 | 81 | UKJ4 | 0.84 | 94 |
| GR24 | -0.24 | 74 | NL31 | 0.81 | 94 | Uкк1 | 0.40 | 86 |
| GR25 | -0.44 | 71 | NL32 | 0.69 | 91 | UKK2 | -0.10 | 77 |
| GR30 | -0.13 | 76 | NL33 | 0.76 | 93 | บккз | -1.18 | 57 |
| GR41 | -3.52 | 15 | NL34 | 0.35 | 85 | UKK4 | -0.49 | 70 |
| GR42 | -3.36 | 18 | NL41 | 0.61 | 90 | UKL1 | -0.29 | 74 |
| GR43 | -3.15 | 21 | NL42 | 0.71 | 92 | UKL2 | -0.05 | 78 |
| ES11 | -0.25 | 74 | AT11 | 0.26 | 84 | UKM2 | -0.05 | 78 |
| ES12 | -0.55 | 69 | AT12 | 0.60 | 90 | Uкмз | 0.17 | 82 |
| ES13 | -0.28 | 74 | AT13 | 1.04 | 98 | Uкм5 | -1.25 | 56 |
| ES21 | -0.10 | 77 | AT21 | 0.10 | 81 | UKM6 | -0.86 | 63 |
| ES22 | -0.22 | 75 | AT22 | 0.01 | 79 | UKNO | -0.35 | 72 |
| ES23 | -0.21 | 75 | AT31 | 0.21 | 83 |  |  |  |
| ES24 | -0.26 | 74 | AT32 | 0.30 | 84 |  |  |  |

Figure 5-13: Histogram of Infrastructure sub-score


Table 25. Institutions pillar sub-rank (from best to worst)

| Infrastructure |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | DE30 | 46 | DEB2 | 91 | DED1 | 136 | UKE1 | 181 | ES13 | 226 | LV00 |
| 2 | UKI | 47 | DE27 | 92 | HU21 | 137 | FR82 | 182 | ES62 | 227 | BG31 |
| 3 | DE50 | 48 | UKJ2 | 93 | UKF1 | 138 | SI01 | 183 | UKL1 | 228 | UKK3 |
| 4 | AT13 | 49 | BE34 | 94 | UKG2 | 139 | ITE1 | 184 | DK05 | 229 | PL63 |
| 5 | DE60 | 50 | FR30 | 95 | DE14 | 140 | AT22 | 185 | IE01 | 230 | RO42 |
| 6 | UKD3 | 51 | BE22 | 96 | DE24 | 141 | FI18 | 186 | HU32 | 231 | GR23 |
| 7 | UKG3 | 52 | DE13 | 97 | DEFO | 142 | DK03 | 187 | UKNO | 232 | GR11 |
| 8 | DEA1 | 53 | DE21 | 98 | AT32 | 143 | FR62 | 188 | FR25 | 233 | GR14 |
| 9 | BE00 | 54 | DE23 | 99 | DE92 | 144 | ITF1 | 189 | ITF6 | 234 | UKM5 |
| 10 | UKD5 | 55 | SK01 | 100 | DE94 | 145 | SE21 | 190 | FI19 | 235 | FR83 |
| 11 | DE71 | 56 | ES30 | 101 | FR41 | 146 | UKL2 | 191 | FI13 | 236 | PL61 |
| 12 | UKJ4 | 57 | FR23 | 102 | NL12 | 147 | UKM2 | 192 | PL51 | 237 | BG32 |
| 13 | DEA5 | 58 | DK02 | 103 | DED2 | 148 | FR63 | 193 | GR25 | 238 | PL33 |
| 14 | DEA2 | 59 | DK01 | 104 | DEGO | 149 | ITD1 | 194 | ITE3 | 239 | PL62 |
| 15 | NL31 | 60 | DE11 | 105 | FR71 | 150 | UKH1 | 195 | FI1A | 240 | ITG2 |
| 16 | DECO | 61 | UKG1 | 106 | ITE4 | 151 | SE31 | 196 | UKK4 | 241 | RO11 |
| 17 | DEB3 | 62 | FR24 | 107 | AT11 | 152 | ITF2 | 197 | ITF4 | 242 | RO22 |
| 18 | UKE3 | 63 | NL21 | 108 | CZO3 | 153 | ITF3 | 198 | PL21 | 243 | GR13 |
| 19 | DE12 | 64 | UKJ3 | 109 | ITC4 | 154 | PT15 | 199 | RO31 | 244 | PL32 |
| 20 | NL33 | 65 | UKH3 | 110 | CZO6 | 155 | DK04 | 200 | ITF5 | 245 | PL31 |
| 21 | BE21 | 66 | BE35 | 111 | AT31 | 156 | UKC1 | 201 | HU23 | 246 | RO12 |
| 22 | FR22 | 67 | DEEO | 112 | PT18 | 157 | CZO5 | 202 | PT11 | 247 | ES53 |
| 23 | FR10 | 68 | FR42 | 113 | ES51 | 158 | ES21 | 203 | SK03 | 248 | ES70 |
| 24 | NL42 | 69 | DE22 | 114 | SE11 | 159 | UKK2 | 204 | ES12 | 249 | CYOO |
| 25 | DEA3 | 70 | DE93 | 115 | UKM3 | 160 | FR61 | 205 | PL11 | 250 | RO21 |
| 26 | UKD2 | 71 | CZO2 | 116 | FR43 | 161 | NL11 | 206 | BG41 | 251 | PL34 |
| 27 | NL32 | 72 | DEA4 | 117 | ITD3 | 162 | GR30 | 207 | ES43 | 252 | RO41 |
| 28 | UKH2 | 73 | ITC1 | 118 | UKD1 | 163 | FR53 | 208 | PL41 | 253 | ITD2 |
| 29 | NL22 | 74 | ITC2 | 119 | AT33 | 164 | FR51 | 209 | PL42 | 254 | GR43 |
| 30 | BE32 | 75 | UKD4 | 120 | ES41 | 165 | ITE2 | 210 | UKC2 | 255 | GR21 |
| 31 | BE33 | 76 | DED3 | 121 | NL23 | 166 | PL22 | 211 | BG33 | 256 | MT00 |
| 32 | CZ01 | 77 | BE25 | 122 | RO32 | 167 | PT16 | 212 | GR12 | 257 | PT30 |
| 33 | DE72 | 78 | FR26 | 123 | ITD5 | 168 | HU31 | 213 | LT00 | 258 | FR92 |
| 34 | DEB1 | 79 | UKK1 | 124 | AT21 | 169 | ES52 | 214 | BG42 | 259 | GR42 |
| 35 | UKE4 | 80 | PT17 | 125 | SK02 | 170 | HU33 | 215 | CZO7 | 260 | GR22 |
| 36 | UKJ1 | 81 | ITC3 | 126 | AT34 | 171 | ES23 | 216 | FR52 | 261 | GR41 |
| 37 | BE23 | 82 | LU00 | 127 | HU10 | 172 | ES22 | 217 | EE00 | 262 | FI20 |
| 38 | DE26 | 83 | ES42 | 128 | ITD4 | 173 | ES61 | 218 | SK04 | 263 | ES63 |
| 39 | DE42 | 84 | CZ04 | 129 | HU22 | 174 | ITG1 | 219 | UKF3 | 264 | ES64 |
| 40 | FR21 | 85 | DE91 | 130 | UKE2 | 175 | GR24 | 220 | BG34 | 265 | FR91 |
| 41 | DE25 | 86 | NL34 | 131 | FR72 | 176 | ES11 | 221 | SE33 | 266 | FR93 |
| 42 | NL41 | 87 | SE22 | 132 | SE23 | 177 | ES24 | 222 | UKM6 | 267 | FR94 |
| 43 | AT12 | 88 | UKF2 | 133 | FR81 | 178 | PL52 | 223 | PL43 | 268 | PT20 |
| 44 | DE41 | 89 | DE80 | 134 | NL13 | 179 | SE32 | 224 | PL12 |  |  |
| 45 | DE73 | 90 | SE12 | 135 | SIO2 | 180 | IE02 | 225 | CZ08 |  |  |

### 5.4 Health

Candidate indicators are described in Section 3.4 and are here recalled with their abbreviations used in the analysis.

Indicators included, in brackets short names:

1. Hospital beds
(hospital_beds)
2. Road fatalities (reversed)
(road_fatalities)
3. Healthy life expectancy
(healthy_life)
4. Infant mortality (reversed)
5. Cancer disease death rate
6. Heart disease death rate
7. Suicide rate
(infant_mortality)
(cancer)
(heart_disease)
(suicide)

## Imputation of missing data

For the indicator on Hospital_beds, 2007 data has been used for most regions. However, for the following countries the most recent available data has been used: for Germany, Estonia, and Sweden - 2006 data (for Germany, NUTS 1 data has been imputed at the NUTS 2 level); for Greece - 2005 data, for Portugal - 2004 data; for the Netherlands - 2002 data.

For the indicator on Road_fatalities, 2004-2006 average has been used. However, in some cases, due to lack of data, different time periods have been considered: for Greece, Spain and France: 2003-2005; for Bulgaria, Ireland, Sweden and the UK: 2002-2004; for Italy: 2001-2003.

For the indicator on Infant_mortality, as 2007 data was not available for some countries, 2006 NUTS 2 data has been used for Belgium, Germany, Ireland, Italy, Poland, and the United Kingdom.

For the indicator on Healthy_life, data for DE 41 and DE 42 has been estimated by DG Regional Policy.

For the indicators on Cancer, Hearth_disease and Suicide, an average of 2006-2008 (or most recent year) has been taken.

## Univariate analysis

Table 26 presents the descriptive statistics for the seven indicators included in the Health pillar. All indicators have a very low percentage of missing values (less than $1 \%$ ) with the exception of the indicator on Hospital_beds $(11.19 \%)$ which is, however, still within the thresholds defined in Section 4.2 and has been included in the final computation of the subscore.

Table 26: Descriptive statistics of Health indicators

| Indicator | Hospital beds | Road fatalities | Healthy life expectancy | Infant mortality | Cancer disease death rate | Heart disease | Suicide |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| description | rate of hospital beds <br> per 100,000 <br> inhabitants | number of deaths in road accidents per million inhabitants | number of years of healthy life expected | number of deaths of children under 1 year of age during the year to the number of live births in that year | standardized cancer death rate for population under 65 | standardized heart diseases death rate for population under 65 | standardized death rate for suicide for population under 65 |
| source | Eurostat Regional Statistics | Eurostat, CARE, ITF, NSIs, DG Regio | Eurostat/DG Regional Policy | Eurostat Regional Statistics | DG Regio | DG Regio, Eurostat | DG Regio, Eurostat |
| reference year | 2007 | 2004-2006 | 2007 | 2007 | 2006-08 | 2006-08 | 2006-08 |
| $\%$ of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 11.19 \\ 592.26 \\ 204.97 \\ 0.35 \end{gathered}$ | $\begin{gathered} 0.75 \\ 103.35 \\ 46.47 \\ 0.45 \end{gathered}$ | $\begin{gathered} 0.37 \\ 62.24 \\ 3.41 \\ 0.05 \end{gathered}$ | $\begin{aligned} & 0.75 \\ & 4.03 \\ & 2.43 \\ & 0.60 \end{aligned}$ | $\begin{gathered} 0.37 \\ 76.38 \\ 15.93 \\ 0.21 \end{gathered}$ | $\begin{gathered} 0.37 \\ 51.57 \\ 31.70 \\ 0.61 \end{gathered}$ | $\begin{aligned} & 0.37 \\ & 9.95 \\ & 4.89 \\ & 0.49 \end{aligned}$ |
| maximum value <br> region corresponding to maximum value <br> minimum value <br> region corresponding to minimum value | $\begin{gathered} 1216.80 \\ \text { DE80 } \\ 165.60 \\ \text { NL23 } \end{gathered}$ | $\begin{gathered} 304.00 \\ \text { GR24 } \\ 17.50 \\ \text { DE50 } \end{gathered}$ | $\begin{gathered} 69.90 \\ \text { MTOO } \\ 52.23 \\ \text { EEOO } \end{gathered}$ | $\begin{array}{r} 14.20 \\ \text { RO21 } \\ 0.00 \\ \text { ITC2 } \\ \hline \end{array}$ | $\begin{aligned} & 143.90 \\ & \text { HU32 } \\ & 38.80 \\ & \text { NL23 } \end{aligned}$ | $\begin{aligned} & 189.80 \\ & \text { BG31 } \\ & 17.70 \\ & \text { NL23 } \end{aligned}$ | $\begin{array}{r} 28.20 \\ \text { LT00 } \\ 2.10 \\ \text { GR12 } \end{array}$ |

## How do EU regions score in each of the indicators?

Southern European and Scandinavian regions have very low numbers of hospital beds. Road fatalities present biggest problem in Southern European regions (Spanish, Greek and Portuguese) as well as in the Baltic countries. UK regions are among the ones with the lowest number of road fatalities. Most of the Scandinavian and Greek regions have very high healthy life expectancy while regions in the Baltic States, Finland, Hungary and Slovakia are among the ones with the lowest performance. Infant mortality is highest in Eastern European regions, Bulgaria and Romania specifically, while best performers are regions in Italy, Greece, Germany and United Kingdom. Cancer rate is highest in a number of Eastern European regions (Romanian, Hungarian, Bulgarian, Baltic) while best performers are parts of Italy, Sweden and Finland. Similarly, heart diseases are most common in Eastern Europe while most rare in Spanish, Portuguese and Southern French regions. Suicide rates are very low in Southern European regions and very high in Northern European regions.



Figure 5-14: Best and worst performing regions for each indicator - Health
As shown in Table 27, two of the indicators (Cancer and Heart_disease) have been transformed with the Box-Cox method while Infant mortality has been transformed logarithmically due to the presence of zero values.

Table 27: Histograms of Health indicators

## Hospital beds



Road fatalities


Healthy life




## Multivariate analysis

A rather low correlation characterizes the indicators included in the pillar (Table 28). This is due to the intrinsic nature of the indicators which describe very different aspects related to the heath conditions of the population. Among the candidate indicators, Hospital_beds shows the most anomalous behaviour, being negatively correlated with almost all the other indicators. The PCA analysis is not expected to show a unique underlying dimension and indeed this may be seen from the scree plot in Figure 5-15. At least two dimensions are
needed to reach about $60 \%$ of total variance (Table 30), with the second dimension mainly related to Hospital_beds and Road_fatalities (Table 29). These results suggest dropping the indicator Hospital_bed, which is also the only one which somehow describes an 'input' factor within the pillar.

Table 28: Correlation matrix between all candidate indicators of the Health pillar

| Correlation Matrix |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Hospital_beds | Road_fatalities_ reversed | Healthy_life | Infant_mortalityr eversed | Cancer_ reversed | Heart_disease reversed | Suicide_ reversed |
| Correlation | Hospital_beds | 1.000 | . 213 | -. 581 | -. 031 | -. 278 | -. 213 | -. 395 |
|  | Road_fatalities_reversed | . 213 | 1.000 | . 073 | . 079 | . 164 | . 252 | . 169 |
|  | Healthy_life | -. 581 | . 073 | 1.000 | . 194 | . 385 | . 419 | . 403 |
|  | Infant_mortality_reversed | -. 031 | . 079 | . 194 | 1.000 | . 391 | . 491 | . 145 |
|  | Cancer_reversed | -. 278 | . 164 | . 385 | . 391 | 1.000 | . 650 | . 474 |
|  | Heart_disease_reversed | -. 213 | . 252 | . 419 | . 491 | . 650 | 1.000 | . 273 |
|  | Suicide_reversed | -. 395 | . 169 | . 403 | . 145 | . 474 | . 273 | 1.000 |
| Sig. (1-tailed) | Hospital_beds |  | . 000 | . 000 | . 317 | . 000 | . 000 | . 000 |
|  | Road_fatalities_reversed | . 000 |  | . 117 | . 102 | . 004 | . 000 | . 003 |
|  | Healthy_life | . 000 | . 117 |  | . 001 | . 000 | . 000 | . 000 |
|  | Infant_mortality_reversed | . 317 | . 102 | . 001 |  | . 000 | . 000 | . 009 |
|  | Cancer_reversed | . 000 | . 004 | . 000 | . 000 |  | . 000 | . 000 |
|  | Heart_disease_reversed | . 000 | . 000 | . 000 | . 000 | . 000 |  | . 000 |
|  | Suicide_reversed | . 000 | . 003 | . 000 | . 009 | . 000 | . 000 |  |



Figure 5-15: PCA analysis of the Health pillar, all candidate indicators - eigenvalues

Table 29: PCA analysis of the Health pillar, all candidate indicators: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| Hospital_beds | -.561 | .707 | .015 | .134 | -.024 | .405 | .052 |
| Road_fatalities_reversed | .223 | .620 | .672 | -.250 | .116 | -.170 | -.093 |
| Healthy_life | .726 | -.360 | .097 | -.434 | .128 | .354 | -.062 |
| Infant_mortality_reversed | .523 | .436 | -.553 | .036 | .468 | -.067 | -.075 |
| Cancer_reversed | .806 | .184 | -.062 | .267 | -.373 | .045 | -.317 |
| Heart_disease_reversed | .777 | .346 | -.181 | -.173 | -.292 | -.053 | .353 |
| Suicide_reversed | .659 | -.190 | .406 | .526 | .229 | .073 | .174 |

Extraction Method: Principal Component Analysis.
a. 7 components extracted.

Table 30: PCA analysis for the Health pillar, all candidate indicators:
explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 2.853 | 40.755 | 40.755 |
| 2 | 1.393 | 19.904 | 60.659 |
| 3 | .970 | 13.852 | 74.511 |
| -4 | .648 | 9.253 | 83.764 |
| 5 | .527 | 7.528 | 91.292 |
| 6 | .333 | 4.755 | 96.047 |
| 7 | .277 | 3.953 | 100.000 |

The multivariate analysis without Hospital_beds is shown in Figure 5-16, Table 31 and Table 32. Results are better even if the first PCA dimension explains only $44 \%$ of total variation, slightly more than in the previous case. However, in this case all the indicators are positively related to the first major PCA dimension (Table 31) and roughly to the same extent (with the exception of Road_fatalities which has a low correlation coefficient, 0.33 ).


Figure 5-16: PCA analysis of the Health pillar, without Hospital_beds - eigenvalues

Table 31: PCA analysis of the Health pillar without Hospital_beds: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: | :---: |
|  | 1 |  | 2 |  | 3 | 4 |  |
| 5 |  |  |  |  |  |  |  |
| Road_fatalities_reversed | .331 | .239 | .898 | .106 | .102 | .072 |  |
| Healthy_life | .644 | .322 | -.337 | .574 | .168 | .102 |  |
| Infant_mortality_reversed | .593 | -.650 | -.003 | -.110 | .458 | .066 |  |
| Cancer_reversed | .833 | -.035 | -.071 | -.233 | -.378 | .322 |  |
| Heart_disease_reversed | .826 | -.270 | .091 | .155 | -.290 | -.357 |  |
| Suicide_reversed | .623 | .564 | -.153 | -.440 | .226 | -.163 |  |

Extraction Method: Principal Component Analysis.
a. 6 components extracted.

Table 32: PCA analysis for the Health pillar, without Hospital_beds: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 2.640 | 44.003 | 44.003 |
| 2 | .974 | 16.241 | 60.244 |
| 3 | .956 | 15.938 | 76.182 |
| -4 | .625 | 10.416 | 86.598 |
| 5 | .526 | 8.773 | 95.371 |
| 6 | .278 | 4.629 | 100.000 |

The final Health_sub-score has been computed as a simple arithmetic mean of the transformed and standardized indicators, excluding Hospital_beds. The geographical distribution of the sub-score across NUTS2 regions is displayed in Figure 5-17 based on values displayed in Table 33. The histogram of the Health sub-score is shown in Figure 5-18, while the ranking of regions are in Table 34.


Figure 5-17: Map of Health sub-score (min-max normalized values)

Table 33: Health sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | -0.10 | 60 | Es30 | 1.00 | 90 | AT33 | 0.30 | 71 |
| BE21 | 0.40 | 73 | ES41 | 0.13 | 66 | AT34 | 0.33 | 71 |
| BE22 | -0.30 | 54 | ES42 | 0.28 | 70 | PL11 | -1.00 | 35 |
| BE23 | -0.32 | 54 | ES43 | 0.23 | 69 | PL12 | -0.72 | 43 |
| BE25 | -0.18 | 58 | ES51 | 0.57 | 78 | PL21 | -0.38 | 52 |
| BE32 | -0.95 | 37 | ES52 | 0.27 | 70 | PL22 | -0.70 | 43 |
| BE33 | -0.57 | 47 | ES53 | 0.32 | 71 | PL31 | -0.80 | 41 |
| BE34 | -1.58 | 20 | ES61 | 0.13 | 66 | PL32 | -0.38 | 52 |
| BE35 | -0.82 | 40 | ES62 | 0.27 | 70 | PL33 | -0.73 | 43 |
| BG31 | -1.15 | 31 | ES63 | 0.33 | 71 | PL34 | -1.00 | 35 |
| BG32 | -1.02 | 35 | ES64 | 0.77 | 83 | PL41 | -1.00 | 35 |
| BG33 | -1.18 | 30 | Es70 | 0.20 | 68 | PL42 | -0.90 | 38 |
| BG34 | -1.45 | 23 | FR10 | 0.62 | 79 | PL43 | -0.88 | 39 |
| BG41 | -0.80 | 41 | FR21 | -0.27 | 55 | PL51 | -0.88 | 39 |
| BG42 | -1.08 | 33 | FR22 | -0.53 | 48 | PL52 | -0.73 | 43 |
| cZ01 | 0.22 | 68 | FR23 | -0.28 | 55 | PL61 | -0.77 | 42 |
| czo2 | -0.62 | 46 | FR24 | -0.13 | 59 | PL62 | -0.80 | 41 |
| cz03 | -0.42 | 51 | FR25 | -0.22 | 57 | PL63 | -0.60 | 46 |
| cz04 | -1.02 | 35 | FR26 | -0.23 | 56 | PT11 | 0.28 | 70 |
| cz05 | -0.32 | 54 | FR30 | -0.53 | 48 | PT15 | -0.62 | 46 |
| cZ06 | -0.38 | 52 | FR41 | -0.27 | 55 | PT16 | 0.15 | 67 |
| CZ07 | -0.45 | 50 | FR42 | 0.13 | 66 | PT17 | 0.08 | 65 |
| cz08 | -0.50 | 49 | FR43 | -0.08 | 60 | PT18 | -0.60 | 46 |
| DK01 | 0.57 | 78 | FR51 | -0.05 | 61 | PT20 | -0.84 | 40 |
| Dк02 | 0.23 | 69 | FR52 | -0.33 | 54 | PT30 | -1.14 | 32 |
| Dк03 | 0.28 | 70 | FR53 | -0.17 | 58 | RO11 | -1.38 | 25 |
| DK04 | 0.17 | 67 | FR61 | 0.00 | 63 | RO12 | -1.42 | 24 |
| Dк05 | 0.08 | 65 | FR62 | 0.28 | 70 | RO21 | -1.22 | 29 |
| DE11 | 0.55 | 77 | FR63 | -0.23 | 56 | RO22 | -1.28 | 28 |
| DE12 | 0.28 | 70 | FR71 | 0.40 | 73 | RO31 | -1.23 | 29 |
| DE13 | 0.08 | 65 | FR72 | -0.28 | 55 | RO32 | -0.72 | 43 |
| DE14 | 0.35 | 72 | FR81 | -0.17 | 58 | R041 | -1.07 | 33 |
| DE21 | 0.38 | 73 | FR82 | 0.12 | 66 | RO42 | -1.45 | 23 |
| DE22 | -0.23 | 56 | FR83 | -0.15 | 58 | S101 | -0.62 | 46 |
| DE23 | -0.22 | 57 | FR91 | -0.20 | 57 | 5102 | -0.33 | 54 |
| DE24 | 0.07 | 64 | FR92 | 0.17 | 67 | SK01 | -0.55 | 48 |
| DE25 | 0.10 | 65 | FR93 | -0.70 | 43 | Sk02 | -0.95 | 37 |
| DE26 | 0.38 | 73 | FR94 | -0.42 | 51 | Sk03 | -1.25 | 29 |
| DE27 | 0.03 | 63 | ITC1 | 0.32 | 71 | SK04 | -1.23 | 29 |
| de30 | 0.43 | 74 | ITC2 | 0.53 | 77 | F113 | -0.38 | 52 |
| DE41 | -0.40 | 52 | ITC3 | 0.62 | 79 | F118 | 0.08 | 65 |
| DE42 | -0.28 | 55 | ITC4 | 0.42 | 74 | F19 | -0.10 | 60 |
| DE50 | -0.25 | 56 | ItD1 | 0.40 | 73 | F11A | -0.18 | 58 |
| DE60 | 0.07 | 64 | ITD2 | 0.15 | 67 | F120 | 1.32 | 98 |
| DE71 | 0.37 | 73 | ItD3 | 0.37 | 73 | SE11 | 1.15 | 94 |
| DE72 | -0.13 | 59 | ItD4 | -0.07 | 61 | SE12 | 0.93 | 88 |
| DE73 | 0.33 | 71 | ITD5 | 0.12 | 66 | SE21 | 0.72 | 82 |
| DE80 | -0.07 | 61 | ITE1 | 0.67 | 81 | SE22 | 0.73 | 82 |
| DE91 | 0.02 | 63 | ITE2 | 0.67 | 81 | SE23 | 0.98 | 89 |
| DE92 | 0.12 | 66 | ITE3 | 0.62 | 79 | SE31 | 0.70 | 82 |
| DE93 | -0.27 | 55 | ITE4 | 0.55 | 77 | SE32 | 0.65 | 80 |
| DE94 | -0.17 | 58 | ITF1 | 0.67 | 81 | SE33 | 0.70 | 82 |
| deal | 0.07 | 64 | ITF2 | 0.23 | 69 | UKC1 | 0.15 | 67 |
| deA2 | 0.15 | 67 | ITF3 | 0.53 | 77 | UKC2 | 0.50 | 76 |
| dea3 | 0.15 | 67 | ITF4 | 0.82 | 85 | UKD1 | 0.38 | 73 |
| dea4 | 0.17 | 67 | ITF5 | 0.42 | 74 | UKD2 | 0.52 | 77 |
| deas | 0.15 | 67 | ITF6 | 0.87 | 86 | UKD3 | 0.07 | 64 |
| deb1 | -0.18 | 58 | ITG1 | 0.73 | 82 | UKD4 | 0.32 | 71 |
| DEB2 | -0.23 | 56 | ITG2 | 0.43 | 74 | UKD5 | 0.30 | 71 |
| deb3 | 0.25 | 69 | croo | 0.65 | 80 | UKE1 | 0.12 | 66 |
| deco | -0.18 | 58 | Lvoo | -2.20 | 3 | UKE2 | 1.07 | 92 |
| dedi | 0.10 | 65 | LToO | -2.30 | 0 | UKE3 | 0.43 | 74 |
| DED2 | 0.22 | 68 | Luoo | 0.32 | 71 | UKE4 | 0.12 | 66 |
| ded3 | 0.03 | 63 | HU10 | -1.35 | 26 | UKF1 | 0.40 | 73 |
| deeo | -0.47 | 50 | HU21 | -1.93 | 10 | UKF2 | 0.67 | 81 |
| DEFO | 0.03 | 63 | HU22 | -1.55 | 20 | UKF3 | 0.30 | 71 |
| DEGO | 0.12 | 66 | HU23 | -1.95 | 10 | UKG1 | 0.20 | 68 |
| eEoo | -1.47 | 23 | HU31 | -2.08 | 6 | UKG2 | 0.40 | 73 |
| IE01 | 0.47 | 75 | HU32 | -2.18 | 3 | UKG3 | 0.47 | 75 |
| $1 E 02$ | 0.22 | 68 | низ3 | -2.07 | 6 | UKH1 | 0.63 | 80 |
| GR11 | -0.23 | 56 | мтоо | 0.87 | 86 | UKH2 | 0.67 | 81 |
| GR12 | 0.28 | 70 | NL11 | 0.18 | 67 | Uкн3 | 0.85 | 86 |
| GR13 | 0.80 | 84 | NL12 | 0.33 | 71 | UKI | 0.72 | 82 |
| GR14 | 0.43 | 74 | NL13 | 0.45 | 75 | UKJ1 | 0.82 | 85 |
| GR21 | 0.57 | 78 | NL21 | 0.35 | 72 | UKI2 | 0.60 | 79 |
| GR22 | 0.80 | 84 | NL22 | 0.12 | 66 | UK13 | 1.03 | 90 |
| GR23 | 0.13 | 66 | NL23 | 1.38 | 100 | UKJ4 | 0.87 | 86 |
| GR24 | 0.05 | 64 | NL31 | 0.62 | 79 | Uкк1 | 0.97 | 89 |
| GR25 | -0.05 | 61 | NL32 | 0.57 | 78 | UKK2 | 1.07 | 92 |
| GR30 | 0.68 | 81 | NL33 | 0.52 | 77 | Uкк3 | 0.70 | 82 |
| GR41 | 0.50 | 76 | NL34 | 0.60 | 79 | UKK4 | 0.72 | 82 |
| GR42 | 0.55 | 77 | NL41 | 0.40 | 73 | UKL1 | 0.17 | 67 |
| GR43 | 0.48 | 76 | NL42 | 0.42 | 74 | UKL2 | 0.28 | 70 |
| ES11 | 0.03 | 63 | AT11 | -0.07 | 61 | UKM2 | -0.05 | 61 |
| ES12 | -0.13 | 59 | AT12 | -0.28 | 55 | UKM3 | -0.28 | 55 |
| ES13 | 0.57 | 78 | AT13 | -0.08 | 60 | UKM5 | -0.32 | 54 |
| ES21 | 0.35 | 72 | AT21 | 0.22 | 68 | UKM6 | -0.42 | 51 |
| ES22 | 0.47 | 75 | AT22 | 0.08 | 65 | UKNO | 0.20 | 68 |
| ES23 | 0.17 | 67 | AT31 | 0.02 | 63 |  |  |  |
| ES24 | 0.13 | 66 | AT32 | 0.12 | 66 |  |  |  |

Figure 5-18: Histogram of Health sub-score


Table 34: Health pillar sub-rank (from best to worst)

| Health |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | NL23 | 46 | ES51 | 91 | AT33 | 136 | AT32 | 181 | FR25 | 226 | PL33 |
| 2 | FI20 | 47 | NL32 | 92 | UKD5 | 137 | UKE1 | 182 | DE22 | 227 | PL52 |
| 3 | SE11 | 48 | DE11 | 93 | UKF3 | 138 | UKE4 | 183 | DEB2 | 228 | PL61 |
| 4 | UKE2 | 49 | GR42 | 94 | DK03 | 139 | DE25 | 184 | GR11 | 229 | BG41 |
| 5 | UKK2 | 50 | ITE4 | 95 | DE12 | 140 | DED1 | 185 | FR26 | 230 | PL31 |
| 6 | UKJ3 | 51 | ITC2 | 96 | GR12 | 141 | DK05 | 186 | FR63 | 231 | PL62 |
| 7 | ES30 | 52 | ITF3 | 97 | ES42 | 142 | DE13 | 187 | DE50 | 232 | BE35 |
| 8 | SE23 | 53 | NL33 | 98 | FR62 | 143 | AT22 | 188 | DE93 | 233 | PT20 |
| 9 | UKK1 | 54 | UKD2 | 99 | PT11 | 144 | PT17 | 189 | FR21 | 234 | PL43 |
| 10 | SE12 | 55 | GR41 | 100 | UKL2 | 145 | Fl18 | 190 | FR41 | 235 | PL51 |
| 11 | ITF6 | 56 | UKC2 | 101 | ES52 | 146 | DE24 | 191 | DE42 | 236 | PL42 |
| 12 | MT00 | 57 | GR43 | 102 | ES62 | 147 | DE60 | 192 | FR23 | 237 | BE32 |
| 13 | UKJ4 | 58 | IE01 | 103 | DEB3 | 148 | DEA1 | 193 | FR72 | 238 | SK02 |
| 14 | UKH3 | 59 | ES22 | 104 | DK02 | 149 | UKD3 | 194 | AT12 | 239 | PL11 |
| 15 | ITF4 | 60 | UKG3 | 105 | ES43 | 150 | GR24 | 195 | UKM3 | 240 | PL34 |
| 16 | UKJ1 | 61 | NL13 | 106 | ITF2 | 151 | DE27 | 196 | BE22 | 241 | PL41 |
| 17 | GR13 | 62 | DE30 | 107 | CZ01 | 152 | DED3 | 197 | BE23 | 242 | BG32 |
| 18 | GR22 | 63 | GR14 | 108 | DED2 | 153 | DEFO | 198 | CZ05 | 243 | CZ04 |
| 19 | ES64 | 64 | ITG2 | 109 | IE02 | 154 | ES11 | 199 | UKM5 | 244 | RO41 |
| 20 | ITG1 | 65 | UKE3 | 110 | AT21 | 155 | DE91 | 200 | FR52 | 245 | BG42 |
| 21 | SE22 | 66 | ITC4 | 111 | ES70 | 156 | AT31 | 201 | SIO2 | 246 | PT30 |
| 22 | SE21 | 67 | ITF5 | 112 | UKG1 | 157 | FR61 | 202 | CZ06 | 247 | BG31 |
| 23 | UKI | 68 | NL42 | 113 | UKNO | 158 | GR25 | 203 | PL21 | 248 | BG33 |
| 24 | UKK4 | 69 | BE21 | 114 | NL11 | 159 | FR51 | 204 | PL32 | 249 | RO21 |
| 25 | SE31 | 70 | FR71 | 115 | DK04 | 160 | UKM2 | 205 | FI13 | 250 | RO31 |
| 26 | SE33 | 71 | ITD1 | 116 | DEA4 | 161 | DE80 | 206 | DE41 | 251 | SK04 |
| 27 | UKK3 | 72 | NL41 | 117 | ES23 | 162 | ITD4 | 207 | CZ03 | 252 | SK03 |
| 28 | GR30 | 73 | UKF1 | 118 | FR92 | 163 | AT11 | 208 | FR94 | 253 | RO22 |
| 29 | ITE1 | 74 | UKG2 | 119 | UKL1 | 164 | FR43 | 209 | UKM6 | 254 | HU10 |
| 30 | ITE2 | 75 | DE21 | 120 | DEA2 | 165 | AT13 | 210 | CZ07 | 255 | R011 |
| 31 | ITF1 | 76 | DE26 | 121 | DEA3 | 166 | BE00 | 211 | DEEO | 256 | RO12 |
| 32 | UKF2 | 77 | UKD1 | 122 | DEA5 | 167 | Fl19 | 212 | CZ08 | 257 | BG34 |
| 33 | UKH2 | 78 | DE71 | 123 | ITD2 | 168 | DE72 | 213 | FR22 | 258 | RO42 |
| 34 | CYOO | 79 | ITD3 | 124 | PT16 | 169 | ES12 | 214 | FR30 | 259 | EE00 |
| 35 | SE32 | 80 | DE14 | 125 | UKC1 | 170 | FR24 | 215 | SK01 | 260 | HU22 |
| 36 | UKH1 | 81 | ES21 | 126 | GR23 | 171 | FR83 | 216 | BE33 | 261 | BE34 |
| 37 | FR10 | 82 | NL21 | 127 | ES24 | 172 | DE94 | 217 | PL63 | 262 | HU21 |
| 38 | ITC3 | 83 | DE73 | 128 | ES41 | 173 | FR53 | 218 | PT18 | 263 | HU23 |
| 39 | ITE3 | 84 | ES63 | 129 | ES61 | 174 | FR81 | 219 | CZ02 | 264 | HU33 |
| 40 | NL31 | 85 | NL12 | 130 | FR42 | 175 | BE25 | 220 | PT15 | 265 | HU31 |
| 41 | NL34 | 86 | AT34 | 131 | DE92 | 176 | DEB1 | 221 | SIO1 | 266 | HU32 |
| 42 | UKJ2 | 87 | ES53 | 132 | DEGO | 177 | DECO | 222 | FR93 | 267 | LV00 |
| 43 | DK01 | 88 | ITC1 | 133 | FR82 | 178 | FI1A | 223 | PL22 | 268 | LTOO |
| 44 | GR21 | 89 | LU00 | 134 | ITD5 | 179 | FR91 | 224 | PL12 |  |  |
| 45 | ES13 | 90 | UKD4 | 135 | NL22 | 180 | DE23 | 225 | RO32 |  |  |

### 5.5 Quality of Primary and Secondary Education

Indicators included in the pillar are discussed in Section 3.5. In the following we recall PISA indicators, related to educational outcomes, included in the analysis with their short names:

Indicators included, in brackets short names:

1. Low achievers in reading (reversed) (PISA_reading)
2. Low achievers in math (reversed)
(PISA_math)
3. Low achievers in science (reversed)
(PISA_science)

All three indicators have been reversed in order to have the same polarity with respect to competitiveness (the higher the better).

As discussed in Section 3.5, the initial set of indicators originally considered for this pillar comprised more indicators, with the intention of describing also the inputs to the education system. To this aim the following indicators have been examined: student to teacher ratio, financial aid ISCED level 1 to 4 , public expenditures level 1 to 4 and rates of participation in education of 4 year old pupils. All these indicators are at the country level. Although, a preliminary analysis of these indicators showed that they are very poorly related with each other. None of their correlation coefficients is statistical significant (Table 35) and, accordingly, PCA loadings have almost the same value across dimensions (Table 36). This suggests that the indicators have very little in common. They represent a mix of different aspects rather then mostly describing the quality of basic education. They were therefore dropped from the analysis.

Table 35: Correlation matrix for additional indicators originally included in the pillar of Quality of Primary and Secondary Education

| Correlation Matrix |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Student to teacher ratio (reversed) | $\begin{aligned} & \text { financial } \\ & \text { aid } \\ & \text { level } 1.4 \end{aligned}$ | public expenditure level 24 | public expenditure level 1 | Early Education (reversed) |
| Correlation | student_teacher_ratio_reversed | 1.000 | -. 230 | -. 013 | . 104 | -. 197 |
|  | financial_aid_1_4 | -. 230 | 1.000 | . 073 | -. 044 | -. 108 |
|  | public_expenditure_2_4 | -. 013 | . 073 | 1.000 | -. 016 | . 114 |
|  | public_expenditure_1 | . 104 | -. 044 | -. 016 | 1.000 | -. 053 |
|  | early_education_reversed | -. 197 | -. 108 | . 114 | -. 053 | 1.000 |
| Sig. <br> (1-tailed) | student_teacher_ratio_reversed |  | . 151 | . 476 | . 319 | . 184 |
|  | financial_aid_1_4 | . 151 |  | . 370 | . 421 | . 321 |
|  | public_expenditure_2_4 | . 476 | . 370 |  | . 470 | . 307 |
|  | public_expenditure_1 | . 319 | . 421 | . 470 |  | . 407 |
|  | early_education reversed | . 184 | . 321 | . 307 | . 407 |  |

Table 36: PCA results on the set of additional indicators originally included in the pillar of Quality of Primary and Secondary Education

| Component | Initial Eigenvalues |  |  |
| :--- | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 1.331 | 26.620 | 26.620 |
| 2 | 1.118 | 22.353 | 48.973 |
| 3 | .999 | 19.988 | 68.961 |
| 4 | .937 | 18.732 | 87.693 |
| 5 | .615 | 12.307 | 100.000 |

## UNIVARIATE ANALYSIS

All three indicators included in the analysis are at the country level. There is no data for Cyprus and Malta leading to $7.41 \%$ of missing data, which is within the threshold of missing data defined in Section 4.2.

Table 37: Descriptive statistics of Quality of primary and secondary education indicators

| Name of indicator | Low achievers in reading | Low achievers in math | Low achievers in science |
| :---: | :---: | :---: | :---: |
| description of indicator | \% of pupils, 15 years old, with reading proficiency level 1 and low on PISA | $\%$ of pupils,15 years old, with math proficiency level 1 and low on PISA | $\%$ of pupils, 15 years old, with science proficiency level 1 and low on PISA |
| source | OECD Programme for International Student Assessment (PISA) | OECD Programme for International Student Assessment (PISA) | OECD Programme for International Student Assessment (PISA) |
| reference year | 2006 | 2006 | 2006 |
| \% of missing values <br> mean value <br> standard deviation (unbiased) <br> coefficient of variation | $\begin{gathered} 7.41 \\ 22.54 \\ 10.49 \\ 0.47 \end{gathered}$ | $\begin{gathered} 7.41 \\ 22.76 \\ 10.93 \\ 0.48 \end{gathered}$ | $\begin{gathered} 7.41 \\ 19.27 \\ 8.93 \\ 0.46 \end{gathered}$ |
| maximum value <br> region corresponding to maximum value <br> minimum value <br> region corresponding to minimum value | $\begin{gathered} 53.50 \\ \text { RO } \\ 4.80 \\ \mathrm{FI} \end{gathered}$ | $\begin{gathered} 53.30 \\ \text { BG } \\ 6.00 \\ \mathrm{FI} \end{gathered}$ | $\begin{gathered} 46.90 \\ \text { RO } \\ 4.10 \\ \text { FI } \end{gathered}$ |

## How do EU countries score in each of the indicators?

Bulgaria and Romania are the countries with the highest percentage of low achievers in reading, math and science. Finland is the top performer in all three fields, together with Ireland (for reading), the Netherlands (for math), and Estonia (for science).


Figure 5-19. Best and worst performing countries for each indicator Quality of primary and secondary education

Table 38 shows the histograms of the three indicators. They all show positive skewness and have been transformed with the Box-Cox method.

Table 38 Histograms of Quality of Primary \& Secondary education indicators PISA reading



PISA math


Trasformed indicator. PISA-math Box-Cox transformation, lambda $=0.05$ skewness $=0.30633$


PISA science


## MULTIVARIATE ANALYSIS

The correlation coefficients between the three PISA indicators clearly indicate a very high level of correlation (Table 39). Accordingly, the PCA analysis highlights a single major dimension (Figure 5-20), which accounts for more than $95 \%$ of total variation (Table 41) and is equally described by all the indicators, as may be seen by the table of component loadings (Table 40). The pillar describing the level of compulsory education in EU regions is statistically consistent and well balanced, thus supporting the choice of the sub-score computation as simple average of the normalized indicators. The geographical distribution of the sub-score across EU countries is displayed in Figure 5-21 while its histogram is shown in Figure 5-22. Countries, reordered from best to worst performers in this pillar are displayed in Table 43.

Table 39: Correlation matrix between indicators included in the pillar on Quality of primary and secondary education

Correlation Matrix

|  |  | PISA_reading <br> _reversed | PISA_math_ <br> reversed | PISA_science <br> _reversed |
| :--- | :--- | ---: | ---: | ---: |
| Correlation | PISA_reading_reversed | 1.000 | .939 | .927 |
|  | PISA_math_reversed | .939 | 1.000 | .937 |
|  | PISA_science_reversed | .927 | .937 | 1.000 |
| Sig. (1-tailed) | PISA_reading_reversed |  | .000 | .000 |
|  | PISA_math_reversed | .000 |  | .000 |
|  | PISA_science_reversed | .000 | .000 |  |

Scree Plot


Figure 5-20: PCA analysis of the pillar on Quality of primary and secondary education- eigenvalues

Table 40: PCA analysis of the pillar on Quality of primary and secondary education: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |
| :--- | ---: | ---: | ---: |
|  | 1 | 2 | 3 |
| PISA_reading_reversed | .977 | -.181 | .113 |
| PISA_math_reversed | .980 | -.019 | -.197 |
| PISA_science_reversed | .976 | .200 | .084 |

Extraction Method: Principal Component Analysis.
a. 3 components extracted.

Table 41: PCA analysis for the pillar on Quality of primary and secondary education: explained variance

Total Variance Explained

| Component | Initial Eigenvalues |  |  | Extraction Sums of Squared Loadings |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative $\%$ | Total | $\%$ of Variance | Cumulative $\%$ |
| 1 | 2.868 | 95.603 | 95.603 | 2.868 | 95.603 | 95.603 |
| 2 | .073 | 2.447 | 98.050 | .073 | 2.447 | 98.050 |
| 3 | .059 | 1.950 | 100.000 | .059 | 1.950 | 100.000 |

Extraction Method: Principal Component Analysis.


Figure 5-21: Quality of primary and secondary education pillar sub-score. (min-max normalized values)

Table 42: Quality of primary and secondary education sub-scores

| country | Subscore | Min_max <br> normalized <br> subscore |
| :---: | :---: | :---: |
| BE | 0.5 | 43 |
| BG | -2.5 | 2 |
| CZ | 0.23 | 39 |
| DK | 0.87 | 48 |
| DE | 0.43 | 42 |
| EE | 2.07 | 65 |
| IE | 1.13 | 52 |
| GR | -0.8 | 25 |
| ES | -0.3 | 32 |
| FR | -0.1 | 35 |
| IT | -0.87 | 24 |
| CY |  |  |
| LV | 0.23 | 39 |
| LT | -0.27 | 33 |
| LU | -0.23 | 33 |
| HU | 0.37 | 41 |
| MT |  |  |
| NL | 1.43 | 56 |
| AT | 0.3 | 40 |
| PL | 0.53 | 44 |
| PT | -0.5 | 29 |
| RO | -2.63 | 0 |
| SI | 0.87 | 48 |
| SK | -0.27 | 33 |
| FI | 4.63 | 100 |
| SE | 0.7 | 46 |
| UK | 0.4 | 42 |
|  |  |  |



Figure 5-22: Histogram of Quality of primary and secondary education sub-scores

Table 43: Quality of primary and secondary education sub-rank (from best to worst)

| Quality of primary and secondary <br> education |  |  |
| :---: | :---: | ---: |
| 1 | FI | Finland |
| 2 | EE | Estonia |
| 3 | NL | Netherlands |
| 4 | IE | Ireland |
| 5 | DK | Denmark |
| 6 | SI | Slovenia |
| 7 | SE | Sweden |
| 8 | PL | Poland |
| 9 | BE | Belgium |
| 10 | DE | Germany |
| 11 | UK | United Kingdom |
| 12 | HU | Hungary |
| 13 | AT | Austria |
| 14 | CZ | Czech republic |
| 15 | LV | Latvia |
| 16 | FR | France |
| 17 | LU | Luxembourg |
| 18 | LT | Lithuania |
| 19 | SK | Slovakia |
| 20 | ES | Spain |
| 21 | PT | Portugal |
| 22 | GR | Greece |
| 23 | IT | Italy |
| 24 | BG | Bulgaria |
| 25 | RO | Romania |
| -- | CY | Cyprus |
| -- | MT | Malta |
|  |  |  |

### 5.6 Higher Education/Training and Lifelong Learning

The full description of indicators included in the pillar is due in Section 3.6. In the following we recall them together with the short names used for the statistical analysis.

## Indicators included, in brackets short names:

1. Share of population 25-64 with higher educational attainment
(tertiary_ed_attainment)
2. Share of population 25-64 involved in education and training
(lifelong_learning)
3. Share of population with low education (reversed) (early_school_leavers)
4. Share of population at $>60$ minutes from university (reversed)
(accessibility)
5. Total expenditures on tertiary education as GDP percentage
(tertiary_ed_expenditure)

Indicators 3. and 4. have been reversed in order to have the same polarity with respect to competitiveness.

## Imputation of missing data

Total expenditure on tertiary level of education (Tertiary_ed_expenditure) is available at the country level. 2006 data has been used with the exception of Denmark, Estonia, Greece, Poland and Malta where 2005 has been used due to lack of more recent data. The number of students in tertiary education, available at the regional level from the Eurostat Education and Training database, is considered as the best proxy for imputing Tertiary_ed_expenditure at the NUTS 2 level. 2006 figures for the number of students (ISCED 5-6 level) in the 20-29 years age brackets at the NUTS 2 level have been used. For Greece and Ireland 2005 figures have been used due to lack of 2006 data. For the UK and Germany, NUTS 2 data on the
number of students has been imputed to the NUTS 2 level. Similarly, for Denmark NUTS 0 level has been imputed to the NUTS 2 level.

## Univariate analysis

Table 44 presents the descriptive statistics for the five indicators included in the Higher Education/Training and Lifelong Learning pillar. The first four indicators are at the regional level, with very low percentage of missing data. The higher education attainment, accessibility to universities and lifelong learning refer to 2007 while data on early school leavers is an average of 2006 and 2007. Data on tertiary education expenditure refers to 2006. The accessibility indicator shows a high coefficient of variation, explained by the fact that there are various regions in Europe, often times due to their geographical position, which are located far (in this case, defined as more than $60^{\prime}$ drive) from a university.

Table 44: Descriptive statistics of Higher Education/Training and Lifelong Learning indicators

| Indicator | Population 25-64 with higher education | Lifelong learning | Early school leavers | Accessibility to universities | Higher education expenditure |
| :---: | :---: | :---: | :---: | :---: | :---: |
| description | Population aged 25-64 with higher educational attainment (ISCED5_6), $\%$ of total population of age group | Participation of adults aged 25-64 in education and training, \% of population aged 25-64 | People with at most lower secondary education and not in further education or training, \% of total population aged 18-24 | Population living at more than 60 minutes from the nearest university, \% of total population | Total public expenditure on education as \% of GDP, at tertiary level of education (ISCED 5-6) |
| source | Eurostat, LFS | Eurostat Regional <br> Education Statistics | Eurostat Structural Indicators | Nordregio/EuroGeographics/ GISCO/ EEA ETC-TE | Total public expenditure on education as \% of GDP, at tertiary level of education (ISCED 5-6) |
| reference year | 2007 | 2007 | average 2006-2007 | 2006 | 2006 |
| $\%$ of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 1.49 \\ 23.04 \\ 7.95 \\ 0.35 \end{gathered}$ | $\begin{aligned} & 1.49 \\ & 9.74 \\ & 6.62 \\ & 0.68 \end{aligned}$ | $\begin{gathered} 1.49 \\ 15.76 \\ 8.67 \\ 0.55 \end{gathered}$ | $\begin{gathered} 2.24 \\ 11.76 \\ 21.39 \\ 1.82 \end{gathered}$ | $\begin{gathered} 11.11 \\ 1.21 \\ 0.39 \\ 0.32 \end{gathered}$ |
| maximum value <br> region corresponding to maximum value <br> minimum value <br> region corresponding to minimum value | $\begin{gathered} 42.50 \\ \text { ES21 } \\ 7.26 \\ \text { CZ04 } \end{gathered}$ | $\begin{gathered} 29.19 \\ \text { DK01 } \\ 0.29 \\ \text { GR41 } \end{gathered}$ | $\begin{array}{r} 56.46 \\ \text { PT20 } \\ 2.82 \\ \text { PL21 } \end{array}$ | $\begin{array}{r} 99.99 \\ \text { GR13 } \\ 0.00 \\ \text { BE00 } \end{array}$ | $\begin{gathered} 2.27 \\ \text { DK } \\ 0.73 \\ \text { BG } \end{gathered}$ |

## How do EU regions score in each of the indicators?

With regards to tertiary education attainment, we can see that a number of UK regions are performing very well while the northern regions of Romania show some of the lowest performance. Northern European regions perform best on the lifelong learning indicator while we can see parts of Romania, Bulgaria and Greece having the lowest percentage of the population participating in lifelong learning activities. A number of Polish regions perform very well on the indicator on early school leavers while Mediterranean regions, especially in Portugal and Spain, are lagging significantly behind. German regions demonstrate a very dense network of universities while Greek regions have the worst accessibility to universities. Denmark and Finland are the countries with highest expenditure on tertiary education as
percentage of GDP while Bulgaria and Italy have the lowest. In general, we could see a rather distinct division between the performance of Northern and Southern European regions in terms of the quality of higher education and training systems.


Figure 5-23. Best and worst performing regions for each indicator
Higher Education/Training and Lifelong learning

[^12]As shown in Table 45, only two of the Higher Education/Training and Lifelong Learning indicators, higher education attainment and lifelong learning, have not been transformed. Data on early school leavers and total expenditure on tertiary education demonstrates high positive skewness and has been transformed using the Box-Cox method as described in Section 4.3. Furthermore, similar to the infrastructure indicators, the data on university accessibility has a lot of zero values and has been, thus, transformed logarithmically as explained in Section 4.3. The graphs show, where relevant, both the distribution of the original data as well as the as that of the transformed indicator.

Table 45: Histograms of Higher Education/Training and Lifelong Learning indicators
Tertiary ed attainment


Lifelong learning



## Tertiary ed expenditure



## Multivariate analysis

The PCA analysis highlights the presence of two prevalent dimensions which together explain about $62 \%$ of total variation (Table 48). The first dimension, which accounts for $40 \%$ of the variance, is described by Tertiary_ed_attainment, Lifelong_learning and Accessibility (Table 47). Early_school_leavers and Tertiary_ed_expenditure contribute to the second component, which explains $22 \%$ of total variation. From the analysis of the scree plot (Figure 5-24) it can be seen that the presence of one unique dimension cannot be fully supported in this case.

Figure 5-25 shows the map of the Higher education sub-score, computed as an arithmetic mean of the five standardized indicators (values in Table 49). In Figure 5-26 the histogram of the higher education sub-score is displayed while Table 50 shows the ranking of regions in this pillar.

Table 46: Correlation matrix between indicators included inthe Higher Education/Training and Lifelong Learning pillar

| Correlation Matrix |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Tertiary_ed attainment | Lifelong_ learning | Early_school_ leavers reversē | Accessibility reversed | Tertiary_ed expenditure |
| Correlation | Tertiary_ed_attainment | 1.000 | .667 | . 158 | . 334 | . 187 |
|  | Lifelong_learning | . 667 | 1.000 | . 121 | . 256 | . 048 |
|  | Early_school_leavers_ reversed | . 158 | . 121 | 1.000 | . 237 | . 226 |
|  | Accessibility_reversed | . 334 | . 256 | . 237 | 1.000 | . 049 |
|  | Tertiary_ed_expenditure | . 187 | . 048 | . 226 | . 049 | 1.000 |
| Sig. (1-tailed) | Tertiary_ed_attainment |  | . 000 | . 005 | . 000 | . 001 |
|  | Lifelong_learning | . 000 |  | . 025 | . 000 | . 220 |
|  | Early_school_leavers_ reversed | . 005 | . 025 |  | . 000 | . 000 |
|  | Accessibility_reversed | . 000 | . 000 | . 000 |  | . 219 |
|  | Tertiary_ed_expenditure | . 001 | . 220 | . 000 | . 219 |  |



Figure 5-24: PCA analysis of the Higher Education/Training and Lifelong learning pillar eigenvalues

Table 47: PCA analysis of the Higher Education/Training and Lifelong Learning pillar: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 5 |
| Tertiary_ed_attainment | . 851 | -. 244 | . 226 | -. 025 | -. 406 |
| Lifelong_learning | . 786 | -. 399 | . 199 | -. 216 | . 369 |
| Early_school_leavers_ reversed | . 442 | . 643 | -. 365 | -. 507 | -. 021 |
| Accessibility_reversed | . 602 | . 020 | -. 605 | . 517 | . 061 |
| Tertiary_ed_expenditure | . 324 | . 696 | . 548 | . 321 | . 087 |

Extraction Method: Principal Component Analysis.
a. 5 components extracted.

Table 48: PCA analysis for the Higher Education/Training and Lifelong learning pillar: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 2.006 | 40.118 | 40.118 |
| 2 | 1.117 | 22.339 | 62.457 |
| -3 | .890 | 17.801 | 80.257 |
| 4 | .675 | 13.500 | 93.758 |
| 5 | .312 | 6.242 | 100.000 |



Figure 5-25: Higher Education/Training and Lifelong Learning sub-score (Min-Max normalized values)

Table 49: Higher Education/Training and Lifelong Learning sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | 0.92 | 89 | Es30 | 0.48 | 81 | AT33 | -0.04 | 71 |
| BE21 | 0.52 | 82 | ES41 | -0.32 | 66 | AT34 | -0.60 | 61 |
| BE22 | 0.30 | 78 | ES42 | -1.03 | 53 | PL11 | -0.07 | 71 |
| BE23 | 0.57 | 83 | ES43 | -1.03 | 53 | PL12 | 0.68 | 85 |
| BE25 | 0.34 | 78 | ES51 | -0.07 | 71 | PL21 | 0.36 | 79 |
| BE32 | 0.00 | 72 | Es52 | -0.18 | 69 | PL22 | 0.40 | 79 |
| вез3 | 0.27 | 77 | ES53 | -1.15 | 51 | PL31 | 0.05 | 73 |
| BE34 | -0.13 | 70 | ES61 | -0.47 | 64 | PL32 | 0.03 | 73 |
| BE35 | 0.12 | 74 | Es62 | -0.50 | 63 | PL33 | -0.07 | 71 |
| BG31 | -1.62 | 43 | Es63 | -1.98 | 36 | PL34 | -0.27 | 67 |
| BG32 | -0.79 | 58 | ES64 | -1.96 | 37 | PL41 | -0.05 | 71 |
| BG33 | -0.70 | 59 | Es70 | -0.79 | 58 | PL42 | -0.37 | 65 |
| BG34 | -1.10 | 52 | FR10 | 0.85 | 88 | PL43 | -0.64 | 61 |
| BG41 | 0.21 | 76 | FR21 | -0.17 | 69 | PL51 | 0.13 | 75 |
| BG42 | -0.91 | 56 | FR22 | -0.38 | 65 | PL52 | -0.29 | 67 |
| czo1 | 0.97 | 90 | FR23 | -0.18 | 69 | PL61 | -0.36 | 66 |
| cz02 | -0.41 | 65 | FR24 | -0.55 | 62 | PL62 | -0.67 | 60 |
| cz03 | 0.01 | 72 | FR25 | -0.41 | 65 | PL63 | -0.23 | 68 |
| czo4 | -0.45 | 64 | FR26 | -0.81 | 57 | PT11 | -0.63 | 61 |
| cz05 | 0.05 | 73 | FR30 | 0.05 | 73 | PT15 | -1.31 | 48 |
| czo6 | 0.04 | 73 | FR41 | 0.08 | 74 | PT16 | -0.49 | 63 |
| czo7 | -0.12 | 70 | FR42 | 0.16 | 75 | PT17 | -0.15 | 69 |
| cz08 | -0.11 | 70 | FR43 | -0.37 | 65 | PT18 | -1.09 | 52 |
| DK01 | 1.43 | 98 | FR51 | -0.04 | 71 | PT20 | -1.94 | 37 |
| DK02 | 0.77 | 86 | FR52 | 0.59 | 83 | PT30 | -1.77 | 40 |
| DK03 | 0.86 | 88 | FR53 | -0.54 | 62 | RO11 | -0.91 | 56 |
| DK04 | 0.89 | 88 | FR61 | -0.29 | 67 | RO12 | -0.86 | 57 |
| DK05 | 0.89 | 88 | FR62 | 0.02 | 73 | RO21 | -1.05 | 53 |
| DE11 | 0.22 | 76 | FR63 | -0.74 | 59 | RO22 | -1.26 | 49 |
| DE12 | 0.19 | 76 | FR71 | 0.15 | 75 | R031 | -1.23 | 50 |
| DE13 | 0.10 | 74 | FR72 | -0.36 | 66 | RO32 | 0.28 | 77 |
| DE14 | 0.11 | 74 | FR81 | -0.32 | 66 | RO41 | -1.13 | 52 |
| DE21 | 0.38 | 79 | FR82 | -0.17 | 69 | RO42 | -0.87 | 56 |
| DE22 | -0.18 | 69 | FR83 | -1.48 | 45 | S101 | 1.13 | 93 |
| DE23 | -0.18 | 69 | FR91 | -2.66 | 24 | S102 | 1.03 | 91 |
| DE24 | -0.26 | 67 | FR92 | -2.61 | 25 | sk01 | 0.96 | 90 |
| DE25 | -0.08 | 71 | FR93 | -3.97 | 0 | Sk02 | 0.29 | 77 |
| DE26 | -0.01 | 72 | FR94 | -1.91 | 37 | Sk03 | -0.16 | 69 |
| DE27 | -0.14 | 70 | ITC1 | -0.71 | 59 | Sk04 | -0.55 | 62 |
| DE30 | 0.40 | 79 | ITC2 | -2.15 | 33 | F13 | 0.53 | 82 |
| DE41 | -0.10 | 70 | ITC3 | -0.54 | 62 | F118 | 1.53 | 100 |
| DE42 | 0.16 | 75 | ITC4 | -0.38 | 65 | F19 | 1.10 | 92 |
| DE50 | -0.19 | 69 | ITD1 | -1.84 | 39 | F11A | 0.70 | 85 |
| DE60 | 0.09 | 74 | ITD2 | -0.67 | 60 | F120 | -0.49 | 63 |
| DE71 | 0.18 | 75 | ITD3 | -0.51 | 63 | SE11 | 1.03 | 91 |
| De72 | -0.01 | 72 | ITD4 | -0.52 | 63 | SE12 | 0.78 | 86 |
| De73 | -0.16 | 69 | ItD5 | -0.34 | 66 | SE21 | 0.26 | 77 |
| DE80 | -0.05 | 71 | ITE1 | -0.45 | 64 | SE22 | 1.02 | 91 |
| DE91 | -0.10 | 70 | ITE2 | -0.38 | 65 | SE23 | 0.89 | 88 |
| DE92 | -0.17 | 69 | ite3 | -0.65 | 60 | SE31 | -0.01 | 72 |
| DE93 | -0.44 | 64 | ITE4 | 0.08 | 74 | SE32 | -0.08 | 71 |
| DE94 | -0.30 | 67 | ITF1 | -0.38 | 65 | SE33 | 0.29 | 77 |
| DEA1 | -0.07 | 71 | ITF2 | -0.80 | 58 | UKC1 | 0.19 | 76 |
| DEA2 | 0.18 | 75 | ITF3 | -0.68 | 60 | UKC2 | 0.35 | 79 |
| DEA3 | -0.04 | 71 | ITF4 | -0.91 | 56 | UKD1 | -0.43 | 64 |
| DEA4 | -0.20 | 69 | ITF5 | -1.26 | 49 | UKD2 | 0.37 | 79 |
| DEA5 | -0.15 | 69 | ITF6 | -1.03 | 53 | UKD3 | 0.47 | 81 |
| DEB1 | -0.30 | 67 | ItG1 | -1.05 | 53 | UKD4 | 0.41 | 80 |
| DEB2 | -0.19 | 69 | ITG2 | -1.12 | 52 | UKD5 | 0.31 | 78 |
| Deb3 | -0.09 | 71 | cyoo | 0.35 | 79 | UKE1 | -0.01 | 72 |
| DECO | -0.53 | 63 | Lvoo | -0.07 | 71 | UKE2 | 0.45 | 80 |
| DED1 | 0.21 | 76 | Lтоо | 0.19 | 76 | UKE3 | 0.10 | 74 |
| DED2 | 0.36 | 79 | Lu00 | 0.14 | 75 | UKE4 | 0.42 | 80 |
| DED3 | 0.30 | 78 | HU10 | 0.33 | 78 | UKF1 | 0.39 | 79 |
| DEEO | -0.03 | 72 | HU21 | -0.25 | 68 | UKF2 | 0.45 | 80 |
| DEFO | -0.27 | 67 | HU22 | -0.17 | 69 | UKF3 | -0.09 | 71 |
| DEGO | 0.23 | 76 | HU23 | -0.45 | 64 | UKG1 | 0.60 | 83 |
| EEOO | 0.20 | 76 | HU31 | -0.37 | 65 | UKG2 | 0.33 | 78 |
| $1 E 01$ | 0.19 | 76 | HU32 | -0.37 | 65 | UKG3 | 0.45 | 80 |
| $1 E 02$ | 0.78 | 86 | нuz3 | -0.35 | 66 | UKH1 | 0.45 | 80 |
| GR11 | -0.95 | 55 | мтоо | -0.86 | 57 | UKH2 | 0.58 | 83 |
| GR12 | -0.31 | 67 | NL11 | 0.89 | 88 | икнз | 0.22 | 76 |
| GR13 | -0.83 | 57 | NL12 | 0.23 | 76 | UKI | 1.24 | 95 |
| GR14 | -0.89 | 56 | NL13 | -0.05 | 71 | UKJ1 | 0.88 | 88 |
| GR21 | -0.83 | 57 | NL21 | 0.67 | 84 | UKJ2 | 1.01 | 91 |
| GR22 | -1.64 | 42 | NL22 | 0.67 | 84 | UKJ3 | 0.46 | 81 |
| GR23 | -0.63 | 61 | NL23 | -0.02 | 72 | UKJ4 | 0.34 | 78 |
| GR24 | -1.36 | 47 | NL31 | 1.18 | 94 | UKK1 | 0.84 | 87 |
| GR25 | -1.37 | 47 | NL32 | 0.94 | 89 | UKK2 | 0.37 | 79 |
| GR30 | 0.37 | 79 | NL33 | 0.96 | 90 | บкк3 | -0.30 | 67 |
| GR41 | -1.43 | 46 | NL34 | 0.03 | 73 | UKK4 | 0.48 | 81 |
| GR42 | -1.57 | 44 | NL41 | 0.74 | 86 | UKL1 | 0.23 | 76 |
| GR43 | -0.91 | 56 | NL42 | 0.41 | 80 | UKL2 | 0.56 | 82 |
| ES11 | -0.42 | 65 | AT11 | -0.09 | 71 | UKM2 | 1.30 | 96 |
| ES12 | -0.50 | 63 | AT12 | -0.14 | 70 | UKм3 | 0.76 | 86 |
| ES13 | -0.53 | 63 | AT13 | 0.73 | 85 | UKM5 | 1.31 | 96 |
| ES21 | 0.63 | 84 | AT21 | 0.07 | 73 | UKM6 | 0.42 | 80 |
| ES22 | 0.19 | 76 | AT22 | 0.22 | 76 | UKNO | 0.26 | 77 |
| ES23 | -0.53 | 63 | AT31 | 0.36 | 79 |  |  |  |
| ES24 | -0.40 | 65 | AT32 | -0.09 | 71 |  |  |  |



Figure 5-26: Histogram of Higher Education/Training
Table 50: Higher Education/Training and Lifelong Learning pillar sub-rank
(from best to worst)

| Higher education and training |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | F118 | 46 | UKD3 | 91 | IE01 | 136 | DE25 | 181 | FR43 | 226 | FR26 |
| 2 | DK01 | 47 | UKJ3 | 92 | ES22 | 137 | SE32 | 182 | HU31 | 227 | GR13 |
| 3 | UKM5 | 48 | UKE2 | 93 | LTOO | 138 | DEB3 | 183 | HU32 | 228 | GR21 |
| 4 | UKM2 | 49 | UKF2 | 94 | UKC1 | 139 | AT11 | 184 | PL42 | 229 | MT00 |
| 5 | UKI | 50 | UKG3 | 95 | DE71 | 140 | AT32 | 185 | FR22 | 230 | RO12 |
| 6 | NL31 | 51 | UKH1 | 96 | DEA2 | 141 | UKF3 | 186 | ITC4 | 231 | RO42 |
| 7 | SIO1 | 52 | UKE4 | 97 | DE42 | 142 | DE41 | 187 | ITE2 | 232 | GR14 |
| 8 | FI19 | 53 | UKM6 | 98 | FR42 | 143 | DE91 | 188 | ITF1 | 233 | BG42 |
| 9 | SIO2 | 54 | NL42 | 99 | FR71 | 144 | CZ08 | 189 | ES24 | 234 | GR43 |
| 10 | SE11 | 55 | UKD4 | 100 | Lu00 | 145 | CZ07 | 190 | CZO2 | 235 | ITF4 |
| 11 | SE22 | 56 | DE30 | 101 | PL51 | 146 | BE34 | 191 | FR25 | 236 | R011 |
| 12 | UKJ2 | 57 | PL22 | 102 | BE35 | 147 | DE27 | 192 | ES11 | 237 | GR11 |
| 13 | CZ01 | 58 | UKF1 | 103 | DE14 | 148 | AT12 | 193 | UKD1 | 238 | ES42 |
| 14 | NL33 | 59 | DE21 | 104 | DE13 | 149 | DEA5 | 194 | DE93 | 239 | ES43 |
| 15 | SK01 | 60 | GR30 | 105 | UKE3 | 150 | PT17 | 195 | CZO4 | 240 | ITF6 |
| 16 | NL32 | 61 | UKD2 | 106 | DE60 | 151 | DE73 | 196 | ITE1 | 241 | ITG1 |
| 17 | BEOO | 62 | UKK2 | 107 | FR41 | 152 | SK03 | 197 | HU23 | 242 | RO21 |
| 18 | DK04 | 63 | DED2 | 108 | ITE4 | 153 | DE92 | 198 | ES61 | 243 | PT18 |
| 19 | DK05 | 64 | AT31 | 109 | AT21 | 154 | FR21 | 199 | PT16 | 244 | BG34 |
| 20 | NL11 | 65 | PL21 | 110 | CZ05 | 155 | FR82 | 200 | F120 | 245 | ITG2 |
| 21 | SE23 | 66 | cyoo | 111 | FR30 | 156 | HU22 | 201 | ES12 | 246 | RO41 |
| 22 | UKJ1 | 67 | UKC2 | 112 | PL31 | 157 | DE22 | 202 | ES62 | 247 | ES53 |
| 23 | DK03 | 68 | BE25 | 113 | CZ06 | 158 | DE23 | 203 | ITD3 | 248 | RO31 |
| 24 | FR10 | 69 | UKJ4 | 114 | NL34 | 159 | ES52 | 204 | ITD4 | 249 | ITF5 |
| 25 | UKK1 | 70 | HU10 | 115 | PL32 | 160 | FR23 | 205 | DECO | 250 | RO22 |
| 26 | IEO2 | 71 | UKG2 | 116 | FR62 | 161 | DE50 | 206 | ES13 | 251 | PT15 |
| 27 | SE12 | 72 | UKD5 | 117 | CZO3 | 162 | DEB2 | 207 | ES23 | 252 | GR24 |
| 28 | DK02 | 73 | BE22 | 118 | BE32 | 163 | DEA4 | 208 | FR53 | 253 | GR25 |
| 29 | UKM3 | 74 | DED3 | 119 | DE26 | 164 | PL63 | 209 | ITC3 | 254 | GR41 |
| 30 | NL41 | 75 | SK02 | 120 | DE72 | 165 | HU21 | 210 | FR24 | 255 | FR83 |
| 31 | AT13 | 76 | SE33 | 121 | SE31 | 166 | DE24 | 211 | SK04 | 256 | GR42 |
| 32 | FI1A | 77 | RO32 | 122 | UKE1 | 167 | DEFO | 212 | AT34 | 257 | BG31 |
| 33 | PL12 | 78 | BE33 | 123 | NL23 | 168 | PL34 | 213 | GR23 | 258 | GR22 |
| 34 | NL21 | 79 | SE21 | 124 | DEEO | 169 | FR61 | 214 | PT11 | 259 | PT30 |
| 35 | NL22 | 80 | UKNO | 125 | DEA3 | 170 | PL52 | 215 | PL43 | 260 | ITD1 |
| 36 | ES21 | 81 | DEGO | 126 | FR51 | 171 | DE94 | 216 | ITE3 | 261 | FR94 |
| 37 | UKG1 | 82 | NL12 | 127 | AT33 | 172 | DEB1 | 217 | ITD2 | 262 | PT20 |
| 38 | FR52 | 83 | UKL1 | 128 | DE80 | 173 | UKK3 | 218 | PL62 | 263 | ES64 |
| 39 | UKH2 | 84 | DE11 | 129 | NL13 | 174 | GR12 | 219 | ITF3 | 264 | ES63 |
| 40 | BE23 | 85 | AT22 | 130 | PL41 | 175 | ES41 | 220 | BG33 | 265 | ITC2 |
| 41 | UKL2 | 86 | UKH3 | 131 | DEA1 | 176 | FR81 | 221 | ITC1 | 266 | FR92 |
| 42 | F113 | 87 | BG41 | 132 | ES51 | 177 | ITD5 | 222 | FR63 | 267 | FR91 |
| 43 | BE21 | 88 | DED1 | 133 | LV00 | 178 | HU33 | 223 | BG32 | 268 | FR93 |
| 44 | ES30 | 89 | EEOO | 134 | PL11 | 179 | FR72 | 224 | ES70 |  |  |
| 45 | UKK4 | 90 | DE12 | 135 | PL33 | 180 | PL61 | 225 | ITF2 |  |  |

### 5.7 Labor market efficiency

As discussed in Section 3.7, indicators included in the pillar are:
Indicators included in the pillar (in brackets short names):

1. Employment rate, not including agriculture (Empl_rate)
2. Long term unemployment (reversed) (Long_term_unempl)
3. Unemployment (reversed)
(Unemployment)
4. Job Mobility
(Job_mobility)
5. Labor productivity
(Labor_productivity)
6. Female-male unemployment rate difference (reversed)(Gender_balance_unemp)
7. Male-female employment rate difference (reversed) (Gender_balance_empl)
8. Female unemployment (reversed)
(Female_unemployment)
9. Labor Market Policy
(LMP)

All indicators are available at the NUTS2 level except for LMP which is available only at the country level. For this indicator the imputation method described in Section 4.2.1 is adopted. The indicator labor productivity (Labor_productivity) has the highest correlartion, 0.66, with labor market policy (LMP) at the country level. All the other correlations are, in absolute values, lower than 0.46 . The regional values of Labor_productivity are then used to impute labor market policies values at the NUTS2 level. In the following the multivariate analysis including the imputed LMP indicator is described.

The indicator on employment rate does not include employment in the agricultural sector as it is considered not a driving factor for competitiveness.

The indicator on male-female employment rate (Gender balance employment) has been transformed from the original female-male employment rate difference by multiplying the original indicators with (-1) due to data transformation needs.

It is worth noting that for the gender balance unemployment indicator, $28 \%$ of the regions show a negative value which means that female unemployment rate is lower than male. One could argue if this can be considered a positive or a negative aspect with respect to labor market efficiency. In order to avoid the possible over-awarding of regions with such values, we have decided to censor at the 0 value, i.e. all negative values of the indicator have been substituted with 0 . Our main concern has been not to award regions with higher male unemployment with respect to females as this goes against the concept of gender balance. Such approach is equivalent to assigning the same score to all those regions which lay further away from the optimal gender balance labor market which should be around the null value.

Similar treatment has been applied to the gender balance employment indicator. However, as negative values were not present, no changes were necessary.

The indicators measuring unemployment, long term unemployment, gender balance employment, gender balance unemployment and female unemployment are all reversed in order to have the same polarity with respect the level of competitiveness (the higher the better).

## Imputation of missing data

For the indicator on labor productivity, due to missing data, 2005 data has been used for UKN0.

## Univariate analysis

From the analysis of Table 51, we can observe very low percentage of missing values in the set of indicators describing the Labor market efficiency pillar. Six out of the nine indicators refer to data from 2008, while only job mobility and labor productivity are based on 2007 data. The indicator on Gender balance unemployment has a very high coefficient of variation (1.97) indicating a very heterogeneous situation among EU regions. Similarly, the indicators on long-term unemployment and labor productivity have somewhat higher coefficients of variation ( 0.97 and 0.71 , respectively) even though much lower than gender balance unemployment.

Table 51: Descriptive statistics of Labor market indicators

| Indicator | Employment rate (excluding agriculture) | Long-term unemployment | Unemployment | Job Mobility |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| description | $\%$ of population years | \% of labor force | \% of active population | \% of total employment (people who started to work for the current employer or as selfemployed in the last 2 years) |  |
| source | Eurostat Regional Employment, LFS | Eurostat Regional Employment, LFS | Eurostat Regional Employment, LFS | Eurostat Regional Employment, LFS |  |
| reference year | 2008 | 2008 | 2008 | 2007 |  |
| $\%$ of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 1.49 \\ 64.04 \\ 9.65 \\ 0.15 \end{gathered}$ | $\begin{aligned} & 0.00 \\ & 2.80 \\ & 2.70 \\ & 0.97 \end{aligned}$ | $\begin{aligned} & 0.00 \\ & 7.01 \\ & 3.74 \\ & 0.53 \end{aligned}$ | $\begin{gathered} 1.49 \\ 16.35 \\ 3.94 \\ 0.24 \end{gathered}$ |  |
| maximum value <br> region corresponding to maximum value <br> minimum value <br> region corresponding to minimum value | $\begin{array}{r} 80.20 \\ \text { DK01 } \\ 34.63 \\ \text { RO21 } \\ \hline \end{array}$ | $\begin{gathered} 19.37 \\ \text { FR91 } \\ 0.13 \\ \text { FI20 } \\ \hline \end{gathered}$ | $\begin{array}{r} 24.80 \\ \text { FR94 } \\ 1.90 \\ \text { CZ01 } \\ \hline \end{array}$ | $\begin{array}{r} 28.76 \\ \text { ES61 } \\ 6.86 \\ \text { GR25 } \\ \hline \end{array}$ |  |
| Indicator | Labor productivity | Gender balance uemployment | Gender balance employment | Female unemployment | Labor market policies |
| description | GDP/person employed in industry and services $(€)$, Index, EU27 = 100 | difference between female and male unemployment rates | difference between male and female employment rates | \% of female unemployed | \% of GDP spent on public expenditure on labour market policies |
| source | Eurostat Regional Employment, LFS | Eurostat, DG Regio | Eurostat, DG Regio | Eurostat Regional Employment, LFS | Eurostat Labor Market Policy Statistics |
| reference year | 2007 | 2008 | 2008 | 2008 | 2007 |
| $\%$ of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 0.00 \\ 93.94 \\ 25.26 \\ 0.27 \end{gathered}$ | $\begin{aligned} & 1.49 \\ & 1.49 \\ & 2.94 \\ & 1.97 \end{aligned}$ | $\begin{gathered} 0.00 \\ 13.70 \\ 6.56 \\ 0.48 \end{gathered}$ | $\begin{aligned} & 0.00 \\ & 7.88 \\ & 4.75 \\ & 0.60 \end{aligned}$ | $\begin{aligned} & 3.70 \\ & 1.27 \\ & 0.90 \\ & 0.71 \end{aligned}$ |
| maximum value <br> region corresponding to maximum value <br> minimum value <br> region corresponding to minimum value | $\begin{aligned} & 193.38 \\ & \text { NL11 } \\ & 28.36 \\ & \text { BG31 } \end{aligned}$ | $\begin{aligned} & 15.10 \\ & \text { ES63 } \\ & -3.50 \\ & \text { DE50 } \end{aligned}$ | $\begin{gathered} 40.50 \\ \text { GR41 } \\ 1.80 \\ \text { FI13 } \\ \hline \end{gathered}$ | $\begin{gathered} 29.60 \\ \text { FR93 } \\ 1.30 \\ \text { UKE2 } \end{gathered}$ | $\begin{gathered} 3.294 \\ \mathrm{BE} \\ 0.154 \\ \mathrm{EE} \\ \hline \end{gathered}$ |

## How do EU regions score in each of the indicators?

As shown in Figure 5-27, we can note that Eastern European regions perform consistently bad on the indicator related to employment rate. Similarly, Southern Italian regions also have among the lowest employment rates in Europe. This pattern is confirmed by the data on long-term and short-term unemployment where together with Southern Italian regions, some parts of Spain are also among the worst performers.
The interpretation of the indicator on job mobility could be controversial as higher mobility could both mean a dynamic labor market or a very volatile and insecure one. In this representation, we have related higher job mobility to better performance but any conclusions should be taken with caution. We see the southern regions of Spain having the highest level of job mobility together with parts of Sweden while some Greek regions and parts of Romania are among the regions with lowest mobility.

Female unemployment is clearly a significant problem in Southern European regions.
As regards labor productivity, we can see Eastern European regions are clearly showing the worst performance.

With regards to the indicator on gender balance unemployment, the highest unemployment difference among males and females can be observed in Southern European regions (parts of Spain, Portugal, Italy and Greece) while low difference, i.e. more gender balance labor market, is observed in parts of Romania and the UK as well as Ireland.

The indicator on gender balance employment shows similar results with the highest gender difference in Southern European regions and the lowest in Scandinavian regions.

Denmark and Belgium have the highest expenditure on labor market policies while Romania and Estonia the lowest.



Figure 5-27. Best and worst performing regions for each indicator - Labor market

Four out of the nine indicators analyzed have been transformed with the Box-Cox method due to asymmetric distribution - long-term unemployment, unemployment rate, gender balance employment and female unemployment, as shown in Table 52. Gender balance unemployment has been transformed logarithmically due to the presence of 0 values.

Table 52: Histograms of Labor market indicators
Employment rate


Long-term unemployment





Gender balance employment


Female unemployment


Trasformed indicator: Female unemployment Box-Cox transformation, lambda $=-0.05$ skewness $=0.09401$


Labor productivity


## Multivariate Analysis

From the analysis of the correlation matrix (Table 53) the indicators Job_mobility and LMP show a peculiar behavior.

Table 53: Correlation matrix between all the indicators included in the LME pillar

| ns |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Empl_rate | Long_term_ unempl_ reversed | Unemployment _reversed | Job_mobility | labor_ productivity | Gender_ balance_ unempl_ reversed | Gender_ balance_ empl_reversed | Female_ unemployment_ reversed | LMP_imputed |
| Empl_rate | Pearson Correlation <br> Sig. (1-tailed) <br> N |  | $\begin{array}{r} \hline .595^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} \hline .565^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} .113^{*} \\ .033 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} .537^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline .419^{\prime \prime} \\ .000 \\ 262 \\ \hline \end{array}$ | $\begin{array}{r} .396 " \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline .591^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r}.266 " \\ .000 \\ 251 \\ \hline\end{array}$ |
| Long_term_unempl_ reversed | Pearson Correlation <br> Sig. (1-tailed) <br> N | $\begin{array}{\|r\|} \hline .595^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $1$ | $\begin{array}{r\|} \hline .824^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline .145^{\prime \prime} \\ .009 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} .254^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline .342^{\prime} \\ .000 \\ 266 \\ \hline \end{array}$ | $\begin{array}{r} .157^{\prime \prime} \\ .005 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline .791^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r}. .024 \\ .354 \\ 255 \\ \hline\end{array}$ |
| Unemployment_ reversed | Pearson Correlation Sig. (1-tailed) <br> N | $\begin{array}{r} \hline .565^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} \hline .824^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | 1 268 | $\begin{array}{r\|} \hline .301 " \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} .239^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r} \hline .375^{\prime \prime} \\ .000 \\ 266 \\ \hline \end{array}$ | .074 <br> .115 <br> 268 | $\begin{array}{r\|} \hline .938^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r}-.054 \\ .197 \\ 255 \\ \hline\end{array}$ |
| Job_mobility | Pearson Correlation <br> Sig. (1-tailed) <br> N | $\begin{array}{r} .113^{*} \\ .033 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} .145 \\ .009 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} \hline .301 " \\ .000 \\ 264 \\ \hline \end{array}$ | 1 264 | $\begin{array}{r} -.132^{*} \\ .016 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r}.116 \\ .030 \\ 262 \\ \hline\end{array}$ | $\begin{array}{r} .280^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} -.199^{\prime \prime} \\ .001 \\ 264 \\ \hline \end{array}$ | .032 <br> .308 <br> 251 |
| labor_productivity | Pearson Correlation Sig. (1-tailed) N | $\begin{array}{r\|} \hline .537^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} .254 \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r} \hline .239 " \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r} \hline .132^{*} \\ .016 \\ 264 \\ \hline \end{array}$ |  | .025 .340 266 | $\begin{aligned} & .020 \\ & .374 \\ & 268 \\ & \hline \end{aligned}$ | $\begin{array}{r} .196 \\ .001 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r}.594 * \\ .000 \\ 255 \\ \hline\end{array}$ |
| Gender_balance_ unempl_reversed | Pearson Correlation <br> Sig. (1-tailed) <br> N | $\begin{array}{r} \hline .419^{\prime \prime} \\ .000 \\ 262 \\ \hline \end{array}$ | $\begin{array}{r} \hline .342^{\prime \prime} \\ .000 \\ 266 \\ \hline \end{array}$ | $\begin{array}{r} \hline .375^{\prime \prime} \\ .000 \\ 266 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline .116^{*} \\ .030 \\ 262 \\ \hline \end{array}$ | $\begin{aligned} & .025 \\ & .340 \\ & 266 \\ & \hline \end{aligned}$ | 266 | $\begin{array}{r} \hline .606^{\prime \prime} \\ .000 \\ 266 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline .627^{*} \\ .000 \\ 266 \\ \hline \end{array}$ | $\begin{array}{r}-.214 * \\ .000 \\ 253 \\ \hline\end{array}$ |
| Gender_balance_empl_reversed | Pearson Correlation <br> Sig. (1-tailed) <br> N | $\begin{array}{r} \hline .396^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{\|c\|} \hline .157^{*} \\ .005 \\ 268 \\ \hline \end{array}$ | $\begin{aligned} & .074 \\ & .115 \\ & 268 \\ & \hline \end{aligned}$ | $\begin{array}{r\|} \hline .280^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{aligned} & .020 \\ & .374 \\ & 268 \\ & \hline \end{aligned}$ | $\begin{array}{r\|} \hline .606{ }^{\prime \prime} \\ .000 \\ 266 \\ \hline \end{array}$ | 1 268 | $\begin{array}{r} \hline .255^{*} \\ .000 \\ 268 \\ \hline \end{array}$ | .097 <br> .061 <br> 255 |
| Female_unemployment_ reversed | Pearson Correlation Sig. (1-tailed) <br> N | $\begin{array}{r} .591^{\prime \prime} \\ .000 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} \hline .791 " \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r} \hline .938^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r\|} \hline-.199^{\prime \prime} \\ .001 \\ 264 \\ \hline \end{array}$ | $\begin{array}{r} \hline .196^{\prime \prime} \\ .001 \\ 268 \\ \hline \end{array}$ | $\begin{array}{r} .627^{\prime \prime} \\ .000 \\ 266 \\ \hline \end{array}$ | $\begin{array}{r} \hline .255^{\prime \prime} \\ .000 \\ 268 \\ \hline \end{array}$ | 1 268 | $\begin{array}{r\|} \hline-.163 * \\ .005 \\ 255 \\ \hline \end{array}$ |
| LMP_imputed | Pearson Correlation <br> Sig. (1-tailed) <br> N | $\begin{array}{r} .266^{\prime \prime} \\ .000 \\ 251 \\ \hline \end{array}$ | $\begin{array}{r} .024 \\ .354 \\ 255 \\ \hline \end{array}$ | $\begin{array}{r} -.054 \\ .197 \\ 255 \end{array}$ | $\begin{array}{r} .032 \\ .308 \\ 251 \\ \hline \end{array}$ | $\begin{array}{r} .594^{\prime \prime} \\ .000 \\ 255 \\ \hline \end{array}$ | $\begin{array}{r} \hline-214^{\prime \prime} \\ .000 \\ 253 \\ \hline \end{array}$ | $\begin{array}{r} .097 \\ .061 \\ 255 \\ \hline \end{array}$ | $\begin{array}{r} \hline .163^{\prime \prime} \\ .005 \\ 255 \\ \hline \end{array}$ | $\begin{array}{r}1 \\ 255 \\ \hline\end{array}$ |

${ }^{* *}$. Correlation is significant at the 0.01 level (1-tailed).
*. Correlation is significant at the 0.05 level ( 1 -tailed)
The former is significantly negatively correlated with Unemployment (reversed), Labor productivity and Female_unemployment (reversed), while it is not correlated with LMP (Sig. 1 -tailed higher than 0.01 ). Job_mobility is positively correlated only with four out of eight indicators included in the pillar: Empl_rate, Long_term_unemployment (reversed), Gender_balance_unemployment (reversed) and Gender_balance_employment (reversed). Indicator LMP is not correlated with Long_term_unemployment (reversed), Unemployment (reversed), Job_mobility and Gender_balance_employment (reversed), while it is significantly negatively correlated with Gender_balance_unemployment and Female unemployment (both reversed). In total, it is positively correlated only with two indicators, Empl_rate and Labor_productivity. The analysis of the correlation matrix suggests that Job_mobility and LMP are describing something else than the aspects the labor market pillar is intended to describe.

Regarding Job_mobility, the indicator is defined as the percentage share of people that in the reference year (2007 in this analysis) were working for the current employer a maximum of two years. The indicator is likely composed by two different aspects: one which actually reflects a dynamic and flexible workforce, thus being positively related to competitiveness; the other reflecting an insecure and unstable job market. It is then reasonable that it does not show a relation with unemployment or productivity measures.

As for LMP, some of the problems it shows may be due to the fact that the indicator is available at the country level only and regional values have been imputed according to the method described in Section 4.2.1.

For the reasons above, indicators Job_mobility and LMP have been excluded from the following PCA analysis.

PCA analysis (excluding Job mobility and LMP)
The PCA analysis on the subset of indicators shows the presence of a unique prevalent dimension which explains more than $53 \%$ of total variance (Figure 5-28 and Table 55). All the indicators contribute almost equally to the first dimension, with Labor_productivity and Gender_balance_empl slightly less relevant than the others - component loadings 0.39 and 0.42 respectively (Table 54).

Overall it can be said that the pillar including Empl_rate, Long_term_unempl, Unemployment, Labor_productivity, Gender_balance_unempl, Gender_balance_empl and Female_unempl is rather balanced and statistically consistent.

The distribution of labor market efficiency sub-score across regions is shown in Figure 5-29 and its histogram is due in Figure 5-30. Reordered regions are listed in Table 57.


Figure 5-28: PCA analysis of the labor market efficiency pillar - eigenvalues

Table 54: PCA analysis labor market efficiency pillar: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| Empl_rate | . 801 | -. 046 | . 424 | -. 198 | -. 372 | . 001 | -. 002 |
| Long_term_unempl_ reversed | . 846 | -. 265 | -. 205 | -. 250 | . 152 | -. 294 | -. 001 |
| Unemployment_reversed | . 872 | -. 304 | -. 308 | . 009 | . 040 | . 208 | . 084 |
| labor_productivity | . 387 | -. 419 | . 766 | . 226 | . 192 | . 001 | . 000 |
| Gender balance unempl_reversed | . 656 | . 615 | -. 060 | . 395 | -. 069 | -. 158 | . 031 |
| Gender_balance_empl_ reversed | . 422 | . 787 | . 273 | -. 271 | . 203 | . 114 | . 001 |
| Female_unemployment_ reversed | . 931 | -. 051 | -. 289 | . 139 | . 021 | . 131 | -. 098 |

Extraction Method: Principal Component Analysis.
a. 7 components extracted.

Table 55: PCA analysis for labor market efficiency pillar: explained variance
Total Variance Explained

| Component | Initial Eigenvalues |  |  | Extraction Sums of Squared Loadings |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% | Total | \% of Variance | Cumulative \% |
| 1 | 3.744 | 53.481 | 53.481 | 3.744 | 53.481 | 53.481 |
| 2 | 1.342 | 19.173 | 72.654 | 1.342 | 19.173 | 72.654 |
| 3 | 1.064 | 15.196 | 87.850 | 1.064 | 15.196 | 87.850 |
| 4 | .401 | 5.733 | 93.584 | .401 | 5.733 | 93.584 |
| 5 | .246 | 3.521 | 97.105 | .246 | 3.521 | 97.105 |
| 6 | .185 | 2.644 | 99.748 | .185 | 2.644 | 99.748 |
| 7 | .018 | .252 | 100.000 | .018 | .252 | 100.000 |

Extraction Method: Principal Component Analysis.


Figure 5-29: Map of Labor Market Efficiency sub-score.
Min-max normalized scores are due in Table 56.

Table 56: Labor market efficiency sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | -0.11 | 52 | Es30 | -0.07 | 53 | ${ }^{\text {AT33 }}$ | 1.24 | 87 |
| BE21 | 0.60 | 70 | ES41 | -0.80 | 34 | AT34 | 0.64 | 71 |
| BE22 | 0.26 | 61 | ES42 | -1.03 | 28 | PL11 | -0.57 | 40 |
| BE23 | 0.71 | 73 | ES43 | -1.56 | 14 | PL12 | 0.00 | 55 |
| BE25 | 0.83 | 76 | Es51 | 0.09 | 57 | PL21 | -0.51 | 41 |
| BE32 | -0.91 | 31 | Es52 | -0.63 | 38 | PL22 | -0.44 | 43 |
| ве33 | -0.67 | 37 | ES53 | -0.13 | 51 | PL31 | -0.59 | 39 |
| BE34 | -0.47 | 43 | ES61 | -1.39 | 19 | PL32 | -0.73 | 36 |
| BE35 | -0.64 | 38 | ES62 | -0.69 | 37 | PL33 | -0.66 | 38 |
| BG31 | -0.29 | 47 | ES63 | -1.83 | 7 | PL34 | -0.43 | 44 |
| BG32 | -0.61 | 39 | ES64 | -2.04 | 2 | PL41 | -0.74 | 36 |
| BG33 | -1.03 | 28 | Es70 | -1.09 | 27 | PL42 | -0.74 | 36 |
| BG34 | -0.36 | 45 | FR10 | 0.60 | 70 | PL43 | -0.53 | 41 |
| BG41 | 0.73 | 73 | FR21 | -0.37 | 45 | PL51 | -0.79 | 34 |
| BG42 | -0.14 | 51 | FR22 | -0.10 | 52 | PL52 | -0.56 | 40 |
| czo1 | 1.09 | 83 | FR23 | -0.44 | 43 | PL61 | -0.97 | 30 |
| CZ02 | 0.54 | 69 | FR24 | 0.13 | 58 | PL62 | -0.87 | 32 |
| czo3 | 0.23 | 61 | FR25 | -0.17 | 50 | PL63 | -0.34 | 46 |
| czo4 | -0.73 | 36 | FR26 | 0.29 | 62 | PT11 | -0.74 | 36 |
| cz05 | -0.16 | 51 | FR30 | -0.57 | 40 | PT15 | -0.49 | 42 |
| czo6 | -0.17 | 50 | FR41 | -0.26 | 48 | PT16 | -0.30 | 47 |
| czo7 | -0.30 | 47 | FR42 | 0.01 | 55 | PT17 | 0.07 | 56 |
| cz08 | -0.84 | 33 | FR43 | -0.41 | 44 | PT18 | -0.81 | 34 |
| DK01 | 1.34 | 89 | FR51 | 0.13 | 58 | PT20 | -0.37 | 45 |
| DK02 | 0.94 | 79 | FR52 | 0.40 | 65 | PT30 | 0.07 | 56 |
| DK03 | 1.07 | 82 | FR53 | 0.13 | 58 | RO11 | 0.30 | 62 |
| DK04 | 1.04 | 81 | FR61 | -0.13 | 51 | RO12 | -0.56 | 40 |
| DK05 | 0.90 | 78 | FR62 | 0.10 | 57 | RO21 | 0.00 | 55 |
| DE11 | 0.53 | 68 | FR63 | 0.27 | 62 | RO22 | -0.66 | 38 |
| DE12 | 0.47 | 67 | FR71 | 0.16 | 59 | RO31 | -0.60 | 39 |
| DE13 | 0.71 | 73 | FR72 | -0.09 | 52 | RO32 | 0.67 | 72 |
| DE14 | 0.61 | 70 | FR81 | -0.27 | 48 | RO41 | -0.41 | 44 |
| DE21 | 1.04 | 81 | FR82 | 0.00 | 55 | RO42 | -0.19 | 50 |
| DE22 | 0.34 | 63 | FR83 | -0.87 | 32 | S101 | -0.06 | 53 |
| DE23 | 0.64 | 71 | FR91 | -1.68 | 11 | S102 | 0.86 | 77 |
| DE24 | 0.14 | 58 | FR92 | -1.20 | 24 | SK01 | 0.89 | 78 |
| DE25 | 0.49 | 67 | FR93 | -2.12 | 0 | Sk02 | -0.60 | 39 |
| DE26 | 0.37 | 64 | FR94 | -1.67 | 12 | SK03 | -1.20 | 24 |
| DE27 | 0.46 | 66 | ITC1 | -0.01 | 54 | SK04 | -1.37 | 19 |
| DE30 | -0.31 | 47 | ITC2 | 0.51 | 68 | F13 | 0.49 | 67 |
| DE41 | -0.23 | 49 | ITC3 | -0.14 | 51 | F138 | 0.89 | 78 |
| DE42 | -0.01 | 54 | ITC4 | 0.37 | 64 | F19 | 0.43 | 66 |
| DE50 | 0.06 | 56 | ITD1 | 0.94 | 79 | F11A | 0.37 | 64 |
| DE60 | 0.46 | 66 | ITD2 | 0.47 | 67 | F120 | 1.76 | 100 |
| DE71 | 0.40 | 65 | ItD3 | 0.21 | 60 | SE11 | 1.23 | 86 |
| DE72 | 0.27 | 62 | ItD4 | -0.03 | 54 | SE12 | 0.33 | 63 |
| DE73 | 0.14 | 58 | ItD5 | 0.56 | 69 | SE21 | 0.84 | 76 |
| DE80 | -0.27 | 48 | ITE1 | -0.16 | 51 | SE22 | 0.44 | 66 |
| DE91 | -0.17 | 50 | ITE2 | -0.16 | 51 | SE23 | 0.96 | 79 |
| DE92 | 0.14 | 58 | וте3 | 0.04 | 56 | SE31 | 0.56 | 69 |
| DE93 | 0.16 | 59 | ITE4 | -0.54 | 41 | SE32 | 0.97 | 80 |
| DE94 | 0.06 | 56 | ITF1 | -0.61 | 39 | SE33 | 0.89 | 78 |
| deal | 0.11 | 57 | ITF2 | -1.13 | 26 | UKC1 | 0.06 | 56 |
| DEA2 | 0.10 | 57 | ITF3 | -1.63 | 13 | UKC2 | 0.36 | 64 |
| dea3 | 0.17 | 59 | ITF4 | -1.54 | 15 | UKD1 | 0.93 | 79 |
| DEA4 | -0.09 | 52 | ITF5 | -1.43 | 18 | UKD2 | 0.87 | 77 |
| deas | -0.19 | 50 | ITF6 | -1.49 | 16 | UKD3 | 0.27 | 62 |
| DEB1 | 0.03 | 55 | ITG1 | -1.64 | 12 | UKD4 | 0.44 | 66 |
| DEB2 | 0.56 | 69 | ITG2 | -1.36 | 20 | UKD5 | 0.09 | 57 |
| DEB3 | 0.36 | 64 | croo | 0.59 | 70 | UKE1 | 0.47 | 67 |
| deco | 0.10 | 57 | Lvoo | 0.11 | 57 | UKE2 | 1.51 | 94 |
| DED1 | -0.47 | 43 | Ltoo | 0.39 | 65 | UKE3 | 0.07 | 56 |
| DED2 | -0.06 | 53 | Lu00 | 0.41 | 65 | UKE4 | 0.31 | 63 |
| DED3 | -0.14 | 51 | HU10 | 0.10 | 57 | UKF1 | 0.54 | 69 |
| deeo | -0.61 | 39 | HU21 | -0.19 | 50 | UKF2 | 0.51 | 68 |
| defo | 0.20 | 60 | HU22 | -0.31 | 47 | UKF3 | -0.06 | 53 |
| DEGO | -0.50 | 42 | HU23 | -0.67 | 37 | UKG1 | 0.76 | 74 |
| EEOO | 0.37 | 64 | HU31 | -0.96 | 30 | UKG2 | 0.77 | 74 |
| $1 E 01$ | 0.11 | 57 | HU32 | -1.00 | 29 | UKG3 | -0.04 | 54 |
| $1 E 02$ | 0.67 | 72 | HU33 | -0.73 | 36 | UKH1 | 0.77 | 74 |
| GR11 | -1.34 | 20 | мтоо | -0.60 | 39 | UKH2 | 0.81 | 76 |
| GR12 | -1.20 | 24 | NL11 | 1.09 | 83 | Uкн3 | 0.53 | 68 |
| GR13 | -1.56 | 14 | NL12 | 1.11 | 83 | UKI | 0.56 | 69 |
| GR14 | -1.11 | 26 | NL13 | 0.76 | 74 | UKJ1 | 1.11 | 83 |
| GR21 | -1.46 | 17 | NL21 | 1.11 | 83 | UKJ2 | 1.00 | 80 |
| GR22 | -0.81 | 34 | NL22 | 1.11 | 83 | UKJ3 | 1.07 | 82 |
| GR23 | -1.49 | 16 | NL23 | 0.89 | 78 | UKJ4 | 0.41 | 65 |
| GR24 | -1.23 | 23 | NL31 | 1.61 | 96 | Uкк1 | 1.09 | 83 |
| GR25 | -0.99 | 29 | NL32 | 1.39 | 90 | UKK2 | 0.87 | 77 |
| GR30 | -0.36 | 45 | NL33 | 1.07 | 82 | บккз | 0.24 | 61 |
| GR41 | -0.89 | 32 | NL34 | 1.30 | 88 | UKK4 | 1.06 | 82 |
| GR42 | -0.74 | 36 | NL41 | 1.20 | 86 | UKL1 | 0.41 | 65 |
| GR43 | -0.69 | 37 | NL42 | 1.01 | 81 | UKL2 | 0.67 | 72 |
| ES11 | -0.50 | 42 | AT11 | 0.71 | 73 | UKM2 | 0.80 | 75 |
| ES12 | -0.57 | 40 | AT12 | 0.64 | 71 | Uкм3 | 0.60 | 70 |
| ES13 | -0.24 | 48 | AT13 | 0.46 | 66 | UKM5 | 1.40 | 91 |
| ES21 | -0.01 | 54 | AT21 | 0.63 | 71 | UKM6 | 0.95 | 79 |
| ES22 | -0.04 | 54 | AT22 | 0.90 | 78 | UKNO | 0.61 | 70 |
| ES23 | -0.31 | 47 | AT31 | 0.96 | 79 |  |  |  |
| ES24 | 0.00 | 55 | AT32 | 1.27 | 87 |  |  |  |

Figure 5-30: Histogram of Labor market efficiency sub-score


Table 57: Labor market efficiency pillar sub-rank (from best to worst)

|  <br>  <br>  |  |
| :---: | :---: |
|  |  <br>  |
|  |  <br>  <br>  |
|  |  |
|  |  <br>  <br>  |
|  |  <br>  |

### 5.8 Market size

Candidate indicators included in the pillar are discussed in Section 3.8 and recalled in the following.

Indicators included, in brackets short names:

1. GDP index EU27=100 (GDP_index)
2. Compensation of employees (Compensation_employees)
3. Disposable income (Disposable_income)
4. Potential GDP (Pot_market_size_GDP)
5. Potential population
(Pot_market_size_pop)

Disposable income is calculated as the regional net disposable income ( B 6 NU ) per head plus the difference between national net disposable income (S14_15_B6N) per head and national net adjusted disposable income (S14_15_B7N) per head multiplied by the total NUTS 2 regional population. Data for Romania is not adjusted while data for Luxembourg is estimated.

Due to the nature of the indicators on disposable income and compensation of employees, combining values for regions UKI00 and BE00, respectively in the UK and Belgium, as described in section 4.1, has been done through aggregation and not weighted average.

## Univariate analysis

Table 58 reports the descriptive statistics for the market size pillar indicators. We have no missing data for the first two indicators on GDP and Compensation of employees, low percentage of missing data on both potential market size expressed in GDP and population $(1.49 \%)$ and disposable income $(2.24 \%)$, all within the pre-defined threshold. This allows us to include all indicators in the construction of the Market size pillar. The indicators on compensation of employees, disposable income and potential GDP in pps have high coefficient of variations indicating a very heterogeneous situation among the different EU regions.

Table 58: Descriptive statistics of Market size indicators

| Indicator | GDP index | Compensation of employees | Disposable income | Potential GDP in PPS | Potential POP |
| :---: | :---: | :---: | :---: | :---: | :---: |
| description | Gross Domestic Product pps index, EU27=100 | Compensation of employees in millions of euros | Net adjusted disposable household income in millions of ppcs | Potential market size expressed in GDP (pps), index EU27=100 | Potential market size expressed in population, index EU27=100 |
| source | Eurostat Regional Economic Accounts | Eurostat Regional Economic Accounts | Eurostat, DG Regional Policy estimates | Eurostat, DG Regional Policy estimates | Eurostat, DG Regional Policy estimates |
| reference year | 2007 | 2006 | 2006 | 2007 | 2000 |
| $\%$ of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 0.00 \\ 95.61 \\ 34.14 \\ 0.36 \end{gathered}$ | $\begin{gathered} 0.00 \\ 21081.34 \\ 27248.80 \\ 1.29 \end{gathered}$ | $\begin{gathered} 2.24 \\ 31709.70 \\ 32598.85 \\ 1.03 \end{gathered}$ | $\begin{gathered} 1.49 \\ 193.20 \\ 216.27 \\ 1.12 \end{gathered}$ | $\begin{gathered} 1.49 \\ 174.78 \\ 156.43 \\ 0.90 \end{gathered}$ |
| maximum value region corresponding to maximum value minimum value region corresponding to minimum value | $\begin{aligned} & 275.23 \\ & \text { LU00 } \\ & 25.58 \\ & \text { BG31 } \\ & \hline \end{aligned}$ | $\begin{gathered} 271315.00 \\ \text { FR10 } \\ 560.40 \\ \text { FI20 } \\ \hline \end{gathered}$ | $\begin{gathered} 281068.70 \\ \text { FR10 } \\ 455.32 \\ \text { FI20 } \end{gathered}$ | $\begin{gathered} 1467.34 \\ \text { UKI } \\ 2.34 \\ \text { SE33 } \\ \hline \end{gathered}$ | $\begin{gathered} 895.04 \\ \text { UKI } \\ 3.00 \\ \text { SE33 } \\ \hline \end{gathered}$ |

## How do EU regions score in each of the indicators?

From Figure 5-31 we can see that Eastern European regions have the lowest performance in terms of the indicator on GDP index. Best performers are regions in some parts of Germany, Northern Europe and the UK. Similar situation can be noted for the indicators on compensation of employees and disposable income. With regards to the indicators on potential market size, we can see that peripheral regions have the lowest scores while regions in Belgium, the Netherlands, Germany and the UK have the highest scores.



Figure 5-31: Best and worst performing regions for each indicator - Market size
The next step in our analysis is the analysis of the distribution of the different indicators and possible transformation. Table 59 shows the initial distribution of each indicator. All four indicators have a clear positive skewness, typical of economic data (Zani, 2000). All but the GDP index have been transformed with the Box-Cox method as described in detail in Section 4.3.

Table 59: Histograms of Market size indicators


## Disposable income



Trasformed indicator. Disposable-income Box-Cox transformation, lambda= $=0.05$ skewness $=0.67142$


Potential market size GDP




Note: In the case of the Potential market size population indicator, the lambda used has been set to 0.15

## Multivariate analysis

The PCA analysis highlights the presence of one prevalent dimension equally described by all the indicators. In fact, the scree plot (Figure 5-32) and the percentage of explained variance (Table 62) show that the first PCA component accounts for more than $68 \%$ of total variance, well detached from the other ones. The table of component loadings (Table 61) indicates that all the indicators contribute almost evenly to the major PCA component. Given the analysis, we can conclude that this pillar has a unique, underlying dimension well captured by all the indicators.

The geographical distribution of the market size sub-score is shown in Figure 5-33 and its histogram is displayed in Figure 5-34. The reordered list of regions is due in Table 64.

Table 60: Correlation matrix between indicators included in the Market Size pillar

| Correlation Matrix |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | GDP_index | Compensation_ employees | Disposable_ income | Pot_market size_GDP | Pot_market_ size_pop |
| Correlation | GDP_index | 1.000 | . 586 | .368 | .461 | . 296 |
|  | Compensation_employees | . 586 | 1.000 | . 949 | . 634 | . 531 |
|  | Disposable_income | . 368 | . 949 | 1.000 | . 618 | . 561 |
|  | Pot_market_size_GDP | . 461 | . 634 | . 618 | 1.000 | . 960 |
|  | Pot_market_size_pop | . 296 | . 531 | . 561 | . 960 | 1.000 |
| Sig. (1-tailed) | GDP_index |  | . 000 | . 000 | . 000 | . 000 |
|  | Compensation_employees | . 000 |  | . 000 | . 000 | . 000 |
|  | Disposable_income | . 000 | . 000 |  | . 000 | . 000 |
|  | Pot_market_size_GDP | . 000 | . 000 | . 000 |  | . 000 |
|  | Pot_market_size_pop | . 000 | . 000 | . 000 | . 000 |  |



Figure 5-32: PCA analysis of the Market Size pillar - eigenvalues

Table 61: PCA analysis for the Market size pillar: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: | :---: |
|  | 1 | 2 |  | 3 | 4 |  |
| GDP_index | .619 | .529 | .579 | .032 | .017 |  |
| Compensation_ | .901 | .337 | -.255 | -.047 | -.087 |  |
| employees |  |  |  |  |  |  |
| Disposable_income | .864 | .183 | -.460 | .047 | .076 |  |
| Pot_market_size_GDP | .898 | -.381 | .189 | -.103 | .042 |  |
| Pot_market_size_pop | .826 | -.542 | .119 | .090 | -.042 |  |

Extraction Method: Principal Component Analysis.
a. 5 components extracted.

Table 62: PCA analysis for the Market size pillar: explained variance
Total Variance Explained

| Component | Initial Eigenvalues |  |  | Extraction Sums of Squared Loadings |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% | Total | \% of Variance | Cumulative \% |
| 1 | 3.431 | 68.623 | 68.623 | 3.431 | 68.623 | 68.623 |
| 2 | .866 | 17.311 | 85.933 | .866 | 17.311 | 85.933 |
| 3 | .662 | 13.238 | 99.172 | .662 | 13.238 | 99.172 |
| 4 | .024 | .483 | 99.654 | .024 | .483 | 99.654 |
| 5 | .017 | .346 | 100.000 | .017 | .346 | 100.000 |

Extraction Method: Principal Component Analys is.


Figure 5-33: Market size sub-score
(Min-max normalized values)

Table 63: Market size sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | 1.09 | 79 | ES30 | 1.03 | 78 | AT33 | -0.38 | 50 |
| BE21 | 0.80 | 74 | ES41 | -0.44 | 49 | AT34 | -0.64 | 45 |
| BE22 | 0.16 | 61 | ES42 | -0.60 | 46 | PL11 | -0.72 | 43 |
| BE23 | 0.43 | 66 | ES43 | -1.30 | 32 | PL12 | -0.06 | 56 |
| BE25 | 0.19 | 61 | ES51 | 0.57 | 69 | PL21 | -0.57 | 46 |
| BE32 | 0.06 | 59 | ES52 | 0.08 | 59 | PL22 | -0.10 | 56 |
| BE33 | -0.02 | 57 | ES53 | -0.83 | 41 | PL31 | -1.27 | 32 |
| BE34 | -1.02 | 37 | ES61 | -0.06 | 56 | PL32 | -1.21 | 34 |
| BE35 | -0.52 | 47 | ES62 | -0.64 | 45 | PL33 | -1.20 | 34 |
| BG31 | -2.16 | 15 | ES63 | -2.16 | 15 | PL34 | -1.66 | 25 |
| BG32 | -1.98 | 18 | ES64 | -2.89 | 0 | PL41 | -0.67 | 44 |
| BG33 | -2.08 | 16 | Es70 | -0.82 | 41 | PL42 | -1.26 | 33 |
| BG34 | -2.07 | 16 | FR10 | 1.86 | 95 | PL43 | -1.36 | 31 |
| BG41 | -1.17 | 34 | FR21 | -0.42 | 49 | PL51 | -0.62 | 45 |
| BG42 | -1.90 | 20 | FR22 | 0.17 | 61 | PL52 | -1.14 | 35 |
| czo1 | 0.24 | 62 | FR23 | 0.09 | 59 | PL61 | -1.01 | 38 |
| czo2 | -0.61 | 46 | FR24 | 0.00 | 58 | PL62 | -1.54 | 27 |
| czo3 | -0.87 | 40 | FR25 | -0.49 | 48 | PL63 | -1.01 | 38 |
| czo4 | -0.74 | 43 | FR26 | -0.43 | 49 | PT11 | -0.39 | 50 |
| czos | -0.70 | 44 | FR30 | 0.40 | 66 | PT15 | -1.64 | 25 |
| Cz06 | -0.65 | 45 | FR41 | -0.04 | 57 | PT16 | -0.65 | 45 |
| czo7 | -0.81 | 42 | FR42 | 0.21 | 62 | PT17 | 0.20 | 62 |
| czo8 | -0.62 | 45 | FR43 | -0.46 | 49 | PT18 | -1.22 | 33 |
| DK01 | 0.32 | 64 | FR51 | 0.03 | 58 | PT20 | -2.70 | 4 |
| DK02 | -0.51 | 48 | FR52 | -0.14 | 55 | Рт30 | -2.02 | 17 |
| Dк03 | -0.36 | 50 | fr53 | -0.45 | 49 | RO11 | -1.36 | 31 |
| DK04 | -0.35 | 51 | FR61 | -0.16 | 54 | RO12 | -1.32 | 31 |
| Dк05 | -0.88 | 40 | fr62 | -0.23 | 53 | RO21 | -1.43 | 29 |
| DE11 | 0.99 | 77 | FR63 | -0.96 | 39 | RO22 | -1.41 | 30 |
| DE12 | 0.84 | 74 | FR71 | 0.49 | 67 | RO31 | -1.02 | 37 |
| DE13 | 0.41 | 66 | FR72 | -0.56 | 47 | RO32 | -0.23 | 53 |
| DE14 | 0.43 | 66 | FR81 | -0.32 | 51 | RO41 | -1.50 | 28 |
| DE21 | 0.98 | 77 | FR82 | 0.24 | 62 | RO42 | -1.46 | 29 |
| DE22 | -0.06 | 56 | FR83 | -1.94 | 19 | S101 | -0.83 | 41 |
| DE23 | -0.07 | 56 | FR91 | -1.12 | 35 | S102 | -0.54 | 47 |
| DE24 | -0.05 | 57 | FR92 | -1.10 | 36 | SK01 | -0.17 | 54 |
| DE25 | 0.36 | 65 | FR93 | -2.04 | 17 | SK02 | -0.68 | 44 |
| DE26 | 0.18 | 61 | FR94 | -1.09 | 36 | SK03 | -1.07 | 36 |
| DE27 | 0.32 | 64 | ITC1 | 0.55 | 69 | SK04 | -1.22 | 33 |
| de30 | 0.54 | 68 | ITC2 | -1.14 | 35 | F113 | -1.70 | 24 |
| DE41 | -0.36 | 50 | ITC3 | -0.14 | 55 | F13 | -0.13 | 55 |
| DE42 | -0.12 | 55 | ITC4 | 1.21 | 82 | F19 | -0.96 | 39 |
| DE50 | 0.15 | 61 | ITD1 | -0.64 | 45 | F11A | -1.90 | 20 |
| DE60 | 0.96 | 77 | ITD2 | -0.43 | 49 | F120 | -2.80 | 2 |
| DE71 | 1.11 | 80 | ItD3 | 0.63 | 70 | SE11 | 0.44 | 66 |
| DE72 | 0.12 | 60 | ITD4 | -0.18 | 54 | SE12 | -0.44 | 49 |
| DE73 | 0.05 | 59 | ItD5 | 0.64 | 70 | SE21 | -1.01 | 38 |
| DE80 | -0.49 | 48 | ITE1 | 0.27 | 63 | SE22 | -0.29 | 52 |
| DE91 | 0.16 | 61 | ITE2 | -0.50 | 48 | SE23 | -0.31 | 51 |
| DE92 | 0.36 | 65 | Ite3 | -0.24 | 53 | SE31 | -1.25 | 33 |
| De93 | 0.02 | 58 | ITE4 | 0.67 | 71 | SE32 | -1.96 | 19 |
| DE94 | 0.21 | 62 | ITF1 | -0.47 | 48 | SE33 | -1.94 | 19 |
| deal | 1.32 | 84 | ITF2 | -1.10 | 36 | UKC1 | -0.25 | 53 |
| deA2 | 1.08 | 79 | ITF3 | 0.18 | 61 | UKC2 | -0.22 | 53 |
| dea3 | 0.65 | 71 | ITF4 | -0.27 | 52 | UKD1 | -0.74 | 43 |
| dea4 | 0.42 | 66 | ITF5 | -1.04 | 37 | UKD2 | 0.52 | 68 |
| deas | 0.85 | 75 | ITF6 | -0.76 | 43 | UKD3 | 0.79 | 73 |
| DEB1 | 0.26 | 63 | ITG1 | -0.23 | 53 | UKD4 | 0.29 | 63 |
| DEB2 | -0.43 | 49 | ITG2 | -0.92 | 39 | UKD5 | 0.24 | 62 |
| DEB3 | 0.48 | 67 | cyoo | -1.18 | 34 | UKE1 | -0.19 | 54 |
| DECO | 0.06 | 59 | Lvoo | -1.37 | 30 | UKE2 | -0.04 | 57 |
| ded1 | -0.12 | 55 | Ltoo | -0.98 | 38 | UKE3 | 0.33 | 64 |
| DED2 | -0.13 | 55 | Lu00 | 0.70 | 72 | UKE4 | 0.64 | 70 |
| DED3 | -0.21 | 53 | HU10 | 0.10 | 60 | UKF1 | 0.61 | 70 |
| deeo | -0.04 | 57 | HU21 | -0.88 | 40 | UKF2 | 0.56 | 69 |
| defo | 0.21 | 62 | HU22 | -1.03 | 37 | UKF3 | -0.36 | 50 |
| dego | -0.03 | 57 | HU23 | -1.49 | 28 | UKG1 | 0.28 | 63 |
| EE00 | -1.42 | 29 | HU31 | -1.16 | 35 | UKG2 | 0.32 | 64 |
| IE01 | -0.77 | 42 | HU32 | -1.23 | 33 | UKG3 | 0.69 | 71 |
| $1 E 02$ | 0.22 | 62 | HU33 | -1.27 | 32 | UKH1 | 0.34 | 64 |
| GR11 | -1.70 | 24 | мто0 | -1.58 | 26 | UKH2 | 0.94 | 76 |
| GR12 | -0.75 | 43 | NL11 | -0.16 | 54 | UKH3 | 0.63 | 70 |
| GR13 | -1.88 | 20 | NL12 | -0.40 | 50 | UKI | 2.12 | 100 |
| GR14 | -1.43 | 29 | NL13 | -0.42 | 49 | UKI1 | 1.16 | 81 |
| GR21 | -1.96 | 19 | NL21 | 0.18 | 61 | UK12 | 1.02 | 78 |
| GR22 | -2.47 |  | NL22 | 0.64 | 70 | UK13 | 0.61 | 70 |
| GR23 | -1.57 | 26 | NL23 | -0.26 | 52 | UKJ4 | 0.51 | 68 |
| GR24 | -1.06 | 37 | NL31 | 0.82 | 74 | Uкк1 | 0.63 | 70 |
| GR25 | -1.44 | 29 | NL32 | 0.90 | 76 | UKK2 | -0.04 | 57 |
| GR30 | 0.53 | 68 | NL33 | 1.05 | 79 | Uкк3 | -1.17 | 34 |
| GR41 | -2.86 | 1 | NL34 | -0.12 | 55 | UKK4 | -0.44 | 49 |
| GR42 | -2.01 | 18 | NL41 | 0.93 | 76 | UKL1 | -0.37 | 50 |
| GR43 | -1.66 | 25 | NL42 | 0.53 | 68 | UKL2 | -0.04 | 57 |
| ES11 | -0.39 | 50 | AT11 | -0.90 | 40 | UKM2 | 0.10 | 60 |
| ES12 | -0.74 | 43 | AT12 | -0.03 | 57 | UKM3 | 0.02 | 58 |
| ES13 | -0.90 | 40 | AT13 | 0.64 | 70 | Uкм5 | -0.70 | 44 |
| ES21 | 0.25 | 63 | AT21 | -0.75 | 43 | UKM6 | -1.58 | 26 |
| ES22 | -0.53 | 47 | AT22 | -0.35 | 51 | UKNO | -0.39 | 50 |
| ES23 | -1.15 | 35 | AT31 | -0.06 | 56 |  |  |  |
| ES24 | -0.68 | 44 | AT32 | -0.48 | 48 |  |  |  |

Figure 5-34: Histogram of Market size sub-score


Table 64: Market size pillar sub-rank (from best to worst)

| Market size |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | UKI | 46 | SE11 | 91 | FR24 | 136 | UKL1 | 181 | GR12 | 226 | PL31 |
| 2 | FR10 | 47 | BE23 | 92 | BE33 | 137 | AT33 | 182 | AT21 | 227 | ES43 |
| 3 | DEA1 | 48 | DE14 | 93 | DEGO | 138 | ES11 | 183 | ITF6 | 228 | RO12 |
| 4 | ITC4 | 49 | DEA4 | 94 | AT12 | 139 | PT11 | 184 | IE01 | 229 | PL43 |
| 5 | UKJ1 | 50 | DE13 | 95 | DEEO | 140 | UKNO | 185 | CZ07 | 230 | RO11 |
| 6 | DE71 | 51 | FR30 | 96 | FR41 | 141 | NL12 | 186 | ES70 | 231 | LV00 |
| 7 | BE00 | 52 | DE25 | 97 | UKE2 | 142 | FR21 | 187 | ES53 | 232 | RO22 |
| 8 | DEA2 | 53 | DE92 | 98 | UKK2 | 143 | NL13 | 188 | SIO1 | 233 | EE00 |
| 9 | NL33 | 54 | UKH1 | 99 | UKL2 | 144 | DEB2 | 189 | CZ03 | 234 | GR14 |
| 10 | ES30 | 55 | UKE3 | 100 | DE24 | 145 | FR26 | 190 | DK05 | 235 | RO21 |
| 11 | UKJ2 | 56 | DK01 | 101 | DE22 | 146 | ITD2 | 191 | HU21 | 236 | GR25 |
| 12 | DE11 | 57 | DE27 | 102 | ES61 | 147 | ES41 | 192 | ES13 | 237 | RO42 |
| 13 | DE21 | 58 | UKG2 | 103 | AT31 | 148 | SE12 | 193 | AT11 | 238 | HU23 |
| 14 | DE60 | 59 | UKD4 | 104 | PL12 | 149 | UKK4 | 194 | ITG2 | 239 | RO41 |
| 15 | UKH2 | 60 | UKG1 | 105 | DE23 | 150 | FR53 | 195 | FR63 | 240 | PL62 |
| 16 | NL41 | 61 | ITE1 | 106 | PL22 | 151 | FR43 | 196 | F119 | 241 | GR23 |
| 17 | NL32 | 62 | DEB1 | 107 | DE42 | 152 | ITF1 | 197 | LTOO | 242 | MT00 |
| 18 | DEA5 | 63 | ES21 | 108 | DED1 | 153 | AT32 | 198 | PL61 | 243 | UKM6 |
| 19 | DE12 | 64 | CZ01 | 109 | NL34 | 154 | DE80 | 199 | PL63 | 244 | PT15 |
| 20 | NL31 | 65 | FR82 | 110 | DED2 | 155 | FR25 | 200 | SE21 | 245 | GR43 |
| 21 | BE21 | 66 | UKD5 | 111 | FI18 | 156 | ITE2 | 201 | BE34 | 246 | PL34 |
| 22 | UKD3 | 67 | IE02 | 112 | FR52 | 157 | DK02 | 202 | RO31 | 247 | GR11 |
| 23 | LU00 | 68 | DE94 | 113 | ITC3 | 158 | BE35 | 203 | HU22 | 248 | FI13 |
| 24 | UKG3 | 69 | DEFO | 114 | FR61 | 159 | ES22 | 204 | ITF5 | 249 | GR13 |
| 25 | ITE4 | 70 | FR42 | 115 | NL11 | 160 | SIO2 | 205 | GR24 | 250 | BG42 |
| 26 | DEA3 | 71 | PT17 | 116 | SK01 | 161 | FR72 | 206 | SK03 | 251 | FI1A |
| 27 | ITD5 | 72 | BE25 | 117 | ITD4 | 162 | PL21 | 207 | FR94 | 252 | FR83 |
| 28 | NL22 | 73 | DE26 | 118 | UKE1 | 163 | ES42 | 208 | FR92 | 253 | SE33 |
| 29 | AT13 | 74 | ITF3 | 119 | DED3 | 164 | CZO2 | 209 | ITF2 | 254 | GR21 |
| 30 | UKE4 | 75 | NL21 | 120 | UKC2 | 165 | CZ08 | 210 | FR91 | 255 | SE32 |
| 31 | ITD3 | 76 | FR22 | 121 | FR62 | 166 | PL51 | 211 | ITC2 | 256 | BG32 |
| 32 | UKH3 | 77 | BE22 | 122 | ITG1 | 167 | ES62 | 212 | PL52 | 257 | GR42 |
| 33 | UKK1 | 78 | DE91 | 123 | RO32 | 168 | ITD1 | 213 | ES23 | 258 | PT30 |
| 34 | UKF1 | 79 | DE50 | 124 | ITE3 | 169 | AT34 | 214 | HU31 | 259 | FR93 |
| 35 | UKJ3 | 80 | DE72 | 125 | UKC1 | 170 | CZ06 | 215 | BG41 | 260 | BG34 |
| 36 | ES51 | 81 | HU10 | 126 | NL23 | 171 | PT16 | 216 | UKK3 | 261 | BG33 |
| 37 | UKF2 | 82 | UKM2 | 127 | ITF4 | 172 | PL41 | 217 | CYOO | 262 | BG31 |
| 38 | ITC1 | 83 | FR23 | 128 | SE22 | 173 | ES24 | 218 | PL33 | 263 | ES63 |
| 39 | DE30 | 84 | ES52 | 129 | SE23 | 174 | SK02 | 219 | PL32 | 264 | GR22 |
| 40 | GR30 | 85 | BE32 | 130 | FR81 | 175 | CZO5 | 220 | PT18 | 265 | PT20 |
| 41 | NL42 | 86 | DECO | 131 | DK04 | 176 | UKM5 | 221 | SK04 | 266 | F120 |
| 42 | UKD2 | 87 | DE73 | 132 | AT22 | 177 | PL11 | 222 | HU32 | 267 | GR41 |
| 43 | UKJ4 | 88 | FR51 | 133 | DK03 | 178 | CZ04 | 223 | SE31 | 268 | ES64 |
| 44 | FR71 | 89 | DE93 | 134 | DE41 | 179 | ES12 | 224 | PL42 |  |  |
| 45 | DEB3 | 90 | UKM3 | 135 | UKF3 | 180 | UKD1 | 225 | HU33 |  |  |

### 5.9 Technological readiness

As discussed in Section 3.9, the pillar has been divided into two sub-pillars, one describing Households and the other one Enterprises. In the following, the two sub-pillars are described separately.

## Sub-pillar Households

Candidate indicators for the sub-pillar are shown in Box 17 (Section 3.9). Here the list of indicators is briefly recalled together with their short names.

Indicators included in the sub-pillar HOUSEHOLDS, in brackets short names:

1. Share of households with access to broadband (Households-access-broadband)
2. Share of individuals who used internet to order goods/services (Individuals-buying-internet)
3. Share of households with internet access (Households-access-internet)

All indicators are positively associated to the concept of regional competitiveness in terms of technological use.

## Imputation of missing data

For the indicators on households broadband and internet access, NUTS 1 level data has been imputed at the NUTS 2 level for the following countries - Germany, Greece, France, Poland and Slovenia. Sweden NUTS 1 has been imputed for household broadband access.

For the indicator on household internet access, due to the lack of 2009 figures, 2008 data has been used for all regions in the Czech Republic and Romania, UKE2, UKG1, UKK4, UKL2, UKM6, and UKN0.

For the indicator on individuals buying over the internet, due to the lack of 2009 figures, 2008 data has been used for all region in the Czech Republic, UKE2, UKG1, UKK4, UKL2, UKM6, and UKN0.

## UNIVARIATE ANALYSIS

Table 65 shows the descriptive statistics for the three indicators included in the Household pillar. All three indicators have low percentage of missing values, close to $5 \%$.

Table 65: Descriptive statistics of Household indicators

| Indicator | Households access to <br> broadband | Individuals buying over <br> internet | Households access to internet |
| :--- | :---: | :---: | :---: |
| description | \% of total households <br> with access to <br> broadband | \% of individuals who <br> ordered goods or <br> services over the <br> internet for private use | \% of total households with <br> internet access |
| source | Eurostat Regional <br> Information Society <br> Statistics | Eurostat Regional <br> Information Society <br> Statistics | Eurostat Regional Information <br> Society Statistics |
| reference year | 2009 | 2009 | 2009 |
| \% of missing values | 4.48 | 4.48 | 4.10 |
| mean value | 55.10 | 36.72 | 62.82 |
| standard deviation (unbiased) | 15.29 | 21.65 | 17.32 |
| coefficient of variation | 0.28 | 0.59 | 0.28 |
| maximum value | 83.87 | 79.78 | 95.34 |
| region corresponding to maximum value | NL32 | UKM6 | NL32 |
| minimum value | 19.64 | 0.93 | 23.10 |
| region corresponding to minimum value | GR21 | RO41 | RO21 |

How do EU regions score in each of the indicators?
We can note from Figure 5-35 that Eastern European, Greek and some Italian regions, perform worst with regards to the access to internet, as well as broadband connection, to households. This is true also for the level of utilization of the internet for purchases by individuals. Northern European regions (parts of the Netherlands, UK and Denmark) show the highest performance in all three indicators.


Figure 5-35: Best and worst performing regions for each indicator - Household sub-pillar
Table 66 shows the frequency distribution of all indicators. No transformation has been performed because the indicators do not present highly asymmetric distribution as shown by the value of the skewness.

Table 66: Histograms of Household indicators




## MULTIVARIATE ANALYSIS

The PCA analysis highlights the presence of one clear prevalent dimension equally described by all the three indicators (see Table 67, Table 68, Table 69 and Figure 5-36).

The sub-score is computed as simple average of the three transformed and standardized indicators (Figure 5-37); sub-score values are shown in Table 70. Note that for five regions (DE50, FR91, FR92, FR93, FR94) the sub-score is missing due to missing values on all the three indicators.

Table 67: Correlation matrix between indicators included in the Technological readiness - Households sub-pillar

Correlation Matrix

|  |  | Household access_ broadband | Individual_ buying_internet | Household access_internet |
| :---: | :---: | :---: | :---: | :---: |
| Correlation | Household_access_ broadband | 1.000 | . 808 | . 844 |
|  | Individual_buying_internet | . 808 | 1.000 | . 899 |
|  | Household_access_internet | . 844 | . 899 | 1.000 |
| Sig. (1-tailed) | Household_access_ broadband |  | . 000 | . 000 |
|  | Individual_buying_internet | . 000 |  | . 000 |
|  | Household_access_internet | . 000 | . 000 |  |



Figure 5-36: PCA analysis of the Technological readiness
Households sub-pillar - eigenvalues

Table 68: PCA analysis for the Technological readiness - Households sub-pillar: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |
| :--- | :---: | :---: | :---: |
|  | 1 | 2 | 3 |
| Household_access_ <br> broadband_ | .930 | .363 | .053 |
| Individual_buying_interne <br> t | .952 | -.241 | .191 |
| Household_access_ <br> internet | .964 | -.113 | -.239 |

Extraction Method: Principal Component Analysis.
a. 3 components extracted.

Table 69: PCA analysis for the Technological readiness
Households sub-pillar: explained variance
Total Variance Explained

| Component | Initial Eigenvalues |  |  | Extraction Sums of Squared Loadings |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative $\%$ | Total | $\%$ of Variance | Cumulative $\%$ |
| 1 | 2.701 | 90.036 | 90.036 | 2.701 | 90.036 | 90.036 |
| 2 | .202 | 6.743 | 96.780 | .202 | 6.743 | 96.780 |
| 3 | .097 | 3.220 | 100.000 | .097 | 3.220 | 100.000 |

Extraction Method: Principal Component Analysis.


Figure 5-37: Map for sub-score of Technological readiness Households sub-pillar (Min-max normalized values)

Table 70: Technological readiness - Households sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | 0.38 | 64 | ES30 | 0.09 | 57 | AT33 | 0.22 | 60 |
| BE21 | 0.53 | 68 | ES41 | -0.97 | 30 | AT34 | 0.38 | 64 |
| BE22 | 0.46 | 66 | ES42 | -0.92 | 32 | PL11 | -0.38 | 45 |
| BE23 | 0.53 | 68 | ES43 | -1.16 | 26 | PL12 | -0.38 | 45 |
| BE25 | 0.46 | 66 | ES51 | -0.01 | 54 | PL21 | -0.42 | 44 |
| BE32 | -0.24 | 49 | ES52 | -0.75 | 36 | PL22 | -0.42 | 44 |
| ве33 | -0.33 | 46 | E553 | -0.05 | 53 | PL31 | -0.67 | 38 |
| BE34 | 0.18 | 59 | ES61 | -0.84 | 34 | PL32 | -0.67 | 38 |
| BE35 | -0.09 | 52 | ES62 | -0.97 | 30 | PL33 | -0.67 | 38 |
| BG31 | -2.10 | 2 | ES63 | -0.84 | 34 | PL34 | -0.67 | 38 |
| BG32 | -2.19 | 0 | ES64 | -0.59 | 40 | PL41 | -0.23 | 49 |
| BG33 | -2.02 | 4 | ES70 | -0.54 | 41 | PL42 | -0.23 | 49 |
| BG34 | -2.08 | 3 | FR10 | 0.52 | 67 | PL43 | -0.23 | 49 |
| BG41 | -1.37 | 20 | FR21 | -0.18 | 50 | PL51 | -0.39 | 45 |
| BG42 | -2.16 | 1 | FR22 | -0.18 | 50 | PL52 | -0.39 | 45 |
| cz01 | -0.15 | 51 | FR23 | -0.18 | 50 | PL61 | -0.35 | 46 |
| cz02 | -0.96 | 31 | FR24 | -0.18 | 50 | PL62 | -0.35 | 46 |
| czo3 | -1.13 | 26 | FR25 | -0.18 | 50 | PL63 | -0.35 | 46 |
| czo4 | -1.45 | 18 | FR26 | -0.18 | 50 | PT11 | -1.00 | 30 |
| cz05 | -1.13 | 26 | FR30 | -0.46 | 43 | PT15 | -0.72 | 37 |
| czo6 | -1.08 | 28 | FR41 | -0.10 | 52 | PT16 | -1.24 | 24 |
| CZ07 | -1.33 | 21 | FR42 | -0.10 | 52 | PT17 | -0.48 | 43 |
| cz08 | -1.19 | 25 | FR43 | -0.10 | 52 | PT18 | -1.33 | 21 |
| DK01 | 1.56 | 93 | FR51 | -0.31 | 47 | PT20 | -1.00 | 30 |
| DK02 | 1.05 | 81 | FR52 | -0.31 | 47 | PT30 | -0.88 | 33 |
| DK03 | 1.15 | 83 | FR53 | -0.31 | 47 | R011 | -0.93 | 31 |
| DK04 | 1.32 | 87 | FR61 | -0.17 | 50 | R012 | -1.07 | 28 |
| DK05 | 1.20 | 84 | FR62 | -0.17 | 50 | R021 | -1.22 | 24 |
| DE11 | 0.77 | 74 | FR63 | -0.17 | 50 | RO22 | -1.08 | 28 |
| DE12 | 0.77 | 74 | FR71 | 0.02 | 55 | R031 | -1.10 | 27 |
| DE13 | 0.77 | 74 | FR72 | 0.02 | 55 | RO32 | -0.77 | 35 |
| DE14 | 0.77 | 74 | FR81 | 0.22 | 60 | R041 | -1.33 | 21 |
| DE21 | 0.92 | 77 | FR82 | 0.22 | 60 | RO42 | -0.84 | 34 |
| DE22 | 0.92 | 77 | FR83 | 0.22 | 60 | S101 | -0.17 | 50 |
| DE23 | 0.92 | 77 | FR91 |  |  | 5102 | -0.17 | 50 |
| DE24 | 0.92 | 77 | FR92 |  |  | Sk01 | -0.45 | 43 |
| DE25 | 0.92 | 77 | FR93 |  |  | SK02 | -0.30 | 47 |
| DE26 | 0.92 | 77 | FR94 |  |  | Sk03 | -0.62 | 39 |
| DE27 | 0.92 | 77 | ITC1 | -1.02 | 29 | Sk04 | -0.66 | 38 |
| DE30 | 0.87 | 76 | ITC2 | -1.05 | 28 | F113 | 0.42 | 65 |
| DE41 | -0.30 | 47 | ITC3 | -1.06 | 28 | F118 | 1.22 | 85 |
| DE42 | 0.35 | 63 | ITC4 | -0.75 | 36 | F119 | 0.76 | 73 |
| DE50 |  |  | ITD1 | -0.75 | 36 | FI1A | 1.06 | 81 |
| DE60 | 1.00 | 79 | ITD2 | -0.75 | 36 | F120 |  |  |
| DE71 | 1.03 | 80 | ITD3 | -0.91 | 32 | SE11 | 1.73 | 98 |
| DE72 | 1.03 | 80 | ITD4 | -0.77 | 35 | SE12 | 1.51 | 92 |
| DE73 | 1.03 | 80 | ITD5 | -0.79 | 35 | SE21 | 1.32 | 87 |
| DE80 | 0.31 | 62 | ITE1 | -0.77 | 35 | SE22 | 1.32 | 87 |
| DE91 | 0.89 | 77 | ITE2 | -0.90 | 32 | SE23 | 1.43 | 90 |
| DE92 | 0.89 | 77 | ITE3 | -0.74 | 36 | SE31 | 1.29 | 87 |
| De93 | 0.89 | 77 | ITE4 | -0.71 | 37 | SE32 | 1.29 | 87 |
| DE94 | 0.89 | 77 | ITF1 | -0.97 | 30 | SE33 | 1.30 | 87 |
| DEA1 | 1.23 | 85 | ITF2 | -1.35 | 21 | UKC1 | 0.32 | 62 |
| DEA2 | 1.23 | 85 | ITF3 | -1.20 | 25 | UKC2 | 1.05 | 81 |
| DEA3 | 1.23 | 85 | ITF4 | -1.54 | 16 | UKD1 |  |  |
| DEA4 | 1.23 | 85 | ITF5 | -1.50 | 17 | UKD2 | 1.12 | 82 |
| DEA5 | 1.23 | 85 | ITF6 | -1.54 | 16 | UKD3 | 0.73 | 73 |
| Deb1 | 0.85 | 76 | ITG1 | -1.35 | 21 | UKD4 | 0.92 | 77 |
| DEB2 | 0.85 | 76 | ITG2 | -0.99 | 30 | UKD5 | 0.47 | 66 |
| deb3 | 0.85 | 76 | croo | -0.73 | 36 | UKE1 | 0.77 | 74 |
| deco | 0.94 | 78 | Lvoo | -0.42 | 44 | UKE2 | 0.81 | 75 |
| DED1 | -0.12 | 51 | Ltoo | -0.65 | 38 | UKE3 | 0.67 | 71 |
| DED2 | -0.12 | 51 | Luoo | 1.21 | 85 | UKE4 | 0.74 | 73 |
| DED3 | -0.12 | 51 | HU10 | -0.15 | 51 | UKF1 | 0.85 | 76 |
| DeE0 | 0.53 | 68 | HU21 | -0.37 | 45 | UKF2 | 1.16 | 83 |
| DEFO | 1.12 | 82 | HU22 | -0.57 | 40 | UKF3 |  |  |
| DEGO | 0.64 | 70 | HU23 | -0.99 | 30 | UKG1 | 1.37 | 89 |
| EEOO | -0.16 | 50 | HU31 | -0.96 | 31 | UKG2 | 1.36 | 88 |
| $1 E 01$ | -0.48 | 43 | HU32 | -1.08 | 28 | UKG3 | 0.70 | 72 |
| $1 E 02$ | 0.21 | 60 | Hu33 | -0.77 | 35 | UKH1 | 1.38 | 89 |
| GR11 | -1.80 | 10 | мтоо | 0.16 | 58 | UKH2 | 1.37 | 89 |
| GR12 | -1.80 | 10 | NL11 | 1.06 | 81 | Uкн3 | 1.21 | 85 |
| GR13 | -1.80 | 10 | NL12 | 1.10 | 82 | UKI | 1.38 | 89 |
| GR14 | -1.80 | 10 | NL13 | 1.57 | 94 | UKJ1 | 1.25 | 86 |
| GR21 | -2.13 | 1 | NL21 | 1.35 | 88 | UKJ2 | 1.62 | 95 |
| GR22 | -2.13 | 1 | NL22 | 1.41 | 90 | UKJ3 | 1.24 | 85 |
| GR23 | -2.13 | 1 | NL23 | 1.48 | 91 | UKJ4 | 1.33 | 88 |
| GR24 | -2.13 | 1 | NL31 | 1.83 | 100 | UKK1 | 1.40 | 89 |
| GR25 | -2.13 | 1 | NL32 | 1.76 | 98 | UKK2 | 1.24 | 85 |
| GR30 | -0.86 | 33 | NL33 | 1.47 | 91 | บккз |  |  |
| GR41 | -1.74 | 11 | NL34 | 1.08 | 81 | Uкк4 | 0.82 | 75 |
| GR42 | -1.74 | 11 | NL41 | 1.42 | 90 | UKL1 | 1.08 | 81 |
| GR43 | -1.74 | 11 | NL42 | 1.11 | 82 | UKL2 | 0.90 | 77 |
| ES11 | -1.12 | 27 | AT11 | 0.11 | 57 | UKM2 | 1.16 | 83 |
| ES12 | -0.51 | 42 | AT12 | 0.22 | 60 | Uкм 3 | 0.32 | 62 |
| ES13 | -0.32 | 47 | AT13 | 0.58 | 69 | Uкм5 |  |  |
| ES21 | -0.23 | 49 | AT21 | -0.11 | 52 | UKM6 | 1.60 | 94 |
| ES22 | -0.30 | 47 | AT22 | -0.19 | 50 | UKNO | -0.25 | 48 |
| ES23 | -0.70 | 37 | AT31 | 0.31 | 62 |  |  |  |
| ES24 | -0.53 | 41 | AT32 | 0.44 | 65 |  |  |  |

## Sub-pillar Enterprises

Indicators included in the pillar are discussed in Section 3.9. In the following we recall them and the short names they are assigned.

## Indicators included in the sub-pillar, in brackets short names:

1. Share of enterprises NOT using computers (reversed)
(Enterprises_no_computer_use)
2. Share of enterprises NOT having access to Internet (reversed) (Enterprises_no_internet_access)
3. Share of enterprises having a website or a webpage
(Enterprises_web)
4. Share of enterprises using Intranet (Enterprises_intranet)
5. Share of enterprises using an internal computer network
(Enterprises_internal_networks)
6. Share of employees using Extranet (Employees_extranet)
7. Share of employees NOT having access to Internet (reversed)
(Employees_no_internet_access)
The indicators Enterprises-no-computer-use, Enterprises-no-internet-access and Employees-no-internet-access have been reversed to have positive polarity with respect to competitiveness.

## Imputation of missing values

For Belgium, due to lack of 2009 data, 2008 values have been used for all indicators except enterprises_intranet where 2007 has been used.

As discussed in Section 3.9, the geographical coverage is not the same for all the indicators. Some of them are available at the NUTS2 level while others at the country level only. However, indicators available at the regional level (suffer from close to $50 \%$ of missing values. For this reason the sub-pillar has been treated at the country level only.

## UNIVARIATE ANALYSIS

Table 71 shows the descriptive statistics for the indicators used to describe the Enterprise sub-pillar. All indicators have no missing data. High coefficients of variation observed for the indicators on enterprise use of computers (1.04), enterprises internet access (0.90) and employees internet aces ( 0.91 ) show diverse situations across EU regions.

Table 71: Descriptive statistics of Enterprise indicators

| Name of indicator | Enterprises use of computers | Enterprises internet access | Enterprises use of websites | Enterprises use of intranet | Enterprises use of internal networks | Employees extranet access | Employees internet access |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| description of indicator | \% of enterprises NOT using computers | $\%$ of enterprises NOT having access to internet in the reference year | \% of enterprises having a website or a homepage | \% of enterprises using Intranet | use an internal computer network (e.g LAN) | $\%$ of persons employed by enterprise using Extranet | $\%$ of persons employed by enterprise NOT having access to the Internet |
| source | Community Survey on ICT usage and e -Commerce | Community Survey on ICT usage and e-Commerce | Community Survey on ICT usage and e-Commerce | Community Survey on ICT usage and e -Commerce | Community Survey on ICT usage and e-Commerce | Community Survey on ICT usage and e -Commerce | Community Survey on ICT usage and e-Commerce |
| reference year | 2009 | 2009 | 2009 | 2009 | 2009 | 2009 | 2009 |
| $\%$ of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 0 \\ 3.89 \\ 4.06 \\ 1.04 \end{gathered}$ | $\begin{gathered} 0 \\ 6.11 \\ 5.48 \\ 0.90 \end{gathered}$ | $\begin{gathered} 0 \\ 65.15 \\ 15.49 \\ 0.24 \end{gathered}$ | $\begin{gathered} 0 \\ 32.41 \\ 9.74 \\ 0.30 \end{gathered}$ | $\begin{gathered} 0 \\ 73.41 \\ 13.17 \\ 0.18 \end{gathered}$ | $\begin{gathered} 0 \\ 0.37 \\ 0.12 \\ 0.32 \end{gathered}$ | $\begin{gathered} 0 \\ 2.48 \\ 2.26 \\ 0.91 \end{gathered}$ |
| maximum value <br> region corresponding to maximum value <br> minimum value <br> region corresponding to minimum value | $\begin{gathered} 19.00 \\ \text { RO } \\ 0.00 \\ \text { NL } \end{gathered}$ | $\begin{gathered} 27.00 \\ \text { RO } \\ 1.21 \\ \text { FI } \\ \hline \end{gathered}$ | $\begin{gathered} 88.00 \\ \text { DK } \\ 28.00 \\ \text { RO } \\ \hline \end{gathered}$ | $\begin{gathered} 54.00 \\ \text { SK } \\ 18.00 \\ \text { CY } \end{gathered}$ | $\begin{gathered} 97.00 \\ \text { LU } \\ 44.00 \\ \text { RO } \end{gathered}$ | $\begin{aligned} & 0.56 \\ & \text { FR } \\ & 0.19 \\ & \text { LV } \\ & \hline \end{aligned}$ | $\begin{gathered} 12.00 \\ \text { RO } \\ 0.00 \\ \mathrm{FI} \end{gathered}$ |

## How do EU countries score in each of the indicators?

Scandinavian countries show very high penetration of ICT in their enterprises with Finland performing best in four out of the seven indicators. Sweden has the highest percentage of enterprises with a website together with Denmark, which also scores best on the indicator on enterprise internet access. Romania and Bulgaria show consistent low penetration of ICT technologies.

Enterprises_no_computer use


Enterprises_no_internet access



Figure 5-38. Best and worst performing regions for each indicator Enterprise sub-pillar
As shown in Table 72, three of the indicators have been transformed due to positive skewness. Enterprises_no_computer_use, enterprises_no_internet_access and Employees_no_internet_access have all been transformed logarithmically as described in Section 4.3, due to the presence of 0 values.

Table 72: Histograms of Enterprise indicators Enterprises no computer use



## Enterprises no internet access






Employees extranet


## Employees no internet access



## MULTIVARIATE ANALYSIS

The PCA analysis highlights the presence of one prevalent dimension which explains more than $70 \%$ of total variance (see Figure 5-39 and Table 75) and is well described by almost all the indicators. Indicators Enterprises_intranet is the only one not showing a high correlation with the others (Table 73) and, accordingly, it has a role in defining the second PCA dimension, which explains about $12 \%$ of total variance (Table 74 and Table 75).

On the basis of the analysis, all the indicators have been included in the computation of the final sub-score at the country level which is shown in Figure 5-40.

Table 73: Correlation matrix between indicators included in the Technological readiness Enterprises sub-pillar

Correlation Matrix

|  |  | Enterprises no_computer reversed | Enterprises no_internetreversed | Enterprises_ web | Enterprises intranet | Enterprises internal networks | Employees extranet | Employees no_internet_reversed |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Correlation | Enterprises_no computer_reversed | 1.000 | . 911 | . 736 | . 449 | . 728 | . 689 | . 801 |
|  | Enterprises_no_internet_ reversed | . 911 | 1.000 | . 781 | . 466 | . 705 | . 736 | . 896 |
|  | Enterprises_web | . 736 | . 781 | 1.000 | . 356 | . 741 | . 631 | . 673 |
|  | Enterprises_intranet | . 449 | . 466 | . 356 | 1.000 | . 257 | . 582 | . 338 |
|  | Enterprises_internal_ networks | . 728 | . 705 | . 741 | . 257 | 1.000 | . 689 | . 647 |
|  | Employees_extranet | . 689 | . 736 | . 631 | . 582 | . 689 | 1.000 | . 629 |
|  | Employees_no_internet_ reversed | . 801 | . 896 | . 673 | . 338 | . 647 | . 629 | 1.000 |
| Sig. (1-tailed) | Enterprises_no computer_rēversed |  | . 000 | . 000 | . 009 | . 000 | . 000 | . 000 |
|  | Enterprises_no_internet_ reversed | . 000 |  | . 000 | . 007 | . 000 | . 000 | . 000 |
|  | Enterprises_web | . 000 | . 000 |  | . 034 | . 000 | . 000 | . 000 |
|  | Enterprises_intranet | . 009 | . 007 | . 034 |  | . 098 | . 001 | . 042 |
|  | Enterprises_internal_ networks | . 000 | . 000 | . 000 | . 098 |  | . 000 | . 000 |
|  | Employees_extranet | . 000 | . 000 | . 000 | . 001 | . 000 |  | . 000 |
|  | Employees_no_internet_ reversed | . 000 | . 000 | . 000 | . 042 | . 000 | . 000 |  |

Scree Plot


Figure 5-39: PCA analysis of the Technological readiness
Enterprises sub-pillar - eigenvalues

Table 74: PCA analysis for the Technological readiness - Enterprises sub-pillar: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| Enterprises_no_ computer_rēversed | . 921 | -. 073 | -. 164 | . 012 | . 232 | -. 241 | . 088 |
| Enterprises_no_internet_ reversed | . 952 | -. 064 | -. 227 | -. 003 | -. 027 | -. 061 | -. 183 |
| Enterprises_web | . 851 | -. 175 | . 175 | . 421 | -. 188 | -. 016 | . 032 |
| Enterprises_intranet | . 549 | . 811 | -. 024 | . 133 | . 100 | . 109 | . 001 |
| Enterprises_internal_ networks | . 828 | -. 268 | . 396 | -. 083 | . 227 | . 161 | -. 028 |
| Employees_extranet | . 842 | . 250 | . 259 | -. 313 | -. 223 | -. 116 | . 025 |
| Employees_no_internet_ reversed | . 869 | -. 179 | -. 362 | -. 124 | -. 097 | . 226 | . 078 |

Extraction Method: Principal Component Analysis.
a. 7 components extracted.

Table 75: PCA analysis the Technological readiness -
Enterprises sub-pillar: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 4.930 | 70.429 | 70.429 |
| 2 | .865 | 12.360 | 82.788 |
| 3 | .465 | 6.638 | 89.426 |
| -4 | .315 | 4.507 | 93.933 |
| 5 | .211 | 3.008 | 96.940 |
| 6 | .164 | 2.348 | 99.288 |
| 7 | .050 | .712 | 100.000 |



Figure 5-40: Technological readiness - Enterprises sub-scores (Min-max normalized values)

Table 76: Enterprises sub-score as arithmetic mean of transformed and standardized indicators.

| country | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: |
| BE | 0.67 | 74 |
| BG | -1.47 | 23 |
| CZ | -0.29 | 51 |
| DK | 0.73 | 76 |
| DE | 0.86 | 79 |
| EE | -0.51 | 46 |
| IE | -0.04 | 57 |
| GR | -0.51 | 46 |
| ES | 0.03 | 59 |
| FR | 0.59 | 72 |
| IT | -0.36 | 50 |
| CY | -0.67 | 42 |
| LV | -1.16 | 31 |
| LT | -0.54 | 45 |
| LU | 0.81 | 77 |
| HU | -1.40 | 25 |
| MT | -0.04 | 57 |
| NL | 0.66 | 74 |
| AT | 0.60 | 73 |
| PL | -1.00 | 35 |
| PT | -0.60 | 44 |
| RO | -2.46 | 0 |
| SI | 0.13 | 61 |
| SK | 0.91 | 80 |
| FI | 1.76 | 100 |
| SE | 0.66 | 74 |
| UK | 0.06 | 60 |

The overall sub-score of Technological readiness is computed as simple arithmetic mean of the two sub-pillar scores. Since for the enterprise sub-pillar the sub-scores are available at the country level only, these values have been equally assigned to all the regions in that country. Sub-scores of the Technological readiness pillar are shown in Table 77 , while Figure 5-41 displays the sub-score histogram. The list of regions reordered form best to worst according to the overall technological readiness sub-score is due in Table 78.

Table 77: Overall technological readiness sub-score

| region | subscore | Min_max normalized score | region | subscore | Min_max normalized score | region | subscore | Min_max normalized score |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BE00 | 0.74 | 74 | ES30 | -0.04 | 54 | AT33 | 0.56 | 69 |
| BE21 | 0.72 | 74 | ES41 | -0.55 | 40 | AT34 | 0.56 | 69 |
| BE22 | 0.81 | 76 | ES42 | -0.56 | 40 | PL11 | -0.56 | 40 |
| BE23 | 0.80 | 76 | ES43 | -0.55 | 40 | PL12 | -0.56 | 40 |
| BE25 | 0.57 | 70 | ES51 | -0.13 | 51 | PL21 | -0.56 | 40 |
| BE32 | 0.33 | 63 | ES52 | -0.41 | 44 | PL22 | -0.56 | 40 |
| вез3 | 0.47 | 67 | Es53 | -0.18 | 50 | PL31 | -0.70 | 36 |
| BE34 | 0.52 | 68 | ES61 | -0.51 | 41 | PL32 | -0.70 | 36 |
| BE35 | 0.52 | 68 | ES62 | -0.55 | 40 | PL33 | -0.70 | 36 |
| BG31 | -1.94 | 3 | Es63 | -0.41 | 44 | PL34 | -0.70 | 36 |
| BG32 | -1.79 | 7 | ES64 | -0.36 | 45 | PL41 | -0.51 | 41 |
| BG33 | -1.72 | 9 | ES70 | -0.35 | 45 | PL42 | -0.51 | 41 |
| BG34 | -1.75 | 8 | FR10 | 0.66 | 72 | PL43 | -0.51 | 41 |
| BG41 | -1.32 | 20 | FR21 | 0.35 | 64 | PL51 | -0.54 | 40 |
| BG42 | -1.77 | 8 | FR22 | 0.35 | 64 | PL52 | -0.54 | 40 |
| CZ01 | 0.10 | 57 | FR23 | 0.35 | 64 | PL61 | -0.58 | 39 |
| cz02 | -0.27 | 47 | FR24 | 0.35 | 64 | PL62 | -0.58 | 39 |
| cz03 | -0.36 | 45 | FR25 | 0.35 | 64 | PL63 | -0.58 | 39 |
| cz04 | -0.50 | 41 | FR26 | 0.35 | 64 | PT11 | -0.80 | 33 |
| cz05 | -0.38 | 45 | FR30 | 0.37 | 64 | PT15 | -0.69 | 36 |
| cz06 | -0.35 | 45 | FR41 | 0.26 | 62 | PT16 | -0.91 | 31 |
| CZ07 | -0.47 | 42 | FR42 | 0.26 | 62 | PT17 | -0.50 | 42 |
| CZ08 | -0.39 | 44 | FR43 | 0.26 | 62 | PT18 | -0.88 | 31 |
| DK01 | 1.71 | 100 | FR51 | 0.40 | 65 | PT20 | -0.81 | 33 |
| DK02 | 1.32 | 90 | FR52 | 0.40 | 65 | PT30 | -0.76 | 34 |
| DK03 | 1.28 | 89 | FR53 | 0.40 | 65 | R011 | -1.96 | 3 |
| DK04 | 1.51 | 95 | FR61 | 0.23 | 61 | R012 | -1.99 | 2 |
| DK05 | 1.09 | 84 | FR62 | 0.23 | 61 | RO21 | -2.06 | 0 |
| DE11 | 0.58 | 70 | FR63 | 0.23 | 61 | RO22 | -1.90 | 4 |
| DE12 | 0.58 | 70 | FR71 | 0.33 | 63 | RO31 | -2.02 | 1 |
| DE13 | 0.58 | 70 | FR72 | 0.33 | 63 | RO32 | -1.59 | 12 |
| DE14 | 0.58 | 70 | FR81 | 0.46 | 67 | RO41 | -2.00 | 2 |
| DE21 | 0.67 | 72 | FR82 | 0.46 | 67 | RO42 | -1.97 | 3 |
| DE22 | 0.67 | 72 | FR83 | 0.46 | 67 | S101 | 0.00 | 55 |
| DE23 | 0.67 | 72 | FR91 | 0.51 | 68 | S102 | 0.00 | 55 |
| DE24 | 0.67 | 72 | FR92 | 0.51 | 68 | Sk01 | 0.31 | 63 |
| DE25 | 0.67 | 72 | FR93 | 0.51 | 68 | Sk02 | 0.27 | 62 |
| DE26 | 0.67 | 72 | FR94 | 0.51 | 68 | Sk03 | 0.22 | 60 |
| DE27 | 0.67 | 72 | ITC1 | -0.65 | 38 | Sk04 | 0.21 | 60 |
| DE30 | 0.64 | 72 | ITC2 | -0.62 | 38 | F113 | 1.21 | 87 |
| DE41 | 0.36 | 64 | ITC3 | -0.66 | 37 | F138 | 1.39 | 91 |
| DE42 | 0.36 | 64 | ITC4 | -0.45 | 43 | F19 | 1.39 | 92 |
| DE50 | 0.31 | 63 | ITD1 | -0.45 | 43 | F11A | 1.29 | 89 |
| DE60 | 0.56 | 69 | ITD2 | -0.50 | 42 | F120 | 1.17 | 86 |
| DE71 | 0.61 | 71 | ITD3 | -0.57 | 40 | SE11 | 1.13 | 85 |
| DE72 | 0.61 | 71 | ITD4 | -0.51 | 41 | SE12 | 1.13 | 85 |
| DE73 | 0.61 | 71 | ITD5 | -0.45 | 43 | SE21 | 1.18 | 86 |
| DE80 | 0.27 | 62 | ITE1 | -0.60 | 39 | SE22 | 1.18 | 86 |
| DE91 | 0.60 | 71 | ITE2 | -0.57 | 40 | SE23 | 1.18 | 86 |
| DE92 | 0.60 | 71 | ITE3 | -0.61 | 38 | SE31 | 0.97 | 80 |
| DE93 | 0.60 | 71 | ITE4 | -0.46 | 43 | SE32 | 0.97 | 80 |
| DE94 | 0.60 | 71 | ITF1 | -0.69 | 36 | SE33 | 0.97 | 80 |
| DEA1 | 0.67 | 72 | ITF2 | -0.75 | 35 | UKC1 | 0.24 | 61 |
| DEA2 | 0.67 | 72 | ITF3 | -0.75 | 35 | UKC2 | 0.16 | 59 |
| DEA3 | 0.67 | 72 | ITF4 | -0.83 | 33 | UKD1 | -0.32 | 46 |
| DEA4 | 0.67 | 72 | ITF5 | -0.75 | 35 | UKD2 | 0.46 | 67 |
| DEA5 | 0.67 | 72 | ITF6 | -0.86 | 32 | UKD3 | 0.18 | 59 |
| DEB1 | 0.64 | 71 | ITG1 | -0.86 | 32 | UKD4 | 0.11 | 58 |
| DEB2 | 0.64 | 71 | ITG2 | -0.63 | 38 | UKD5 | 0.23 | 61 |
| Deb3 | 0.64 | 71 | cyoo | -0.80 | 34 | UKE1 | 0.02 | 55 |
| DECO | 0.63 | 71 | Lvoo | -0.75 | 35 | UKE2 | 0.54 | 69 |
| DED1 | 0.23 | 61 | LTOO | -0.60 | 39 | UKE3 | 0.42 | 66 |
| DED2 | 0.23 | 61 | LU00 | 1.10 | 84 | UKE4 | 0.35 | 64 |
| DED3 | 0.23 | 61 | HU10 | -0.79 | 34 | UKF1 | 0.44 | 66 |
| deeo | 0.26 | 61 | HU21 | -0.92 | 30 | UKF2 | 0.60 | 70 |
| DEFO | 0.53 | 69 | HU22 | -1.02 | 28 | UKF3 | 0.55 | 69 |
| DEGO | 0.42 | 66 | HU23 | -1.17 | 24 | UKG1 | 0.80 | 76 |
| eeoo | -0.09 | 52 | HU31 | -1.16 | 24 | UKG2 | 0.20 | 60 |
| IE01 | 0.23 | 61 | HU32 | -1.19 | 23 | UKG3 | 0.31 | 63 |
| 1 E02 | 0.23 | 61 | hu33 | -1.09 | 26 | UKH1 | 0.63 | 71 |
| GR11 | -1.21 | 23 | мто0 | 0.09 | 57 | UKH2 | 0.56 | 69 |
| GR12 | -1.21 | 23 | NL11 | 1.55 | 96 | UKH3 | 0.59 | 70 |
| GR13 | -1.21 | 23 | NL12 | 1.42 | 92 | UKI | 0.56 | 69 |
| GR14 | -1.21 | 23 | NL13 | 1.21 | 87 | UKJ1 | 0.81 | 76 |
| GR21 | -1.29 | 20 | NL21 | 1.54 | 95 | UKJ2 | 0.76 | 75 |
| GR22 | -1.29 | 20 | NL22 | 1.49 | 94 | UKJ3 | 0.54 | 69 |
| GR23 | -1.29 | 20 | NL23 | 1.43 | 92 | UKJ4 | 0.60 | 70 |
| GR24 | -1.29 | 20 | NL31 | 1.59 | 97 | UKK1 | 0.63 | 71 |
| GR25 | -1.29 | 20 | NL32 | 1.61 | 97 | UKK2 | 0.51 | 68 |
| GR30 | -0.75 | 35 | NL33 | 1.51 | 95 | UKK3 | 0.58 | 70 |
| GR41 | -1.18 | 23 | NL34 | 1.34 | 90 | UKK4 | 0.54 | 69 |
| GR42 | -1.18 | 23 | NL41 | 1.37 | 91 | UKL1 | 0.38 | 65 |
| GR43 | -1.18 | 23 | NL42 | 1.34 | 90 | UKL2 | 0.44 | 66 |
| ES11 | -0.61 | 38 | AT11 | 0.55 | 69 | UKM2 | 0.37 | 64 |
| ES12 | -0.29 | 47 | AT12 | 0.50 | 68 | UKM3 | 0.26 | 61 |
| ES13 | -0.25 | 48 | AT13 | 0.75 | 75 | UKM5 | 0.02 | 55 |
| ES21 | -0.25 | 48 | AT21 | 0.39 | 65 | UKM6 | 0.91 | 79 |
| ES22 | -0.27 | 48 | AT22 | 0.51 | 68 | UKNO | 0.04 | 56 |
| ES23 | -0.32 | 46 | AT31 | 0.55 | 69 |  |  |  |
| ES24 | -0.30 | 47 | AT32 | 0.56 | 69 |  |  |  |

Figure 5-41: Histogram of Technological readiness sub-score


Table 78: Technological readiness pillar sub-rank (from best to worst)

| Technological readiness |  |  |
| :---: | ---: | ---: |
| 1 | FI | Finland |
| 2 | SK | Slovakia |
| 3 | DE | Germany |
| 4 | LU | Luxembourg |
| 5 | DK | Denmark |
| 6 | BE | Belgium |
| 7 | NL | Netherlands |
| 8 | SE | Sweden |
| 9 | AT | Austria |
| 10 | FR | France |
| 11 | SI | Slovenia |
| 12 | UK | United Kingdom |
| 13 | ES | Spain |
| 14 | IE | Ireland |
| 15 | MT | Malta |
| 16 | CZ | Czech republic |
| 17 | IT | Italy |
| 18 | EE | Estonia |
| 19 | GR | Greece |
| 20 | LT | Lithuania |
| 21 | PT | Portugal |
| 22 | CY | Cyprus |
| 23 | PL | Poland |
| 24 | LV | Latvia |
| 25 | HU | Hungary |
| 26 | BG | Bulgaria |
| 27 | RO | Romania |

### 5.10 Business sophistication

All the indicators included in the pillar are available at the regional NUTS2 level (Section 3.10). In the following, they are recalled with their short names used in the statistical analysis. Indicators included in the pillar, in brackets short names:

1. Share of employment in 'sophisticated' sectors (Employment_JK)
2. Share of GVA in ‘sophisticated’ sectors (GVA_JK)
3. New foreign firms per (mill.) inhabitants (FDI_intensity)
4. Strength of regional clusters (Regional_clusters)

As discussed in Section 3.10, three indicators on venture capital have been considered but have resulted having more than $35 \%$ of missing values and have been thus, discarded from the analysis.

## Univariate analysis

As can be seen from Table 79, the indicators included in the pillar have a very low percentage of missing values $(0.37 \%$ for Employment JK, $0 \%$ for GVA, $0.75 \%$ for new foreign firms, and $4.1 \%$ for strength of regional clusters). Thus, all indicators have been included in the analysis.

The coefficient of variation is quite high for the indicator on new foreign firms (2.84) suggesting very diverse situations among EU regions.

Table 79: Descriptive statistics of Business sophistication indicators

| Indicator | Employment, JK sector | GVA, JK sector | FDI intensity | Stength of regional clusters |
| :---: | :---: | :---: | :---: | :---: |
| description | employment in the <br> "Financial <br> intermediation, real estate, renting and business activities" sector (J_K) as \% of total employment | GVA in the "Financial intermediation, real estate, renting and business activities" sector (J_K) as \% of total GVA | Number of new foreign firms per million inhabitants | for description of the derivation of the indicator, see Appendix B |
| source | Eurostat Regional Statistics | Eurostat Regional Statistics | ISLA - Bocconi | European Cluster Observatory |
| reference year | 2007 | 2007 | 2005-2007 | 2006 |
| \% of missing values mean value standard deviation (unbiased) coefficient of variation | $\begin{gathered} 0.37 \\ 12.38 \\ 5.41 \\ 0.44 \end{gathered}$ | $\begin{gathered} 0.00 \\ 23.36 \\ 6.59 \\ 0.28 \end{gathered}$ | $\begin{gathered} 0.75 \\ 173.43 \\ 493.33 \\ 2.84 \end{gathered}$ | $\begin{gathered} 4.10 \\ 14.39 \\ 8.50 \\ 0.59 \end{gathered}$ |
| maximum value region corresponding to maximum value minimum value region corresponding to minimum value | $\begin{array}{r} 29.05 \\ \text { NL31 } \\ 2.53 \\ \text { BG31 } \end{array}$ | $\begin{array}{r} 48.63 \\ \text { LU00 } \\ 9.59 \\ \text { CZ04 } \end{array}$ | $\begin{gathered} 6813.10 \\ \text { RO32 } \\ 0.00 \\ \text { GR22 } \end{gathered}$ | $\begin{gathered} 52.00 \\ \text { ITC4 } \\ 2.00 \\ \text { ITF6 } \end{gathered}$ |

## How do EU regions score in each of the indicators?

As we can see from Figure 5-42, employment in 'sophisticated sectors' is lowest in Eastern European regions and parts of Greece and Portugal. Similar situation is seen for the indicator on GVA with some UK and Central European regions also among the worst performers. The indicator on new foreign firms shows high FDI intensity in a number of Romanian and UK regions which are among the best performers. Worst performance in terms of FDI can be seen in Southern European regions, Italy and Greece, in particular. The indicator on the strength of regional clusters shows a very diverse situation across regions. Northern Italian regions show very strong regional clusters activity, being among the best performers together with parts of Southern Germany, Belgium, Denmark, and Spain.


Figure 5-42: Best and worst performing regions for each indicator - Business sophistication

Table 80 shows the histograms of the four indicators. Two indicators have been transformed due to positive skewness. The indicator on new foreign firms has been transformed logarithmically due to the presence of zero values while the indicator on regional clusters has been transformed with the Box-Cox method.

Table 80: Histograms of Business sophistication indicators
Employment_JK



## Multivariate analysis

The correlation matrix (Table 81) shows a discrete correlation pattern between indicators with the highest correlation between Employment and GVA in J-K sectors. The PCA analysis highlights the presence of a first prevalent dimension (Figure 5-43) that accounts for about $56 \%$ of total variation (Table 83). As expected from the correlation coefficients, the first two indicators mostly contribute to the first dimension with component loadings of about 0.9 , with the loadings of the remaining indicators below 0.59 (Table 82). The second dimension, which explains about $20 \%$ of variance (Table 83), is mainly due to the indicator

FDI_intensity which has a correlation of 0.70 with this second component (Table 82). Overall, PCA outcomes support the hypothesis of a single major dimension underlying the pillar. Figure 5-44 shows the geographical distribution of the business sophistication subscore computed as arithmetic mean of all four indicators. The histogram of the sub-score is displayed in Figure 5-45 while reordered regions are listed in Table 85.

Table 81: Correlation matrix between indicators included in the Business sophistication pillar Correlation Matrix

|  |  | Employment_ <br> JK | GVA_JK | FDI_intensity | Regional_ <br> clusters |
| :--- | :--- | ---: | ---: | ---: | ---: |
| Correlation | Employment_JK | 1.000 | .834 | .346 | .393 |
|  | GVA_JK | .834 | 1.000 | .277 | .291 |
|  | FDI_intensity | .346 | .277 | 1.000 | .226 |
|  | Regional_clusters | .393 | .291 | .226 | 1.000 |
| Sig. (1-tailed) | Employment_JK |  | .000 | .000 | .000 |
|  | GVA_JK | .000 |  | .000 | .000 |
|  | FDI_intensity | .000 | .000 |  | .000 |
|  | Regional_clusters | .000 | .000 | .000 |  |



Figure 5-43: PCA analysis for the Business sophistication pillar - eigenvalues

Table 82: PCA analysis for the Business sophistication pillar: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |
| :--- | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 |
| Employment_JK | .917 | -.265 | -.067 | -.290 |
| GVA_JK | .870 | -.390 | -.140 | .267 |
| FDI_intensity | .553 | .699 | -.453 | .021 |
| Regional_clusters | .590 | .333 | .735 | .037 |

Extraction Method: Principal Component Analysis.
a. 4 components extracted.

Table 83: PCA analysis for the Business sophistication pillar: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 2.252 | 56.295 | 56.295 |
| 2 | .822 | 20.553 | 76.848 |
| -769 | 19.232 | 96.080 |  |
| 3 | .157 | 3.920 | 100.000 |



Figure 5-44: Business sophistication sub-score.
Values of the min-max normalized sub-scores are shown in Table 84

Table 84: Business sophistication sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | 1.37 | 86 | ES30 | 0.78 | 70 | AT33 | -0.24 | 43 |
| BE21 | 0.98 | 76 | ES41 | -0.76 | 29 | AT34 | 0.03 | 50 |
| BE22 | 0.58 | 65 | ES42 | -1.42 | 11 | PL11 | -0.43 | 38 |
| BE23 | 0.64 | 67 | ES43 | -1.52 | 8 | PL12 | 0.75 | 70 |
| BE25 | 0.50 | 63 | Es51 | 0.44 | 61 | PL21 | -0.35 | 40 |
| BE32 | 0.02 | 50 | ES52 | -0.22 | 43 | PL22 | -0.36 | 40 |
| BE33 | 0.08 | 51 | ES53 | -0.81 | 27 | PL31 | -0.85 | 26 |
| BE34 | -0.22 | 43 | ES61 | -0.42 | 38 | PL32 | -0.87 | 26 |
| BE35 | -0.13 | 46 | Es62 | -1.05 | 21 | PL33 | -1.07 | 20 |
| BG31 | -1.34 | 13 | ES63 | -1.44 | 11 | PL34 | -0.97 | 23 |
| BG32 | -1.06 | 21 | Es64 | -1.60 | 6 | PL41 | -0.40 | 39 |
| BG33 | -1.08 | 20 | ES70 | -1.06 | 21 | PL42 | -0.52 | 35 |
| BG34 | -1.24 | 16 | FR10 | 1.88 | 100 | PL43 | -0.81 | 27 |
| BG41 | 0.31 | 58 | FR21 | -0.45 | 37 | PL51 | -0.25 | 43 |
| BG42 | -1.02 | 22 | FR22 | -0.11 | 46 | PL52 | -0.59 | 33 |
| CZ01 | 0.87 | 73 | fR23 | 0.05 | 51 | PL61 | -0.71 | 30 |
| czo2 | -0.86 | 26 | FR24 | -0.13 | 46 | PL62 | -0.78 | 28 |
| czo3 | -0.68 | 31 | FR25 | -0.38 | 39 | PL63 | -0.37 | 39 |
| cz04 | -0.86 | 26 | FR26 | -0.38 | 39 | PT11 | -0.66 | 32 |
| czos | -0.73 | 30 | FR30 | 0.26 | 56 | PT15 | -1.35 | 13 |
| czo6 | -0.49 | 36 | FR41 | -0.09 | 47 | PT16 | -0.96 | 23 |
| cz07 | -0.91 | 25 | FR42 | 0.15 | 53 | PT17 | 0.25 | 56 |
| Cz08 | -1.13 | 19 | FR43 | -0.06 | 48 | PT18 | -1.32 | 14 |
| DK01 | 1.22 | 82 | FR51 | 0.02 | 50 | PT20 | -1.83 | 0 |
| DK02 | 0.10 | 52 | FR52 | -0.37 | 39 | Рт30 | -1.34 | 13 |
| DK03 | 0.09 | 52 | fr53 | -0.27 | 42 | RO11 | -0.49 | 36 |
| DK04 | 0.14 | 53 | FR61 | -0.13 | 46 | RO12 | -0.19 | 44 |
| DK05 | 0.01 | 50 | FR62 | 0.06 | 51 | RO21 | -0.74 | 29 |
| DE11 | 0.41 | 60 | FR63 | -0.63 | 32 | RO22 | -0.58 | 34 |
| DE12 | 0.47 | 62 | FR71 | 0.77 | 70 | RO31 | -0.37 | 39 |
| DE13 | -0.21 | 44 | FR72 | -0.33 | 40 | RO32 | 0.51 | 63 |
| DE14 | -0.08 | 47 | FR81 | -0.25 | 43 | RO41 | -0.56 | 34 |
| DE21 | 1.04 | 77 | FR82 | 0.27 | 57 | RO42 | -0.14 | 46 |
| DE22 | -0.38 | 39 | FR83 | -0.84 | 27 | 5101 | -0.77 | 29 |
| DE23 | -0.03 | 49 | FR91 | -0.34 | 40 | 5102 | -0.13 | 46 |
| DE24 | -0.17 | 45 | FR92 | -0.43 | 38 | SK01 | 0.53 | 64 |
| DE25 | 0.40 | 60 | FR93 | -0.41 | 38 | SK02 | -0.86 | 26 |
| DE26 | -0.03 | 49 | FR94 | -0.62 | 33 | Sk03 | -1.27 | 15 |
| DE27 | -0.05 | 48 | ITC1 | 0.44 | 61 | SK04 | -1.22 | 16 |
| DE30 | 0.77 | 70 | ITC2 | -0.54 | 35 | F113 | -1.17 | 18 |
| DE41 | -0.33 | 40 | ITC3 | -0.30 | 41 | F118 | 0.25 | 56 |
| DE42 | -0.11 | 46 | ITC4 | 0.87 | 73 | F19 | -0.75 | 29 |
| DE50 | 0.28 | 57 | ITD1 | -0.56 | 34 | F11A | -0.92 | 25 |
| DE60 | 1.15 | 80 | ITD2 | -0.47 | 37 | F120 | -0.72 | 30 |
| DE71 | 1.42 | 88 | ITD3 | 0.21 | 55 | SE11 | 1.20 | 82 |
| DE72 | -0.16 | 45 | ItD4 | -0.15 | 45 | SE12 | -0.21 | 44 |
| De73 | -0.15 | 45 | ItD5 | 0.34 | 58 | SE21 | -0.68 | 31 |
| DE80 | -0.37 | 39 | ITE1 | 0.11 | 52 | SE22 | -0.33 | 40 |
| DE91 | -0.10 | 47 | ITE2 | -0.62 | 33 | SE23 | -0.21 | 44 |
| DE92 | 0.14 | 53 | ITE3 | -0.46 | 37 | SE31 | -0.84 | 27 |
| De93 | -0.52 | 35 | ITE4 | 0.45 | 61 | SE32 | -0.82 | 27 |
| De94 | -0.29 | 42 | ITF1 | -0.98 | 23 | SE33 | -1.12 | 19 |
| deal | 0.81 | 71 | ITF2 | -0.93 | 24 | UKC1 | -0.47 | 37 |
| deA2 | 0.75 | 70 | ITF3 | -0.48 | 36 | UKC2 | 0.02 | 50 |
| DEA3 | -0.06 | 48 | ITF4 | -0.96 | 23 | UKD1 | -1.28 | 15 |
| DEA4 | -0.08 | 47 | ITF5 | -0.98 | 23 | UKD2 | 0.53 | 64 |
| deas | -0.05 | 48 | ITF6 | -1.57 | 7 | UKD3 | 0.56 | 64 |
| DEB1 | -0.27 | 42 | ITG1 | -1.02 | 22 | UKD4 | -0.22 | 43 |
| DEB2 | -0.43 | 38 | ITG2 | -1.09 | 20 | UKD5 | 0.08 | 51 |
| DEB3 | -0.09 | 47 | cyoo | -0.47 | 37 | UKE1 | -0.60 | 33 |
| DECO | 0.02 | 50 | Lvoo | -0.38 | 39 | UKE2 | 0.06 | 51 |
| DED1 | -0.16 | 45 | Ltoo | -0.87 | 26 | UKE3 | -0.02 | 49 |
| DED2 | -0.13 | 46 | Luoo | 1.36 | 86 | UKE4 | 0.42 | 61 |
| ded3 | 0.04 | 50 | HU10 | 0.47 | 62 | UKF1 | -0.06 | 48 |
| deeo | -0.05 | 48 | HU21 | -0.69 | 31 | UKF2 | 0.18 | 54 |
| Defo | -0.11 | 46 | HU22 | -0.62 | 33 | UKF3 | -0.72 | 30 |
| DEGO | -0.27 | 42 | HU23 | -0.82 | 27 | UKG1 | -0.09 | 47 |
| EEOO | -0.22 | 43 | HU31 | -1.03 | 22 | UKG2 | -0.34 | 40 |
| 1 IEO | -0.18 | 44 | HU32 | -1.07 | 20 | UKG3 | 0.48 | 62 |
| $1 E 02$ | 0.94 | 75 | нuz3 | -1.30 | 14 | UKH1 | 0.37 | 59 |
| GR11 | -0.93 | 24 | мтоо | -1.00 | 22 | UKH2 | 0.73 | 69 |
| GR12 | -1.08 | 20 | NL11 | -0.28 | 42 | UKH3 | 0.41 | 60 |
| GR13 | -1.29 | 15 | NL12 | -0.20 | 44 | UKI | 1.69 | 95 |
| GR14 | -1.11 | 19 | NL13 | -0.15 | 45 | UKI1 | 1.20 | 82 |
| GR21 | -1.28 | 15 | NL21 | 0.14 | 53 | UK12 | 1.10 | 79 |
| GR22 | -1.53 | 8 | NL22 | 0.42 | 61 | UK13 | 0.73 | 69 |
| GR23 | -1.31 | 14 | NL23 | 0.90 | 74 | UKJ4 | 0.21 | 55 |
| GR24 | -1.35 | 13 | NL31 | 1.55 | 91 | UKк1 | 0.82 | 71 |
| GR25 | -1.60 | 6 | NL32 | 1.43 | 88 | UKK2 | 0.23 | 56 |
| GR30 | -0.35 | 40 | NL33 | 0.92 | 74 | Uккз | -0.72 | 30 |
| GR41 | -1.50 | 9 | NL34 | 0.11 | 52 | UKK4 | -0.34 | 40 |
| GR42 | -1.50 | 9 | NL41 | 0.69 | 68 | UKL1 | -0.91 | 25 |
| GR43 | -1.56 | 7 | NL42 | 0.45 | 61 | UKL2 | 0.14 | 53 |
| ES11 | -0.85 | 26 | AT11 | -0.64 | 32 | UKM2 | 0.51 | 63 |
| ES12 | -0.92 | 25 | AT12 | -0.87 | 26 | Uкмз | 0.11 | 52 |
| ES13 | -0.91 | 25 | AT13 | 0.99 | 76 | UKM5 | 0.15 | 53 |
| ES21 | -0.46 | 37 | AT21 | -0.27 | 42 | UKM6 | -1.12 | 19 |
| ES22 | -0.86 | 26 | AT22 | -0.16 | 45 | UKNO | -0.04 | 48 |
| ES23 | -0.96 | 23 | AT31 | 0.08 | 51 |  |  |  |
| ES24 | -0.80 | 28 | AT32 | -0.08 | 47 |  |  |  |



Figure 5-45: Histogram of Business sophistication sub-score

Table 85: Business sophistication pillar sub-rank (from best to worst)

| Business sophistication |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | FR10 | 46 | NL22 | 91 | DE27 | 136 | AT21 | 181 | ITE2 | 226 | TF1 |
| 2 | UKI | 47 | UKE4 | 92 | DEA5 | 137 | NL11 | 182 | HU22 | 227 | ITF5 |
| 3 | NL31 | 48 | DE11 | 93 | DEEO | 138 | DE94 | 183 | FR63 | 228 | MT00 |
| 4 | NL32 | 49 | UKH3 | 94 | DEA3 | 139 | ITC3 | 184 | AT11 | 229 | BG42 |
| 5 | DE71 | 50 | DE25 | 95 | FR43 | 140 | DE41 | 185 | PT11 | 230 | ITG1 |
| 6 | BE00 | 51 | UKH1 | 96 | UKF1 | 141 | FR72 | 186 | CZ03 | 231 | HU31 |
| 7 | LU00 | 52 | ITD5 | 97 | DE14 | 142 | SE22 | 187 | SE21 | 232 | ES62 |
| 8 | DK01 | 53 | BG41 | 98 | DEA4 | 143 | FR91 | 188 | HU21 | 233 | BG32 |
| 9 | SE11 | 54 | DE50 | 99 | AT32 | 144 | UKG2 | 189 | PL61 | 234 | ES70 |
| 10 | UKJ1 | 55 | FR82 | 100 | DEB3 | 145 | UKK4 | 190 | FI20 | 235 | HU32 |
| 11 | DE60 | 56 | FR30 | 101 | FR41 | 146 | GR30 | 191 | UKF3 | 236 | PL33 |
| 12 | UKJ2 | 57 | PT17 | 102 | UKG1 | 147 | PL21 | 192 | UKK3 | 237 | BG33 |
| 13 | DE21 | 58 | FI18 | 103 | DE91 | 148 | PL22 | 193 | CZ05 | 238 | GR12 |
| 14 | AT13 | 59 | UKK2 | 104 | DE42 | 149 | DE80 | 194 | RO21 | 239 | ITG2 |
| 15 | BE21 | 60 | ITD3 | 105 | DEFO | 150 | FR52 | 195 | F119 | 240 | GR14 |
| 16 | IE02 | 61 | UKJ4 | 106 | FR22 | 151 | PL63 | 196 | ES41 | 241 | SE33 |
| 17 | NL33 | 62 | UKF2 | 107 | BE35 | 152 | RO31 | 197 | SIO1 | 242 | UKM6 |
| 18 | NL23 | 63 | FR42 | 108 | DED2 | 153 | DE22 | 198 | PL62 | 243 | CZ08 |
| 19 | CZ01 | 64 | UKM5 | 109 | FR24 | 154 | FR25 | 199 | ES24 | 244 | Fl13 |
| 20 | ITC4 | 65 | DK04 | 110 | FR61 | 155 | FR26 | 200 | ES53 | 245 | SK04 |
| 21 | UKK1 | 66 | DE92 | 111 | SIO2 | 156 | LVOO | 201 | PL43 | 246 | BG34 |
| 22 | DEA1 | 67 | NL21 | 112 | RO42 | 157 | PL41 | 202 | HU23 | 247 | SK03 |
| 23 | ES30 | 68 | UKL2 | 113 | DE73 | 158 | FR93 | 203 | SE32 | 248 | GR21 |
| 24 | DE30 | 69 | ITE1 | 114 | ITD4 | 159 | ES61 | 204 | FR83 | 249 | UKD1 |
| 25 | FR71 | 70 | NL34 | 115 | NL13 | 160 | DEB2 | 205 | SE31 | 250 | GR13 |
| 26 | DEA2 | 71 | UKM3 | 116 | DE72 | 161 | FR92 | 206 | ES11 | 25 | HU33 |
| 27 | PL12 | 72 | DK02 | 117 | DED1 | 162 | PL11 | 207 | PL31 | 252 | GR23 |
| 28 | UKH2 | 73 | DK03 | 118 | AT22 | 163 | FR21 | 208 | CZO2 | 253 | PT18 |
| 29 | UKJ3 | 74 | BE33 | 119 | DE24 | 164 | ES21 | 209 | CZ04 | 254 | BG31 |
| 30 | NL41 | 75 | AT31 | 120 | IE01 | 165 | TE3 | 210 | ES22 | 255 | PT30 |
| 31 | BE23 | 76 | UKD5 | 121 | RO12 | 166 | ITD2 | 211 | SK02 | 256 | GR24 |
| 32 | BE22 | 77 | FR62 | 122 | NL12 | 167 | CYOO | 212 | LTOO | 257 | PT15 |
| 33 | UKD3 | 78 | UKE2 | 123 | DE13 | 168 | UKC1 | 213 | AT12 | 258 | ES42 |
| 34 | SK01 | 79 | FR23 | 124 | SE12 | 169 | ITF3 | 214 | PL32 | 259 | ES63 |
| 35 | UKD2 | 80 | DED3 | 125 | SE23 | 170 | CZ06 | 215 | CZ07 | 260 | GR41 |
| 36 | RO32 | 81 | AT34 | 126 | BE34 | 171 | RO11 | 216 | ES13 | 261 | GR42 |
| 37 | UKM2 | 82 | BE32 | 127 | EEOO | 172 | DE93 | 217 | UKL1 | 262 | ES43 |
| 38 | BE25 | 83 | DECO | 128 | ES52 | 173 | PL42 | 218 | ES12 | 263 | GR22 |
| 39 | UKG3 | 84 | FR51 | 129 | UKD4 | 174 | ITC2 | 219 | FI1A | 264 | GR43 |
| 40 | DE12 | 85 | UKC2 | 130 | AT33 | 175 | ITD1 | 220 | GR11 | 265 | ITF6 |
| 41 | HU10 | 86 | DK05 | 131 | FR81 | 176 | RO41 | 221 | ITF2 | 266 | GR25 |
| 42 | ITE4 | 87 | UKE3 | 132 | PL51 | 177 | RO22 | 222 | ES23 | 267 | ES64 |
| 43 | NL42 | 88 | DE23 | 133 | DEB1 | 178 | PL52 | 223 | ITF4 | 268 | PT20 |
| 44 | ES51 | 89 | DE26 | 134 | DEGO | 179 | UKE1 | 224 | PT16 |  |  |
| 45 | ITC1 | 90 | UKNO | 135 | FR53 | 180 | FR94 | 225 | PL34 |  |  |

### 5.11 Innovation

Candidate indicators are discussed in Section3.11. In the following we recall them together with the short names used in the analysis.

Indicators included, in brackets short names:

1. Innovation patent applications per mill. Inhabitants (Inno_patent_appl)
2. Total patent applications per mill. Inhabitants (Total_patent_appl)
3. Core Creative class employment (share of population) (Core_creative_class)
4. Knowledge workers (share of total employment)
(Knowledge_workers)
5. Scientific publications per mill. Inhabitants (Scientific_publications)
6. Intramural R\&D expenditure (share of GDP)
(GERD)
7. Human resources in Science \& Technology (share of labor force ${ }^{15}$ )
(HRST)
8. Employment in Tech.\& knowledge-intensive sectors (share of total employment)
(High_tech_emp)
9. High-tech EPO applications per mill. Inhabitants
(High_tech_inventors)
10. ICT EPO applications per mill. Inhabitants
(ICT_inventors)
11. Biotechnology EPO applications per mill. Inhabitants
(Biotech_inventors)

## Imputation of missing data

All indicators have the same positive orientation with respect to the level of competitiveness. For the indicator on Core creative class, NUTS 0 data has been imputed to the NUTS 2 level for Denmark.

For the indicator on Scientific publications, NUTS 0 data has been imputed to the NUTS 2 for Denmark and Slovenia, while NUTS 1 (UKI) data has been imputed to the NUTS 2 level (UKI 1 and 2).

[^13]For the indicator on GERD, NUTS 1 data has been imputed to the NUTS 2 level for Belgium. Due to lack of more recent data, 2004 data has been used for France, 2005 - for Italy, and 2003 - for the Netherlands.

For the indicator on high-tech employment, due to lack of more recent data, 2007 data has been used for Bulgaria, Poland, Slovenia, Sweden, DE22, DE80, DEC0, DED1, and DED3, 2004 - for DE50 andGR13, and 2006 - for GR14.

## Univariate analysis

Table 86 reports the descriptive statistics for the innovation pillar indicators. Most indicators have a very low percentage of missing data (below $3 \%$ ). The only two exceptions are the indicators on knowledge workers ( $8.21 \%$ ) and on employment in knowledge and technology intensive sectors $(4.48 \%)$, but both are below the threshold of missing data defined in Section 4.2. Thus, all indicators have been included in the analysis. All indicators related to patents have a high coefficient of variation, a sign of the very diverse innovation output activities across EU regions.

Table 86：Descriptive statistics of Innovation indicators

|  |  |  | － | $\bigcirc$ |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  | O- |  |  |  |  |  | $\bigcirc$ |  |
|  |  |  | 菅 | $\underset{\infty}{-1} \stackrel{\infty}{\infty}$ | $\mid$ |  |  |  |  |  |  |
|  |  |  |  | ¢ | $\stackrel{\circ}{\circ} \underset{\underset{\sim}{7}}{\sim}$ |  |  |  |  |  |  |
|  |  |  |  | o웅 | $\left\lvert\, \begin{array}{ccc} \vec{\sim} \\ \underset{\sim}{c} & \underset{z}{z} & 8 \\ \hline \end{array}\right.$ |  |  |  | O | $\stackrel{\substack{\text { ¢ }}}{\substack{\text { ¢ }}}$ | $\stackrel{\text { m }}{\text { m }}$ |
|  |  | 들 $\stackrel{\rightharpoonup}{u}$ 0 0 0 |  |  |  |  | $\begin{aligned} & \stackrel{y y}{0} \\ & \stackrel{4}{\circ} \\ & \vdots \\ & \stackrel{0}{0} \\ & \stackrel{\circ}{\circ} \\ & \stackrel{\circ}{0} \\ & 0 \end{aligned}$ |  | $\stackrel{\sim}{\circ}$ | $\underset{\sim}{\text { 寸i }} \underset{\sim}{\omega} \underset{\sim}{\omega} \underset{\sim}{\infty} \underset{\infty}{0} \underset{\sim}{\sim}$ |  |
|  | $\begin{aligned} & \text { 흘 } \\ & \text { 흔 } \\ & \text { ition } \\ & \hline \end{aligned}$ | ジֶun |  |  |  |  |  | \＃゙った | （1） |  |  |

## How do EU regions score in each of the indicators?

We can see from Figure 5-46 that Scandinavian regions have very high scores on all innovation indicators. Eastern European regions (Bulgarian and Romanian, in particular) have the worst performance. Some Southern European regions in Greece, Portugal, Spain and Italy show low performance as well.

The blue banana, the banana-shaped metropolitan axis, running from London through Benelux and the Rhine area to the Northern part of Italy, often identified as the area with greatest development potential in Europe in terms of innovation, still seems to be an accurate representation of innovation patterns across EU regions.



## Biotech inventors ${ }^{1}$



Figure 5-46: Best and worst performing regions for each indicator - Innovation
The next step in our analysis is the analysis of the distribution of the different indicators and their transformation. Table 87 shows the initial distribution of each indicator and the method used for its transformation. The approach adopted has been described in detail in Section 4.3. All indicators which have been transformed show a clear positive skewness. Due to the presence of zero values, all indicators on patents have been transformed logarithmically. The indicators on scientific publications and intramural R\&D expenditure have been transformed with the Box-Cox method.

[^14]Table 87: Histograms of Innovation indicators
Innovation patent application



Total patent application




Knowledge workers


Scientific publications


GERD




High tech inventors



ICT inventors


## Biotech inventors



Note: In the case of the Scientific Publications indicator, the lambda used has been set to 0.15

## MULTIVARIATE ANALYSIS

Despite the high number of indicators which describe this pillar and their different sources, the PCA analysis depicts a pillar with a clear single latent dimension, well represented by all the selected indicators (see Figure 5-47 and Table 89). Table 90 shows that the first PCA component alone explains more than $73 \%$ of total variation and from the component loadings (Table 89) one can see that the contribution of each indicator to this component is approximately the same. The analysis fully supports the starting hypothesis of a unique underlying dimension and, consequently, the simple choice of equal weights for the computation of the Innovation sub-score, which is displayed in Figure 5-48. The histogram of the Innovation sub-score is shown in Figure 5-49 while Table 92 lists the reordered regions (from best to worst).

Table 88: Correlation matrix between indicators included in the Innovation pillar

|  |  | Inno_patent_ <br> appl | Total_patent_ appl | Core_creative_ <br> class | Knowledge_ workers | Scientific_ publications | Gerd | HRST | High_tech_empl | High_tech_ inventors | $\begin{gathered} \text { ICT_- }_{2} \\ \text { inventors } \end{gathered}$ | Biotech_ inventors |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Correlation | Inno_patent_appl | 1.000 | . 936 | . 654 | . 719 | . 609 | . 763 | . 725 | . 640 | . 979 | . 994 | . 805 |
|  | Total_patent_appl | . 936 | 1.000 | . 591 | . 690 | . 590 | . 756 | . 705 | . 541 | . 883 | . 933 | . 727 |
|  | Core_creative_class | . 654 | . 591 | 1.000 | . 765 | . 606 | . 604 | . 797 | . 614 | . 646 | . 645 | . 657 |
|  | Knowledge_workers | . 719 | 690 | . 765 | 1.000 | . 581 | . 613 | . 838 | . 685 | . 713 | . 709 | . 657 |
|  | Scientific_publications | . 609 | . 590 | . 606 | . 581 | 1.000 | . 667 | . 624 | . 485 | . 600 | . 593 | . 656 |
|  | Gerd | . 763 | . 756 | . 604 | . 613 | . 667 | 1.000 | . 682 | . 594 | . 751 | . 749 | . 689 |
|  | HRST | . 725 | . 705 | . 797 | . 838 | . 624 | . 682 | 1.000 | . 650 | . 699 | . 714 | . 672 |
|  | High_tech_empl | . 640 | . 541 | . 614 | . 685 | 485 | . 594 | . 650 | 1.000 | . 653 | . 635 | . 585 |
|  | High_tech_inventors | . 979 | . 883 | . 646 | . 713 | . 600 | . 751 | . 699 | . 653 | 1.000 | . 973 | . 817 |
|  | ICT_inventors | . 994 | . 933 | . 645 | . 709 | . 593 | . 749 | . 714 | . 635 | . 973 | 1.000 | . 773 |
|  | Biotech_inventors | . 805 | . 727 | . 657 | . 657 | 656 | . 689 | . 672 | . 585 | . 817 | . 773 | 1.000 |
| Sig. (1-tailed) | Inno_patent_appl |  | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |
|  | Total_patent_appl | . 000 |  | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |
|  | Core_creative_class | . 000 | . 000 |  | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |
|  | Knowledge_workers | . 000 | . 000 | . 000 |  | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |
|  | Scientific_publications | . 000 | . 000 | . 000 | . 000 |  | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |
|  | Gerd | . 000 | . 000 | . 000 | . 000 | . 000 |  | . 000 | . 000 | . 000 | . 000 | . 000 |
|  | HRST | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |  | . 000 | . 000 | . 000 | . 000 |
|  | High_tech_empl | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |  | . 000 | . 000 | . 000 |
|  | High_tech_inventors | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |  | . 000 | . 000 |
|  | ICT_inventors | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |  | . 000 |
|  | Biotech_inventors | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 | . 000 |  |

Scree Plot


Figure 5-47: PCA analysis of the Innovation pillar - eigenvalues

Table 89: PCA analysis Innovation pillar: correlation coefficients between indicators and PCA components

Component Matrix ${ }^{\text {a }}$

|  | Component |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 |
| Inno_patent_appl | . 946 | -. 288 | -. 089 | -. 037 | . 023 | . 039 | . 064 | -. 016 | -. 041 | -. 044 | -. 049 |
| Total_patent_appl | . 896 | -. 316 | -. 030 | -. 126 | -. 136 | . 081 | . 005 | -. 025 | . 230 | . 038 | . 003 |
| Core_creative_class | . 800 | . 420 | -. 010 | -. 212 | . 094 | -. 210 | . 284 | . 053 | . 041 | . 004 | . 000 |
| Knowledge_workers | . 845 | . 325 | -. 181 | -. 154 | -. 049 | . 174 | -. 186 | . 238 | -. 016 | -. 008 | . 000 |
| Scientific_publications | . 737 | . 170 | . 588 | . 099 | . 039 | . 253 | . 078 | -. 012 | -. 011 | . 000 | . 000 |
| Gerd | . 837 | -. 078 | . 228 | . 195 | -. 336 | -. 278 | -. 069 | . 093 | -. 030 | -. 006 | . 000 |
| HRST | . 859 | . 325 | -. 054 | -. 188 | -. 149 | -. 008 | -. 136 | -. 272 | -. 059 | . 004 | . 001 |
| High_tech_empl | . 746 | . 277 | -. 310 | . 508 | . 036 | . 057 | . 048 | -. 049 | . 054 | . 003 | . 000 |
| High_tech_inventors | . 934 | -. 274 | -. 096 | . 009 | . 087 | . 018 | . 054 | . 026 | -. 154 | . 089 | . 006 |
| ICT_inventors | . 935 | -. 299 | -. 111 | -. 041 | . 001 | . 060 | . 097 | -. 018 | -. 048 | -. 071 | . 039 |
| Biotech_inventors | . 856 | -. 076 | . 140 | . 027 | . 394 | -. 176 | -. 226 | -. 015 | . 053 | -. 009 | . 003 |

Extraction Method: Principal Component Analysis.
a. 11 components extracted.

Table 90: PCA analysis for Innovation pillar: explained variance

| Component | Initial Eigenvalues |  |  |
| :---: | ---: | ---: | ---: |
|  | Total | \% of Variance | Cumulative \% |
| 1 | 8.072 | 73.378 | 73.378 |
| 2 | .852 | 7.746 | 81.124 |
| 3 | .580 | 5.270 | 86.394 |
| 4 | .430 | 3.908 | 90.302 |
| 5 | .331 | 3.007 | 93.310 |
| 6 | .262 | 2.383 | 95.693 |
| 7 | .214 | 1.947 | 97.640 |
| 8 | .147 | 1.332 | 98.972 |
| 9 | .093 | .841 | 99.814 |
| 10 | .017 | .150 | 99.964 |
| 11 | .004 | .036 | 100.000 |



Figure 5-48: Map of Innovation sub-score.
Min-max normalized values are shown in Table 91

Table 91: Innovation sub-score as arithmetic mean of transformed and standardized indicators.

| region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore | region | Subscore | Min_max normalized subscore |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BEOO | 1.28 | 84 | ES30 | 0.58 | 66 | AT33 | 0.26 | 58 |
| BE21 | 0.72 | 69 | ES41 | -0.58 | 36 | AT34 | -0.01 | 51 |
| BE22 | 0.11 | 54 | ES42 | -1.03 | 25 | PL11 | -1.07 | 24 |
| BE23 | 0.93 | 75 | ES43 | -1.13 | 22 | PL12 | -0.27 | 44 |
| BE25 | 0.22 | 57 | ES51 | -0.02 | 51 | PL21 | -0.82 | 30 |
| BE32 | -0.01 | 51 | ES52 | -0.56 | 37 | PL22 | -0.98 | 26 |
| BE33 | 0.48 | 63 | ES53 | -0.94 | 27 | PL31 | -1.26 | 19 |
| BE34 | -0.01 | 51 | ES61 | -0.84 | 30 | PL32 | -1.12 | 22 |
| BE35 | 0.70 | 69 | ES62 | -0.96 | 27 | PL33 | -1.66 | 9 |
| BG31 | -1.61 | 10 | ES63 | -1.69 | 8 | PL34 | -1.30 | 18 |
| BG32 | -1.66 | 9 | ES64 | -1.46 | 14 | PL41 | -1.12 | 22 |
| BG33 | -1.62 | 10 | ES70 | -1.12 | 22 | PL42 | -0.97 | 26 |
| BG34 | -1.73 | 7 | FR10 | 1.54 | 90 | PL43 | -0.97 | 26 |
| BG41 | -0.38 | 41 | FR21 | -0.62 | 35 | PL51 | -0.84 | 30 |
| BG42 | -1.57 | 11 | FR22 | -0.34 | 42 | PL52 | -1.29 | 18 |
| cz01 | 0.95 | 75 | FR23 | -0.33 | 43 | PL61 | -1.35 | 17 |
| cz02 | -0.46 | 39 | FR24 | 0.08 | 53 | PL62 | -1.36 | 16 |
| czo3 | -0.64 | 35 | FR25 | 0.10 | 54 | PL63 | -0.89 | 28 |
| CZ04 | -1.09 | 23 | FR26 | -0.29 | 44 | PT11 | -1.16 | 21 |
| czo5 | -0.62 | 35 | FR30 | -0.24 | 45 | PT15 | -1.33 | 17 |
| czo6 | -0.45 | 40 | FR41 | -0.32 | 43 | PT16 | -1.11 | 23 |
| czo7 | -0.85 | 29 | FR42 | 0.39 | 61 | PT17 | -0.14 | 47 |
| czo8 | -0.80 | 31 | FR43 | -0.13 | 48 | PT18 | -1.12 | 22 |
| DK01 | 1.67 | 94 | FR51 | -0.08 | 49 | PT20 | -1.75 | 6 |
| DK02 | 0.67 | 68 | FR52 | 0.40 | 61 | Ртз0 | -1.66 | 9 |
| DK03 | 0.28 | 58 | FR53 | -0.41 | 41 | RO11 | -1.59 | 10 |
| DK04 | 0.62 | 67 | FR61 | -0.06 | 49 | RO12 | -1.72 | 7 |
| DK05 | 0.45 | 63 | FR62 | 0.88 | 73 | RO21 | -1.67 | 8 |
| DE11 | 1.11 | 79 | FR63 | -0.43 | 40 | R022 | -2.00 | 0 |
| DE12 | 1.43 | 88 | FR71 | 0.79 | 71 | RO31 | -1.81 | 5 |
| DE13 | 1.14 | 80 | FR72 | 0.26 | 58 | R032 | -0.16 | 47 |
| DE14 | 0.99 | 76 | FR81 | 0.22 | 57 | RO41 | -1.86 | 4 |
| DE21 | 1.77 | 96 | FR82 | 0.53 | 65 | RO42 | -1.49 | 13 |
| DE22 | -0.06 | 49 | FR83 | -0.59 | 36 | S101 | -0.51 | 38 |
| DE23 | 0.74 | 70 | FR91 | -1.13 | 22 | S102 | 0.23 | 57 |
| DE24 | 0.50 | 64 | FR92 | -1.12 | 22 | Sk01 | 0.47 | 63 |
| DE25 | 1.28 | 84 | FR93 | -0.83 | 30 | SK02 | -1.08 | 23 |
| DE26 | 0.63 | 67 | FR94 | -1.28 | 18 | Sk03 | -1.25 | 19 |
| DE27 | 0.20 | 56 | ITC1 | 0.15 | 55 | SK04 | -1.11 | 23 |
| DE30 | 1.38 | 86 | ITC2 | -0.49 | 39 | F113 | 0.36 | 60 |
| DE41 | 0.09 | 53 | ITC3 | 0.18 | 56 | F18 | 1.61 | 92 |
| DE42 | 0.52 | 64 | ITC4 | 0.29 | 58 | F19 | 0.91 | 74 |
| DE50 | 0.62 | 67 | ITD1 | -0.48 | 39 | F11A | 1.05 | 78 |
| DE60 | 1.16 | 81 | ITD2 | -0.17 | 47 | F120 | -0.37 | 42 |
| DE71 | 1.21 | 82 | ITD3 | -0.25 | 45 | SE11 | 1.92 | 100 |
| DE72 | 0.80 | 71 | ITD4 | -0.06 | 49 | SE12 | 1.19 | 81 |
| DE73 | -0.10 | 48 | ITD5 | 0.07 | 53 | SE21 | -0.11 | 48 |
| DE80 | 0.04 | 52 | ITE1 | -0.03 | 50 | SE22 | 1.42 | 87 |
| DE91 | 0.86 | 73 | ITE2 | -0.47 | 39 | SE23 | 1.03 | 77 |
| DE92 | 0.89 | 74 | ITE3 | -0.51 | 38 | SE31 | -0.20 | 46 |
| DE93 | 0.01 | 51 | ITE4 | 0.23 | 57 | SE32 | -0.24 | 45 |
| DE94 | -0.31 | 43 | ITF1 | -0.37 | 42 | SE33 | 0.82 | 72 |
| DEA1 | 0.63 | 67 | ITF2 | -0.90 | 28 | UKC1 | -0.02 | 51 |
| DEA2 | 1.12 | 80 | ITF3 | -0.54 | 37 | UKC2 | -0.07 | 49 |
| DEA3 | 0.20 | 56 | ITF4 | -0.77 | 31 | UKD1 | -0.53 | 38 |
| DEA4 | 0.53 | 65 | ITF5 | -0.74 | 32 | UKD2 | 0.75 | 70 |
| DEA5 | 0.13 | 54 | ITF6 | -0.91 | 28 | UKD3 | 0.18 | 56 |
| Deb1 | -0.07 | 49 | ITG1 | -0.49 | 39 | UKD4 | -0.18 | 46 |
| DEB2 | -0.09 | 49 | ITG2 | -0.73 | 32 | UKD5 | 0.01 | 51 |
| DEB3 | 0.97 | 76 | cyoo | -0.76 | 32 | UKE1 | -0.57 | 36 |
| DECO | 0.26 | 58 | Lvoo | -0.80 | 31 | UKE2 | 0.44 | 62 |
| DED1 | -0.02 | 51 | Ltoo | -0.74 | 32 | UKE3 | 0.08 | 53 |
| DED2 | 0.73 | 70 | Luoo | 0.46 | 63 | UKE4 | 0.03 | 52 |
| DED3 | 0.44 | 62 | HU10 | 0.35 | 60 | UKF1 | 0.27 | 58 |
| DEEO | -0.03 | 50 | HU21 | -0.90 | 28 | UKF2 | 0.40 | 61 |
| DEFO | 0.34 | 60 | HU22 | -1.10 | 23 | UKF3 | -0.67 | 34 |
| DEGO | 0.42 | 62 | HU23 | -0.78 | 31 | UKG1 | 0.44 | 62 |
| eeoo | -0.22 | 45 | HU31 | -1.18 | 21 | UKG2 | -0.21 | 46 |
| 1 E01 | -0.01 | 51 | HU32 | -0.91 | 28 | UKG3 | -0.02 | 51 |
| $1 E 02$ | 0.46 | 63 | HU33 | -0.76 | 32 | UKH1 | 1.17 | 81 |
| GR11 | -1.41 | 15 | мтоо | -0.46 | 39 | UKH2 | 1.03 | 77 |
| GR12 | -0.71 | 33 | NL11 | 0.91 | 74 | Uкнз | 0.55 | 65 |
| GR13 | -1.40 | 15 | NL12 | -0.24 | 45 | UKI | 0.92 | 74 |
| GR14 | -1.07 | 24 | NL13 | 0.14 | 55 | UKJ1 | 1.63 | 93 |
| GR21 | -0.88 | 29 | NL21 | 0.40 | 61 | UKJ2 | 1.04 | 78 |
| GR22 | -1.80 | 5 | NL22 | 0.87 | 73 | UKJ3 | 1.02 | 77 |
| GR23 | -1.01 | 25 | NL23 | 0.37 | 60 | UKJ4 | 0.18 | 56 |
| GR24 | -1.62 | 10 | NL31 | 1.47 | 89 | UKK1 | 1.01 | 77 |
| GR25 | -1.57 | 11 | NL32 | 0.96 | 76 | Uкк2 | 0.06 | 53 |
| GR30 | -0.21 | 46 | NL33 | 0.93 | 75 | Uкк3 | -0.52 | 38 |
| GR41 | -1.17 | 21 | NL34 | -0.09 | 49 | UKK4 | -0.09 | 49 |
| GR42 | -1.55 | 11 | NL41 | 1.34 | 85 | UKL1 | -0.30 | 43 |
| GR43 | -0.64 | 35 | NL42 | 0.81 | 72 | UKL2 | 0.42 | 62 |
| ES11 | -0.66 | 34 | AT11 | -0.72 | 33 | UKM2 | 0.69 | 69 |
| ES12 | -0.58 | 36 | AT12 | 0.01 | 51 | UKM3 | 0.14 | 55 |
| ES13 | -0.63 | 35 | AT13 | 1.17 | 81 | UKM5 | 0.72 | 69 |
| ES21 | 0.23 | 57 | AT21 | -0.01 | 51 | UKM6 | 0.12 | 54 |
| ES22 | 0.13 | 54 | AT22 | 0.38 | 61 | UKNO | -0.03 | 50 |
| ES23 | -0.81 | 30 | AT31 | 0.02 | 52 |  |  |  |
| ES24 | -0.50 | 38 | AT32 | 0.16 | 55 |  |  |  |

Figure 5-49: Histogram of Innovation sub-score


Table 92: Innovation pillar sub-rank (from best to worst)

| Innovation |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | SE11 | 46 | DE23 | 91 | BE25 | 136 | DEB2 | 181 | ES12 | 226 | PT16 |
| 2 | DE21 | 47 | DED2 | 92 | FR81 | 137 | NL34 | 182 | ES41 | 227 | SK04 |
| 3 | DK01 | 48 | BE21 | 93 | DE27 | 138 | UKK4 | 183 | FR83 | 228 | ES70 |
| 4 | UKJ1 | 49 | UKM5 | 94 | DEA3 | 139 | DE73 | 184 | CZ05 | 229 | FR92 |
| 5 | FI18 | 50 | BE35 | 95 | ITC3 | 140 | SE21 | 185 | FR21 | 230 | PL32 |
| 6 | FR10 | 51 | UKM2 | 96 | UKD3 | 141 | FR43 | 186 | ES13 | 231 | PL41 |
| 7 | NL31 | 52 | DK02 | 97 | UKJ4 | 142 | PT17 | 187 | CZ03 | 232 | PT18 |
| 8 | DE12 | 53 | DE26 | 98 | AT32 | 143 | RO32 | 188 | GR43 | 233 | ES43 |
| 9 | SE22 | 54 | DEA1 | 99 | ITC1 | 144 | ITD2 | 189 | ES11 | 234 | FR91 |
| 10 | DE30 | 55 | DK04 | 100 | NL13 | 145 | UKD4 | 190 | UKF3 | 235 | PT11 |
| 11 | NL41 | 56 | DE50 | 101 | UKM3 | 146 | SE31 | 191 | GR12 | 236 | GR41 |
| 12 | BEOO | 57 | ES30 | 102 | DEA5 | 147 | GR30 | 192 | AT11 | 237 | HU31 |
| 13 | DE25 | 58 | UKH3 | 103 | ES22 | 148 | UKG2 | 193 | ITG2 | 238 | SK03 |
| 14 | DE71 | 59 | DEA4 | 104 | UKM6 | 149 | EEOO | 194 | ITF5 | 239 | PL31 |
| 15 | SE12 | 60 | FR82 | 105 | BE22 | 150 | FR30 | 195 | LTOO | 240 | FR94 |
| 16 | AT13 | 61 | DE42 | 106 | FR25 | 151 | NL12 | 196 | CYOO | 241 | PL52 |
| 17 | UKH1 | 62 | DE24 | 107 | DE41 | 152 | SE32 | 197 | HU33 | 242 | PL34 |
| 18 | DE60 | 63 | BE33 | 108 | FR24 | 153 | ITD3 | 198 | ITF4 | 243 | PT15 |
| 19 | DE13 | 64 | SK01 | 109 | UKE3 | 154 | PL12 | 199 | HU23 | 244 | PL61 |
| 20 | DEA2 | 65 | IE02 | 110 | ITD5 | 155 | FR26 | 200 | CZ08 | 245 | PL62 |
| 21 | DE11 | 66 | LU00 | 111 | UKK2 | 156 | UKL1 | 201 | LV00 | 246 | GR13 |
| 22 | FI1A | 67 | DK05 | 112 | DE80 | 157 | DE94 | 202 | ES23 | 247 | GR11 |
| 23 | UKJ2 | 68 | DED3 | 113 | UKE4 | 158 | FR41 | 203 | PL21 | 248 | ES64 |
| 24 | SE23 | 69 | UKE2 | 114 | AT31 | 159 | FR23 | 204 | FR93 | 249 | RO42 |
| 25 | UKH2 | 70 | UKG1 | 115 | DE93 | 160 | FR22 | 205 | ES61 | 250 | GR42 |
| 26 | UKJ3 | 71 | DEGO | 116 | AT12 | 161 | ITF1 | 206 | PL51 | 251 | BG42 |
| 27 | UKK1 | 72 | UKL2 | 117 | UKD5 | 162 | FI20 | 207 | CZ07 | 252 | GR25 |
| 28 | DE14 | 73 | FR52 | 118 | BE32 | 163 | BG41 | 208 | GR21 | 253 | RO11 |
| 29 | DEB3 | 74 | NL21 | 119 | BE34 | 164 | FR53 | 209 | PL63 | 254 | BG31 |
| 30 | NL32 | 75 | UKF2 | 120 | IE01 | 165 | FR63 | 210 | ITF2 | 255 | BG33 |
| 31 | CZ01 | 76 | FR42 | 121 | AT21 | 166 | CZ06 | 211 | HU21 | 256 | GR24 |
| 32 | BE23 | 77 | AT22 | 122 | AT34 | 167 | CZO2 | 212 | ITF6 | 257 | BG32 |
| 33 | NL33 | 78 | NL23 | 123 | DED1 | 168 | MT00 | 213 | HU32 | 258 | PL33 |
| 34 | UKI | 79 | FI13 | 124 | ES51 | 169 | ITE2 | 214 | ES53 | 259 | PT30 |
| 35 | NL11 | 80 | HU10 | 125 | UKC1 | 170 | ITD1 | 215 | ES62 | 260 | RO21 |
| 36 | FI19 | 81 | DEFO | 126 | UKG3 | 171 | ITC2 | 216 | PL42 | 261 | ES63 |
| 37 | DE92 | 82 | ITC4 | 127 | DEEO | 172 | ITG1 | 217 | PL43 | 262 | RO12 |
| 38 | FR62 | 83 | DK03 | 128 | ITE1 | 173 | ES24 | 218 | PL22 | 263 | BG34 |
| 39 | NL22 | 84 | UKF1 | 129 | UKNO | 174 | ITE3 | 219 | GR23 | 264 | PT20 |
| 40 | DE91 | 85 | DECO | 130 | DE22 | 175 | SIO1 | 220 | ES42 | 265 | GR22 |
| 41 | SE33 | 86 | FR72 | 131 | FR61 | 176 | UKK3 | 221 | GR14 | 266 | RO31 |
| 42 | NL42 | 87 | AT33 | 132 | ITD4 | 177 | UKD1 | 222 | PL11 | 267 | RO41 |
| 43 | DE72 | 88 | ES21 | 133 | DEB1 | 178 | ITF3 | 223 | SK02 | 268 | RO22 |
| 44 | FR71 | 89 | ITE4 | 134 | UKC2 | 179 | ES52 | 224 | CZ04 |  |  |
| 45 | UKD2 | 90 | SIO2 | 135 | FR51 | 180 | UKE1 | 225 | HU22 |  |  |

## 6 The Regional Competitiveness Index

The final setting up of the RCI is based upon the sub-score values computed for the eleven different pillars presented in Chapter 5.

For the final aggregation we followed the approach that World Economic Forum adopts for the Global Competitiveness Index (Schwab and Porter, 2007; Schwab, 2009), as discussed in Section 2.1. Given the high level of heterogeneity of European regions, especially after the 2004 and 2007 enlargements, our aim is to weight different regions according to their level of development. It is, in fact, clear that different pillars affect different regions differently: the competitiveness of a region close to London or Ile de France is driven by factors which are intrinsically different than those which can drive the competitiveness of Eastern European regions. As regions move along the path of development, their socio-economic conditions change and other determinants become more important for the regional level of competitiveness. For this reason, the best way to improve competitiveness of more developed regions is not the same as the best way to make less developed regions catch up.

WEF classifies the countries into three major groups of 'basic', 'efficiency' and 'innovation' driven economies, and two 'transition' groups with feature intermediate stages between the three major groups. The WEF classification is based upon two criteria: the level of GDP per capita at market exchange rates and the extent to which countries are driven by factor endowments (mostly unskilled labor and natural resources). Being not directly measured, the second criterion is approximated by the share of export of mineral goods in total exports. This last criterion is clearly not applicable to the RCI case.

In order to get a first impression on where EU regions are placed in terms of their stage of development as defined by WEF, we have used as a reference the WEF GCI 2010 thresholds for classifying EU regions on the basis of their stage of development. Given that the thresholds are defined in US dollars, we have used the purchasing-power-parity (PPP) conversion method in order to obtain equivalents in euros. The PPP method provides a more accurate comparison than the exchange rate conversion as it is the rate at which the currency of one country needs to be converted into that of a second country to ensure that a given amount of the first country's currency will purchase the same volume of goods and
services in the second country. We use the OECD PPP for GDP data, taking as a reference year 2007. With the premise of all the limitations of such a conversion methodology, the results still give an indication as to how EU regions are placed across the stages of development defined by the WEF GCI. The majority of EU regions (89.5\%) fall under the innovation-driven stage of development, as defined by WEF, while $10 \%$ belongs to the transition stage between efficiency and innovation driven economies. Only one region out of 268 is placed in the efficiency-driven group. This suggests that the classification method used by WEF is not discriminating enough among European regions. The WEF approach has been consequently modified to better describe the European situation.

In the RCI case, regional economies are divided into 'medium', 'transition' and 'high' stage of development. The development stage of the regions is computed on the basis of the regional GDP at current market prices (year 2007) measured as PPP per inhabitants and expressed as percentage of the EU average - GDP\%. The table showing the singles stage of development for each EU region is shown in Appendix F. EU regions are then classified into three groups of medium, transition or high stage according to a GDP\% respectively lower than $75 \%$, between $75 \%$ and $100 \%$ and above $100 \%$, (Table 93).

Table 93: Thresholds (\% GDP) for the definition of stages of development

| Stage of development | $\%$ of GDP (PPP/inhabitants) |
| :--- | :--- |
| Medium | $<75 \quad\left(\mathrm{t}_{1}=75\right)$ |
| Transition | $\geq 75$ and $<100$ |
| High | $\geq 100 \quad\left(\mathrm{t}_{2}=100\right)$ |

The threshold which defines the level 'medium' ( $\mathrm{t}_{1}=75 \%$ of EU average) is the value defined by the EU Commission - Regional Policy 2007-2013 - to identify regions eligible for the 'Convergence’ objective. This threshold is highly relevant as it affects EU policy funding. The second threshold, $\mathrm{t}_{2}=100 \%$, is instead more arbitrary and has been examined by an uncertainty analysis, as discussed in Section 6.2 below.

Different regions are weighted differently according to their level of development. Three groups of pillars are identified, mostly coinciding with the WEF groups (the only exception is for the Technological readiness pillar which is assigned to the intermediate group by WEF and it is here assigned to the third group). The first group of pillars includes Institutions, Macroeconomic Stability Infrastructure, Health, and Quality of Primary and Secondary Education (see Figure 6-1). These are considered as factors which are strictly necessary for the basic functioning of any economy. They, in fact, cover aspects such as unskilled or low skilled labor force, infrastructures, quality of governance and public health. The simple average of these pillars gives the first competitiveness sub-index (sub_index 1, Figure 6-1). The second group of pillars includes Higher Education/Training and Lifelong Learning, Labor Market Efficiency and Market Size. They describe an economy which is more sophisticated, with a higher potential skilled labor force and a structured labor market. These pillars are used for the computation of the second sub_index as the average of the three pillar sub-scores (sub_index 2, Figure 6-1). The last group of pillars comprises all the high tech and innovation related pillars: Technological Readiness, Business Sophistication and Innovation. A region with high scores in these sectors is expected to have the most competitive economy. The third sub_index is computed as simple average of the sub-scores of the third group of pillars (sub_index 3, Figure 6-1).

Given the pillar classification, EU regions are assigned different weights according to their development stage. The set of weights assigned for the RCI computation stems from the WEF approach with some modifications to accommodate for the fact that EU regions do not show the same level of heterogeneity, in terms of stages of development, as the countries covered by WEF. EU regions show on average a medium-high level of development, with only $25 \%$ of the regions ( 68 out of 268 ) in the medium stage of development according to the WEF criteria (Table 94). We have, thus, tried to avoid an excessive penalization of the sub_index related to the innovative aspect of competitiveness by using a slightly different weighting scheme. In any case, the choice of the weighting scheme has been examined by a full robustness analysis, as detailed in Section 6.2 of this chapter.

The regions classified into the 'medium' stage are assigned the weights that WEF assigns to the efficiency-driven economy (corresponding to the WEF intermediate group), while the weights of the 'high' stage are those which WEF uses for the innovative-driven economy.

The weights of the 'intermediate' stage of development have been chosen as the middle point between the weights of the first and third stages. Figure 6-1 displays a sketch of the pillar-groups and development stage weights.

For each region, the stage of development is assessed and the three sub_indices corresponding to the three groups of pillars are computed as simple average of the pillar sub-scores. For the computation of the overall RCI index, each sub_index is then weighted differently to reflect its relevance in defining the final index on the basis of the region's development stage. For medium economies the set of weights is: $\mathrm{w}_{\mathrm{M} 1}=0.4$ for sub_index 1 , $\mathrm{w}_{\mathrm{M} 2}=0.5$ for sub_index 2 and $\mathrm{w}_{\mathrm{M} 3}=0.1$ for sub_index 3 . This reflects a situation where, given that the economy is mostly driven by basic and intermediate socio-economic factors, the first and second groups of pillars are assigned almost all the weight $(90 \%$ ), while the innovationrelated group is assigned the lowest weight ( $10 \%$ ). For intermediate economies, the set of weights is: $\mathrm{w}_{11}=0.3$ for sub_index $1, \mathrm{w}_{12}=0.5$ for sub_index 2 and $\mathrm{w}_{13}=0.2$ for sub_index 3 . With respect to the medium-stage, the role of the third group of pillars is given more relevance. For high-stage economies weights are defined as: $\mathrm{w}_{\mathrm{H} 1}=0.2$ for sub_index 1 , $\mathrm{w}_{\mathrm{H} 2}=0.5$ for sub_index 2 and $\mathrm{w}_{\mathrm{H} 3}=0.3$ for sub_index 3. In this type of economies basic factors have the lowest relevance while the innovative group of pillars is assigned a relatively high importance.

It can be seen that for all development stages the highest weight is assigned to the second pillar group. The importance of the first group of pillar decreases going from medium to high stage of development, while the last pillar group is correspondingly gaining importance.

It is worth noting that, in general, theoretical weights assigned to the components of a composite do not necessarily reflect their effective weight in the final composite. In fact, when combining multiple indicators into a single linear index, the weights of the linear combination determine the tradeoff between indicators rather than their effective relevance in the final score (Patil and Taillie, 2004). Their actual weight depends on the observed data, the sub-score distributions and the number of indicators included in each pillar group, if any. In the RCI case theoretical weights are expected to be close to the effective weights as subscore distributions have similar variances (Chapter 5) and the three groups of pillars include roughly the same number of indicators.


Figure 6-1: The 11 pillars of RCI classified into three groups and weighting scheme for each development stage

### 6.1 RCI regional scores

Table 94 shows GDP $\%$, development stage and value of the three sub_indices for all the 268 NUTS2 EU regions.

Table 94：RCI sub＿indices

|  | com． |  |  |  |  |  | cop， | stact |  |  |  | nection |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| ，mien |  | mct | ${ }_{\text {a }}^{0}$ |  |  |  |  |  |  |  | ${ }^{10,03}$ |  |  |  |  |  |  |
| \％eas | cos | meen | cose |  |  | ${ }_{\text {cks }}$ | cin |  | 200 |  |  | Sun |  |  | cois | 边 |  |
|  | cin |  |  |  |  | ssi |  | mer | coicce | 0129 |  | 保 | cis | Meum |  | coin |  |
| ${ }^{\text {asa }}$ |  |  | coind |  |  | ciss | ${ }^{2} 590$ |  | ．oas | ${ }_{0}$ oso | ${ }_{\substack{\text { Oss }}}^{\substack{\text { Os7 }}}$ |  | 3380 |  | 0.58 |  |  |
| ${ }^{\text {ara }}$ |  |  | coin | （inct | cisis |  | cosize |  | 0 |  |  |  |  |  |  | \％ |  |
| deas |  |  | ．136 | cin |  | $\underset{\substack{\text { man } \\ \text { man }}}{ }$ | 9，20 |  |  |  |  | cis | 2020 |  |  | 203 |  |
| coin |  | Hes | ， | coide | － |  | ${ }_{935}$ |  | $\substack { \text { and } \\ \begin{subarray}{c}{0.180{ \text { and } \\ \begin{subarray} { c } { 0 . 1 8 0 } } \end{subarray}$ | coin |  |  | ， |  |  | ， |  |
| cose | cind | Moum | 20as | coicle | cois | ${ }_{\text {fras }}$ | ${ }_{\text {and }}^{2}$ |  |  |  | coiss | ¢tis |  | neoum | 038 | 58 |  |
| （tas |  |  |  | －0，40 | cosion | com |  |  | （eate |  |  | ， |  | $\substack{\text { memoun } \\ \text { muctur }}$ |  | 200 |  |
| cose |  | ， |  |  |  |  | cin |  |  | coicle | cois | cois | cise | Meoum | ${ }^{0.306}$ |  | 䞨 |
| coicle |  | ${ }_{\text {mest }}^{\text {men }}$ | \％osm |  | cis |  | cois |  |  |  | ${ }^{\text {dasion }}$ | coin | （2002 |  |  | 202 |  |
|  | $\xrightarrow{\text { natas }}$ | ${ }_{\text {mes }}^{\text {mic }}$ |  | ${ }^{\text {oseo }}$ | ${ }_{\text {a }}^{0.0}$ | ${ }_{\substack{\text { mama } \\ \text { mams }}}$ | 旡 |  |  | －0， |  | 20 |  |  |  |  |  |
| coil |  |  |  |  |  |  | cosiose |  |  |  |  |  | $\substack { \text { and } \\ \begin{subarray}{c}{220 \\ 820{ \text { and } \\ \begin{subarray} { c } { 2 2 0 \\ 8 2 0 } } \end{subarray}$ |  |  |  |  |
|  |  |  |  |  | $\underbrace{\substack{1235 \\ 0.150}}_{\substack{\text { a }}}$ | $\substack{\begin{subarray}{c}{\text { mand } \\ \text { max }} }} \end{subarray}$ | coisisi | wrem |  |  |  | （ion | cisize | mean |  |  |  |
|  | $\substack{12120 \\ 1230}$ | ${ }_{\text {med }}$ | ， | （incos |  |  | ， | wrem | cose | coiction |  | \％ | cosin | mown |  | 边 |  |
| coil | $\pm$ | lick |  |  |  | $\underset{\substack{\text { mad } \\ \text { mad }}}{ }$ | $\substack{\begin{subarray}{c}{\text { and } \\ \text { anco } \\ \text { anc }} }} \end{subarray}$ |  |  |  | 为 | smas | cosisisi | meoun | cosio | asion |  |
| coid |  |  | cosid | coiol |  | ${ }_{\text {coser }}$ |  |  |  |  | $\underset{\substack{0.027 \\ 020}}{\substack{027}}$ | \％ | cosisisi |  | Sta | 200 |  |
|  |  | $\underbrace{\substack{\text { Hemet }}}_{\text {Hest }}$ | $\underbrace{0.0}_{\substack{\text { cose } \\ 0.505 \\ 0.50}}$ |  | ${ }_{\text {cosio }}^{0.500}$ | coin |  |  |  |  |  |  | coin |  |  |  |  |
|  |  | meem | ， | 0 | ${ }_{0}^{0}$ | cis | $\xrightarrow{11250}$ | Hem |  |  |  | ¢ | （1020 |  |  |  |  |
|  |  |  |  | ， | \％osis | ma |  |  |  |  |  |  | and |  | \％ | cois |  |
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| coiction |  | ${ }_{\text {Hen }}^{\text {Hem }}$ |  |  | coicl | ${ }^{1+1}$ | $\substack{\text { sise } \\ 6550}$ | （remeiant |  |  |  | cos |  | wrater | coid |  |  |
| Oens | cos | \＃6er |  | ， |  |  | ${ }_{\substack{\text { cise } \\ \text { cise }}}$ |  |  | 隹 | cose | \％oz |  |  |  |  | 238 |
| coil |  |  |  | （0， | ， | ${ }^{102}$ |  | Meime |  |  |  |  |  |  |  |  |  |
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|  |  |  |  | － | ， | nuo |  |  |  |  |  | wri | cois | med |  |  |  |
| cot |  |  |  |  | cois | cin |  | Meile |  |  |  |  | coin | Humen |  | （ismo |  |
| coit | $\underbrace{}_{\substack{\text { cise } \\ \text { geo }}}$ | mira |  | ，0， | ，028 | Hisi |  |  |  |  |  |  | and | wrem |  |  |  |
| come | 2620 |  |  |  | \％ais | muid |  | meme |  | coice |  |  |  |  |  |  | 退 |
|  |  |  |  |  | 20， | min |  |  |  |  |  |  |  |  | 迷 |  | \％ |
|  | cise | Mroum | ${ }^{1.1226}$ | ${ }^{2124}$ | ${ }^{121}$ |  |  |  |  |  |  |  | （1208 | ныe |  |  |  |
|  | $\substack { \text { sis } \\ \begin{subarray}{c}{350{ \text { sis } \\ \begin{subarray} { c } { 3 5 0 } } \end{subarray}$ |  |  | 1220 |  | mis |  |  |  | ${ }^{1.07}$ |  |  |  | matem |  |  |  |
| ${ }^{\text {ceatem }}$ |  |  |  | ${ }_{\text {cose }}^{0.128}$ |  |  |  | tuen |  |  |  |  |  |  |  |  | 3 |
|  | cisisi |  |  |  |  | $\underset{\substack{\text { and } \\ \text { and } \\ \text { and }}}{ }$ | $\substack{\text { liatao } \\ \text { aita }}$ |  |  |  |  |  |  |  |  |  |  |
| ${ }_{\substack{\text { cis } \\ \text { sisi }}}^{1}$ | $\underbrace{\substack{\text { a }}}_{\substack{\text { cise } \\ \text { iscos }}}$ |  | 0.022 | ${ }_{\text {cose }}^{\substack{\text { cosis }}}$ |  | $\underset{\text { arin }}{\substack{\text { ariz }}}$ | $\underbrace{}_{\substack{20.10 \\ 1030}}$ |  |  | 0.50 |  | ams | cose |  | coios | （otas | （107 |
| （ism |  | $\substack{\text { moct } \\ \text { coct }}$ | coiccose | cos |  |  | coicle | $\substack{\text { mown } \\ \text { moctu}}$ |  |  | ${ }_{\text {a }}^{0}$ |  |  |  |  | coide | ${ }_{\text {a }}^{0.0055}$ |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |

Figure 6－2，Figure 6－3 and Figure 6－4 below show the regional maps of the three RCI sub－ indices．We can see an increasing heterogeneity in the performance of regions across the three pillar groups with more regions having similar scores in the Basic pillar group，as expected，with the exception of some of the newest EU Member States．Performance in the Innovation pillar group shows highest diversity across regions，suggesting the different levels of sophistication of regional economies．


Figure 6-2. Basic RCI sub-index


Figure 6-3. Efficiency RCI sub-index


Figure 6-4. Innovation RCI sub-index

For each country, the coefficient of variation ( $\mathrm{CV}=$ standard deviation/arithmetic mean) of the three sub_indices has been computed across the regions to identify the level of heterogeneity of the regions with respect to the partial RCI scores. In Figure 6-5 the absolute value of CV for the three sub_indices is displayed for each country (countries with only one region are not displayed). The graph is not displaying the CV value of the second sub-index (intermediate economy) for Finland since its value is over 40, almost four times the highest value. This is due to the fact that three out of five Finland regions, Itä-Suomi, Pohjois-Suomi and Åland, score very low (negative) values on the second sub-index; while the remaining two regions, Etelä-Suomi and Länsi-Suomi score higher values (positive ones). The case of CV of sub-index 3 of France is instead due to overseas regions (Guadeloupe, Martinique, Guyane, Reunion; FR91-FR94) which score extreme low values for innovative pillars.

Apart from these two cases, the coefficient of variation follows the same pattern across countries for all the three sub_indices, meaning that heterogeneity, or dually homogeneity, of one country does not depend on the particular group of pillars.


Figure 6-5: Absolute value of CV of the three sub_indices within each country (countries with only one region are not displayed)

Further, a parametric and non-parametric analysis of variance is used to test whether the classification of the regions into the three development stages is well reflected by the three sub_indices. To this aim an ANOVA (Knoke et al., 2002) is applied to test the difference in the means of the sub_indices using the development stage as classification variable. Table 95 shows basic descriptive statistics of the three sub_indices for each development stage. It can be seen that, as expected, mean values of all the sub_indices increase as the development stage increases. ANOVA results show that average values of the sub_indices depend on the region's development stage (ANOVA test is statistically significant for all the three sub_indices, see Table 96). Further, a non-parametric ANOVA, the Kruskall-Wallis test (Hollander and Wolfe, 1973), has been applied to the same data (Table 97). Results of the nonparametric test also support the discriminating power of the three sub_indices with respect to the development stage.

Table 95: Descriptive statistics of the sub_indices for each development stage

|  |  | N | Mean | Std. Deviation |
| :---: | :---: | :---: | :---: | :---: |
| sub_index 1 | MEDIUM | 66 | -. 76321 | . 505118 |
|  | INTERMEDIATE | 85 | -. 00748 | . 485122 |
|  | HIGH | 117 | . 40897 | . 464911 |
|  | Total | 268 | -. 01178 | . 668743 |
| sub_index 2 | MEDIUM | 66 | -. 80305 | . 483567 |
|  | INTERMEDIATE | 85 | -. 34909 | . 599923 |
|  | HIGH | 117 | . 30174 | . 429088 |
|  | Total | 268 | -. 17675 | . 675937 |
| sub_index 3 | MEDIUM | 66 | -. 94856 | . 341325 |
|  | INTERMEDIATE | 85 | -. 21920 | . 522323 |
|  | HIGH | 117 | . 40380 | . 484713 |
|  | Total | 268 | -. 12684 | . 713641 |

Table 96: Analysis of variance of the three sub_indices on the basis of the development stage
ANOVA

|  |  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: |
| sub_index 1 | Between Groups | 57.981 | 2 | 28.991 | 125.071 | .000 |
|  | Within Groups | 61.426 | 265 | .232 |  |  |
|  | Total | 119.407 | 267 |  |  |  |
| sub_index2 | Between Groups | 55.201 | 2 | 27.600 | 109.510 | .000 |
|  | Within Groups | 66.789 | 265 | .252 |  |  |
|  | Total | 121.990 | 267 |  |  |  |
| sub_index 3 | Between Groups | 78.235 | 2 | 39.118 | 179.521 | .000 |
|  | Within Groups | 57.743 | 265 | .218 |  |  |
|  | Total | 135.979 | 267 |  |  |  |

Table 97: Non parametric ANOVA of the three sub_indices on the basis of development stage
Hypothesis Test Summary

| Null Hypothesis | Test | Sig. | Decision |
| :--- | :--- | :--- | :--- |
| $\mathbf{1}$The distribution of sub_index 1 is <br> the same across categories of <br> STAGE. | Independent- <br> Samples <br> Kruskal- <br> Wallis Test | .000 | Reject the <br> null <br> hypothesis. |
| The distribution of sub_index 2 is <br> the same across categories of <br> STAGE. | Independent- <br> Samples <br> Kruskal- <br> Wallis Test | .000 | Reject the <br> null <br> hypothesis. |
| $\mathbf{3}$The distribution of sub_index 3 is <br> the same across categories of <br> STAGE. | Independent- <br> Samples <br> Kruskal- <br> Wallis Test | .000 | Reject the <br> null <br> hypothesis. |

Asymptotic significances are displayed. The significance level is .05 .

The final RCI is computed as weighted average of the three sub_indices, with weights defined on the basis of the development stage of the region. These scores are named 'weighted RCI' and are displayed in Table 99. The map of the weighted RCI at the regional level is shown in Figure 6-6. Values are normalized using the min-max transformation and classified into six classes.


Figure 6-6: Map of the weighted RCI (min-max normalized values)
For the sake of completeness, we have also computed the un-weighted RCI as simple average of the $11^{\text {th }}$ sub-scores. Figure 6-7 shows the corresponding map.


Figure 6-7: Map of the unweighted RCI (min-max normalized values)
When comparing Figure 6-6 and Figure 6-7, it can be noted that the unweighted approach leads to slightly lower scores and this is more evident for medium-stage regions - Eastern European regions, some regions in Greece, Southern Italy, Spain and Portugal - which score less with respect to the weighted RCI computation.

Figure 6-8 displays the scatter-plot of the weighted (on the x -axis) and unweighted (on the y axis) RCI scores separately for the three development stages. Trend lines are the least squares regression lines computed for the three groups. For all the regression lines the coefficient of determination $\mathrm{R}^{2}$ is very high ( $>0.90$ ) and slope coefficients are all above 0.93 , meaning that for all the three groups of countries the unweighted RCI score is slightly lower that the weighted one.


Figure 6-8: Scatterplot between weighted and unweighted RCI scores

As done for the three sub_indices, an ANOVA test of the weighted RCI score is computed with the development stage as classification variable. RCI averages are significantly different for the different development stages with increasingly higher means corresponding to increasing level of the region's development (Table 98).

Table 98: Comparison of average RCI scores across different development stages

|  |  |  |  |
| :--- | :---: | :---: | :---: |
|  | N | Mean | Std. Deviation |
| MEDIUM | 66 | -.8017 | .40467 |
| INTERMEDIATE | 85 | -.2206 | .50493 |
| HIGH | 117 | .3538 | .39763 |
| Total | 268 | -.1130 | .63653 |

ANOVA
weighted_RCI

|  | Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | ---: | ---: | ---: | ---: | :---: |
| Between Groups | 57.780 | 2 | 28.890 | 151.900 | .000 |
| Within Groups | 50.401 | 265 | .190 |  |  |
| Total | 108.181 | 267 |  |  |  |

Finally, Table 99 shows reordered regions, from best to worst, their weighted RCI score and the corresponding rank (low ranks are associated to high RCI scores). Hereafter, these ranks are referred as 'reference ranks' and the weighted RCI is simply called RCI.

Table 99: RCI scores and ranks

| reordered regions (best to worst) | weighted RCI | reference rank | reordered regions (best to worst) | weighted RCI | reference rank | reordered regions (best to worst) | weighted RCI | reference rank |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| NL31 | 1.253 | 1 | UKK4 | 0.230 | 91 | ITE2 | -0.370 | 181 |
| DK01 | 1.130 | 2 | DEFO | 0.229 | 92 | ES11 | -0.393 | 182 |
| NL32 | 1.116 | 3 | DED2 | 0.227 | 93 | czo7 | -0.406 | 183 |
| UKI | 1.082 | 4 | UKE3 | 0.216 | 94 | ITD2 | -0.413 | 184 |
| SE11 | 1.081 | 5 | ITC4 | 0.211 | 95 | PT16 | -0.432 | 185 |
| F118 | 1.031 | 6 | SE21 | 0.208 | 96 | Es41 | -0.446 | 186 |
| NL33 | 1.024 | 7 | DE42 | 0.199 | 97 | PL51 | -0.448 | 187 |
| FR10 | 1.017 | 8 | De73 | 0.181 | 98 | ES13 | -0.451 | 188 |
| NL41 | 0.993 | 9 | DED3 | 0.180 | 99 | ITF1 | -0.451 | 189 |
| UKJ1 | 0.954 | 10 | FR42 | 0.179 | 100 | ES61 | -0.460 | 190 |
| DE21 | 0.876 | 11 | DE24 | 0.179 | 101 | ITD1 | -0.478 | 191 |
| UKJ2 | 0.871 | 12 | Deb1 | 0.167 | 102 | ES12 | -0.482 | 192 |
| NL22 | 0.835 | 13 | ES51 | 0.155 | 103 | CZ04 | -0.491 | 193 |
| Uкк1 | 0.759 | 14 | FR82 | 0.152 | 104 | PT11 | -0.493 | 194 |
| De71 | 0.758 | 15 | DECO | 0.151 | 105 | PL11 | -0.495 | 195 |
| NL42 | 0.752 | 16 | UKC2 | 0.141 | 106 | ES62 | -0.495 | 196 |
| Be00 | 0.729 | 17 | DE22 | 0.140 | 107 | czo8 | -0.503 | 197 |
| UKH2 | 0.711 | 18 | DEB2 | 0.138 | 108 | PL41 | -0.511 | 198 |
| AT13 | 0.700 | 19 | DEGO | 0.138 | 109 | ITF3 | -0.530 | 199 |
| DE60 | 0.687 | 20 | AT12 | 0.128 | 110 | LToO | -0.538 | 200 |
| NL21 | 0.682 | 21 | FR52 | 0.112 | 111 | PL63 | -0.543 | 201 |
| UKJ3 | 0.678 | 22 | ES21 | 0.106 | 112 | ES23 | -0.560 | 202 |
| BE21 | 0.658 | 23 | DE93 | 0.097 | 113 | BG41 | -0.562 | 203 |
| DE11 | 0.635 | 24 | DE94 | 0.097 | 114 | PL52 | -0.568 | 204 |
| DE12 | 0.633 | 25 | FR62 | 0.096 | 115 | Es53 | -0.609 | 205 |
| SE23 | 0.630 | 26 | UKNO | 0.092 | 116 | ES42 | -0.621 | 206 |
| DEA2 | 0.627 | 27 | AT21 | 0.083 | 117 | HU21 | -0.628 | 207 |
| NL11 | 0.623 | 28 | SE33 | 0.082 | 118 | PL32 | -0.652 | 208 |
| DK04 | 0.614 | 29 | DED1 | 0.080 | 119 | PL42 | -0.654 | 209 |
| DK02 | 0.608 | 30 | be33 | 0.079 | 120 | HU22 | -0.658 | 210 |
| Lu00 | 0.600 | 31 | ITD5 | 0.060 | 121 | ITF4 | -0.668 | 211 |
| SE22 | 0.593 | 32 | UKL1 | 0.056 | 122 | ITC2 | -0.674 | 212 |
| deal | 0.585 | 33 | AT34 | 0.049 | 123 | ITG1 | -0.676 | 213 |
| BE23 | 0.578 | 34 | SE31 | 0.048 | 124 | PL31 | -0.679 | 214 |
| Dко3 | 0.572 | 35 | UKE1 | 0.035 | 125 | PL33 | -0.684 | 215 |
| czo1 | 0.567 | 36 | FR51 | 0.035 | 126 | Lvoo | -0.700 | 216 |
| UKM2 | 0.565 | 37 | DeE0 | 0.032 | 127 | SK03 | -0.700 | 217 |
| NL23 | 0.564 | 38 | F120 | 0.032 | 128 | PL43 | -0.718 | 218 |
| UKD2 | 0.550 | 39 | $1 E 01$ | 0.031 | 129 | PL61 | -0.726 | 219 |
| UKH1 | 0.530 | 40 | AT11 | 0.021 | 130 | Es70 | -0.742 | 220 |
| F19 | 0.528 | 41 | UKC1 | 0.015 | 131 | PT18 | -0.756 | 221 |
| SE12 | 0.515 | 42 | FR30 | 0.007 | 132 | ITF6 | -0.772 | 222 |
| $1 E 02$ | 0.512 | 43 | ITE4 | 0.006 | 133 | мтоо | -0.775 | 223 |
| DE30 | 0.506 | 44 | DE41 | 0.004 | 134 | GR12 | -0.783 | 224 |
| NL34 | 0.496 | 45 | DE80 | 0.003 | 135 | ITF2 | -0.788 | 225 |
| DE25 | 0.484 | 46 | 5101 | 0.003 | 136 | ES43 | -0.815 | 226 |
| UKE2 | 0.480 | 47 | FR24 | -0.018 | 137 | PL34 | -0.823 | 227 |
| DE13 | 0.472 | 48 | SE32 | -0.025 | 138 | SK04 | -0.829 | 228 |
| DE14 | 0.461 | 49 | FR41 | -0.027 | 139 | FR83 | -0.849 | 229 |
| DK05 | 0.454 | 50 | FR22 | -0.035 | 140 | PL62 | -0.866 | 230 |
| UKH3 | 0.447 | 51 | BE35 | -0.043 | 141 | HU33 | -0.874 | 231 |
| UKF2 | 0.434 | 52 | BE32 | -0.049 | 142 | HU31 | -0.905 | 232 |
| UKD3 | 0.430 | 53 | PT17 | -0.050 | 143 | PT15 | -0.906 | 233 |
| UKG1 | 0.429 | 54 | HU10 | -0.057 | 144 | ITG2 | -0.915 | 234 |
| BE25 | 0.428 | 55 | FR23 | -0.058 | 145 | ITF5 | -0.918 | 235 |
| Es30 | 0.427 | 56 | ITD3 | -0.067 | 146 | HU23 | -0.923 | 236 |
| UKJ4 | 0.417 | 57 | PL12 | -0.070 | 147 | HU32 | -0.937 | 237 |
| Deb3 | 0.410 | 58 | FR61 | -0.081 | 148 | GR14 | -1.026 | 238 |
| NL12 | 0.392 | 59 | ITC1 | -0.084 | 149 | FR92 | -1.049 | 239 |
| UKM5 | 0.386 | 60 | UKM6 | -0.091 | 150 | GR23 | -1.103 | 240 |
| UKF1 | 0.373 | 61 | UKD1 | -0.092 | 151 | GR24 | -1.115 | 241 |
| UKE4 | 0.366 | 62 | FR81 | -0.114 | 152 | GR43 | -1.135 | 242 |
| SK01 | 0.366 | 63 | FR72 | -0.146 | 153 | BG42 | -1.144 | 243 |
| dea3 | 0.365 | 64 | GR30 | -0.152 | 154 | RO11 | -1.146 | 244 |
| FR71 | 0.360 | 65 | ITE1 | -0.154 | 155 | GR25 | -1.172 | 245 |
| AT31 | 0.357 | 66 | ES22 | -0.156 | 156 | FR94 | -1.173 | 246 |
| UKK2 | 0.353 | 67 | FR26 | -0.158 | 157 | GR11 | -1.178 | 247 |
| DE26 | 0.349 | 68 | UKF3 | -0.170 | 158 | RO42 | -1.193 | 248 |
| NL13 | 0.346 | 69 | FR21 | -0.176 | 159 | RO31 | -1.197 | 249 |
| UKG3 | 0.345 | 70 | fR53 | -0.176 | 160 | Ртз0 | -1.202 | 250 |
| UKL2 | 0.333 | 71 | FR43 | -0.177 | 161 | FR91 | -1.219 | 251 |
| DE92 | 0.331 | 72 | EEOO | -0.178 | 162 | GR13 | -1.233 | 252 |
| F13 | 0.324 | 73 | FR25 | -0.198 | 163 | RO21 | -1.260 | 253 |
| UKG2 | 0.322 | 74 | czo3 | -0.212 | 164 | BG32 | -1.275 | 254 |
| DE72 | 0.313 | 75 | ES52 | -0.217 | 165 | BG34 | -1.291 | 255 |
| BE22 | 0.312 | 76 | Cz06 | -0.221 | 166 | BG33 | -1.294 | 256 |
| DE23 | 0.307 | 77 | BE34 | -0.225 | 167 | RO12 | -1.294 | 257 |
| deas | 0.307 | 78 | PL22 | -0.230 | 168 | GR21 | -1.311 | 258 |
| DE27 | 0.304 | 79 | czo2 | -0.238 | 169 | RO41 | -1.369 | 259 |
| F11A | 0.300 | 80 | ITC3 | -0.255 | 170 | GR42 | -1.376 | 260 |
| Uкм3 | 0.291 | 81 | czos | -0.261 | 171 | RO22 | -1.385 | 261 |
| DE50 | 0.285 | 82 | ITD4 | -0.275 | 172 | BG31 | -1.387 | 262 |
| Атз3 | 0.280 | 83 | Uккз | -0.281 | 173 | GR22 | -1.465 | 263 |
| AT32 | 0.275 | 84 | FR63 | -0.291 | 174 | ES63 | -1.483 | 264 |
| UKD4 | 0.273 | 85 | cYoo | -0.298 | 175 | PT20 | -1.485 | 265 |
| dea4 | 0.266 | 86 | PL21 | -0.325 | 176 | GR41 | -1.511 | 266 |
| AT22 | 0.256 | 87 | RO32 | -0.339 | 177 | ES64 | -1.597 | 267 |
| 5102 | 0.248 | 88 | Es24 | -0.356 | 178 | FR93 | -1.750 | 268 |
| UKD5 | 0.231 | 89 | SK02 | -0.361 | 179 |  |  |  |
| DE91 | 0.230 | 90 | ITE3 | -0.362 | 180 |  |  |  |

### 6.2 Country competitiveness scores - CCI

An indicator of competitiveness on the country level has been computed as a population weighted average of the regional competitiveness scores RCI of each country. Table 100 shows the individual country scores (a) and the country ranking (b), while Figure 6-9 shows the country score map.

Table 100: Competitiveness scores at the country level
a) Country competitiveness index

| conuntry-code | CCI | Min_max <br> normalized CCI |
| :---: | :---: | :---: |
| BE | 0.416 | 76 |
| BG | -1.072 | 5 |
| CZ | -0.223 | 46 |
| DK | 0.742 | 92 |
| DE | 0.391 | 75 |
| EE | -0.178 | 48 |
| IE | 0.383 | 75 |
| GR | -0.743 | 20 |
| ES | -0.214 | 46 |
| FR | 0.169 | 65 |
| IT | -0.250 | 44 |
| CY | -0.298 | 42 |
| LV | -0.700 | 23 |
| LT | -0.538 | 30 |
| LU | 0.600 | 85 |
| HU | -0.612 | 27 |
| MT | -0.775 | 19 |
| NL | 0.904 | 100 |
| AT | 0.312 | 71 |
| PL | -0.468 | 34 |
| PT | -0.437 | 35 |
| RO | -1.167 | 0 |
| SI | 0.116 | 62 |
| SK | -0.501 | 32 |
| FI | 0.721 | 91 |
| SE | 0.552 | 83 |
| UK | 0.488 | 80 |
|  |  |  |

b) Country Competitiveness Index ranking

| CCI ranking |  |
| :---: | :---: |
| 1 | NL |
| 2 | DK |
| 3 | FI |
| 4 | LU |
| 5 | SE |
| 6 | UK |
| 7 | BE |
| 8 | DE |
| 9 | IE |
| 10 | AT |
| 11 | FR |
| 12 | SI |
| 13 | EE |
| 14 | ES |
| 15 | CZ |
| 16 | IT |
| 17 | CY |
| 18 | PT |
| 19 | PL |
| 20 | SK |
| 21 | LT |
| 22 | HU |
| 23 | LV |
| 24 | GR |
| 25 | MT |
| 26 | BG |
| 27 | RO |



Figure 6-9: Country Competitiveness Index map (min-max normalized values as shown in Table 100)

In Table 101 we compare the country ranking from the Country Competitiveness Index with the one of the 2009/2010 edition ${ }^{17}$ of the Global Competitiveness Index, the world's reference index for country competitiveness and the source of the framework structure for the computation of the RCI. As expected the differences are not large. For eight countries the shift in rank is higher or equal to four, with only Luxembourg which moves upward six positions with respect to WEF-GCI (i.e, we rank Luxembourg better than WEF). These differences may be easily explained by the fact that, even if the framework of the two composites is similar, data sources, geographical level and the method followed for the construction of the RCI score are substantially different in the two cases.

[^15]Table 101: Comparison between CCI 2010 and GCI 2009-2010

| country-code | CCI rank | GCI 2009-2010 rank | diff |
| :---: | :---: | :---: | :---: |
| NL | 1 | 5 | -4 |
| DK | 2 | 2 | 0 |
| FI | 3 | 3 | 0 |
| LU | 4 | 10 | -6 |
| SE | 5 | 1 | 4 |
| UK | 6 | 6 | 0 |
| BE | 7 | 9 | -2 |
| DE | 8 | 4 | 4 |
| IE | 9 | 11 | -2 |
| AT | 10 | 8 | 2 |
| FR | 11 | 16 | 4 |
| SI | 12 | 15 | -2 |
| EE | 13 | 13 | 1 |
| ES | 14 | 12 | 3 |
| CZ | 15 | 20 | -4 |
| IT | 16 | 14 | 3 |
| CY | 17 | 17 | 1 |
| PT | 18 | 18 | 1 |
| PL | 19 | 19 | 1 |
| SK | 20 | 22 | -1 |
| LT | 21 | 23 | -1 |
| HU | 22 | 25 | -2 |
| LV | 23 | 26 | -2 |
| GR | 24 | 21 | 4 |
| MT | 25 | 27 | -1 |
| BG | 26 | 24 | 3 |
| RO | 27 |  |  |

Figure 6-10 provides a clear picture of the countries whose rank mostly deviates from the WEF-GCI rank.


Figure 6-10: CCI 2010 - GCI 2009-2010 rank difference

### 6.3 Robustness analysis of the RCI

As always in composite indicator analysis, the setting up of the final index is based upon a series of choices. Some of them may be subjective, at least to some extent, or driven by mathematical simplicity or experts' opinion. The aim of the robustness analysis is to examine the extent to which the final ranking depends on the set of choices made and the analysis typically involves the simultaneous variation of the set of the uncertain parameters in a preselected interval.

The framework of a composite is usually assumed to be fixed as its choice is mainly driven by socio-economic aspects and experts' opinion. The indicators which populate the pillars in the framework are generally chosen by integrating experts' judgment, data availability and checks on statistical consistency, as in the RCI case. Transformation and normalization methods may be also checked via uncertainty analysis. For RCI the adopted transformations have been fully justified by a detailed univariate analysis carried out indicator by indicator (section 5). The aggregation and weighting scheme is another important source of uncertainty in CIs. In the case of RCI, the choice of simple average aggregation at the pillar level has been verified and supported case by case by multivariate statistical analyses (Section 5). Thus, other choices have been considered uncertain and checked by means of an uncertain analysis -UA - (OECD, 2008; Saltelli et al. 2008) detailed in the following.

In the RCI construction the uncertain analysis is carried out considering the following parameters:

- the second threshold for the computation of the development stage $-\mathrm{t}_{2}$;
- the set of weights assigned to the three groups of pillars of the medium, intermediate and high stages of development $-\mathrm{w}_{\mathrm{M} 1}, \mathrm{w}_{\mathrm{N} 2}, \mathrm{w}_{\mathrm{M} 3}, \mathrm{w}_{\mathrm{I} 1}, \mathrm{w}_{\mathrm{I} 2}, \mathrm{w}_{\mathrm{I} 3}, \mathrm{w}_{\mathrm{H} 1}, \mathrm{w}_{\mathrm{H} 2}, \mathrm{w}_{\mathrm{H} 3} ;$
with a the total number of runs of 1200 , each corresponding to a different set of parameter values. Each run can be viewed as a particular scenario for the RCI computation.

Parameter $t_{2}$ is simply sampled from the continuous uniform distribution $U[95,105]$ centered in the reference value, $\mathrm{t}_{2 \text { ref }}=100$.

Parameters $\mathrm{w}_{\mathrm{i}} \mathrm{s}$ are instead limited by the constraint:

$$
\sum_{\mathrm{i}=1}^{3} \mathrm{w}_{\mathrm{ji}}=1 \quad \mathrm{j}=\mathrm{M}, \mathrm{I}, \mathrm{H}
$$

The sampling strategy for $\mathrm{w}_{\mathrm{i}} \mathrm{s}$ is slightly more complicated. First, the initial distribution of each parameter is assumed to be a continuous uniform distribution centered in the corresponding reference value (reference values are displayed in Figure 6-1). The choice of the range of uncertainty was driven by to two opposite needs: on the one hand, there is the need to anticipate the criticism that the assumptions of the uncertainty analysis are not 'wide enough'; on the other hand, there is the need to not completely spoil the weighting structure of the RCI, which would make the classification of regions into different development stages pointless. Following this trade off the distributions assigned to the set of weights of RCI are shown in Table 102 and sketched in Figure 6-11.

Table 102: range of variation assigned to weights $w_{i}$

| Parameter | Reference value | Range of variability |
| :---: | :---: | :---: |
| $\mathrm{w}_{\mathrm{M} 1}$ | 0.4 | $\mathrm{U}[0.3,0.5]$ |
| $\mathrm{w}_{\mathrm{M} 2}$ | 0.5 | $\mathrm{U}[0.4,0.6]$ |
| $\mathrm{w}_{\mathrm{M} 3}$ | 0.1 | $\mathrm{U}[0.05,0.15]$ |
| $\mathrm{w}_{\mathrm{I} 1}$ | 0.3 | $\mathrm{U}[0.2,0.4]$ |
| $\mathrm{w}_{\mathrm{I} 2}$ | 0.5 | $\mathrm{U}[0.4,0.6]$ |
| $\mathrm{w}_{\mathrm{I} 3}$ | 0.2 | $\mathrm{U}[0.1,0.3]$ |
| $\mathrm{w}_{\mathrm{H} 1}$ | 0.2 | $\mathrm{U}[0.1,0.3]$ |
| $\mathrm{w}_{\mathrm{H} 2}$ | 0.5 | $\mathrm{U}[0.4,0.6]$ |
| $\mathrm{w}_{\mathrm{H} 3}$ | 0.3 | $\mathrm{U}[0.2,0.4]$ |



Figure 6-11: Sketch of uncertainty ranges assigned to the RCI set of weights

Due to weight constraints, values of $\mathrm{w}_{\mathrm{i}} \mathrm{s}$ cannot be independently sampled from these distributions. Instead, a check is added in order to end up with a consistent set of weights for each development stage and a weight permutation is performed to balance the sample. For this reason the final distributions of weights is no more perfectly uniform, but has some 'very' low and high values as shown in Figure 6-12, Figure 6-13 and Figure 6-14.


Figure 6-12: Final distributions of weights for the MEDIUM development stage (1200 runs)


Figure 6-13: Final distributions of weights for the INTERMEDIATE development stage (1200 runs)


Figure 6-14: Final distributions of weights for the HIGH development stage (1200 runs)

UA results are displayed in Figure 6-15. For each region, it shows the boxplot of rank differences $R D$, i.e. the difference between the rank corresponding to the modified scenario and the reference rank. Vertical lines which cross the boxes represent all the 1200 values of rank difference computed for the region, actually showing the whole distribution of $R D$. Two horizontal lines at the values -30 and +30 have been added to the figure to show a tolerance interval of about $\pm 10 \%$ of shift of RD . At a first glance, it can be seen that the ranking is rather robust. For only 9 regions out of 268 (about $2 \%$ of the cases) $R D$ values go outside the $[-30 ;+30]$ band. They are listed in Figure 6-15 next to the picture.


Regions which show highest variation:

HU10 154 Közép-Magyarország
SK01 216 Bratislavský kraj
FI13 220 Itä-Suomi
FI19 222 Länsi-Suomi
FI1A 223 Pohjois-Suomi
FI20 224 Åland
SE21 227 Småland med öarna
SE31 230 Norra Mellansverige
SE32 231 Mellersta Norrland

Figure 6-15: Boxplots of ranking differences and regions with shifts higher than $\pm 30$ positions

Results shown in Table 103 are the frequencies of each region rank calculated over all the 1200 simulated scenarios. The higher the score the lower the rank (that is best performers are assigned the lowest ranks).

Frequency distributions are classified into 27 classes (1-|10, 11-|20, ... 261-|268). Such frequency matrix has a twofold aim: to show most and least stable regions while providing a synthesized picture of the region ranking. The most stable regions, with frequencies higher or equal to $95 \%$ in one interval, are highlighted in blue. 'Volatile' regions are considered as those regions whose rank values spam at least four rank intervals. They are highlighted in orange. The top elements in the matrix correspond to the regions with very stable and high score on competitiveness. Within this group, the ones which are always in the top ten for each simulated scenario are:

NL31 168 Utrecht-The Netherlands
DK01 24 Hovedstaden-Denmark
NL32 169 Noord-Holland-The Netherlands
UKI00 253 Inner London + Outer London-United Kingdom
SE11 225 Stockholm-Sweden

FI18 221 Etelä-Suomi-Finland
NL33 170 Zuid-Holland- The Netherlands

At the other end of the RCI classification one finds:
PT20 204 Região Autónoma dos Açores
GR41 81 Voreio Aigaio-Greece
ES64 101 Ciudad Autónoma de Melilla-Spain
FR93 127 Guyane-France
These regions are really low performers as they rank among the worst ten for all the 1200 different choices of RCI parameters.

Table 103: Frequency matrix of the regions rank for the RCI (low ranks correspond to high RCI values)

| Region | 1-110 | 11-120 | 21-\|30 | 31-140 | 41-150 | 51-160 | 61-170 | 71-\|80 | 81-190 | 91-100 | 101-\|110 | 111-\|120 | 121-130 | 131-\|140 | 141-\|150 | 151-\|160 | 161-1170 | 171-\|180 | 181-\|190 | 191-\|200 | 201-210 | 211-\|220 | 221-1230 | 231-240 | 241-250 | 251-260 | 261-268 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| NL31 | 100 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DK01 | 100 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL32 | 100 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UK100 | 98.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| SE11 | 100 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| F118 | 100 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL33 | 99.25 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| FR10 | 75.167 | 24.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL41 | 77 | 23 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKJ1 | 44.25 | 55.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE21 | 35.917 | 64 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKJ2 | 25.333 | 74.667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL22 | 33.167 | 66.167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKK1 |  | 99.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE71 | 10.75 | 71.667 | 16.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL42 |  | 100 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| BE00 |  | 99.5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKH2 |  | 82.75 | 17.25 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| AT13 |  | 65 | 33.333 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE60 |  | 50.083 | 48.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL21 |  | 6.1667 | 93.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKJ3 |  | 14.333 | 82.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| BE21 |  | 35.167 | 41.167 | 22.333 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE11 |  | 23.333 | 62 | 13.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE12 |  |  | 99.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| SE23 |  |  | 94.833 | 5.1667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DEA2 |  |  | 93.167 | 6.5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL11 |  | 29.417 | 31.75 | 23.75 | 11.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DK04 |  |  | 64.75 | 34.667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DK02 |  |  | 57.583 | 42.417 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| LU00 |  |  | 38.667 | 50.25 | 10.667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| SE22 |  |  | 23.083 | 75.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DEA1 |  |  | 33.833 | 35.333 | 26.917 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| BE23 |  |  | 6.8333 | 89.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DK03 |  |  |  | 99.667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| Cz01 |  |  | 6.4167 | 83.583 | 10 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKM2 |  |  | 8.1667 | 69.667 | 22.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL23 |  |  |  | 56.667 | 43.333 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKD2 |  |  | 12.583 | 57.917 | 27.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKH1 |  |  |  | 42 | 53.417 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| F119 |  |  |  |  | 94.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| SE12 |  |  |  | 31.833 | 56.167 | 11.25 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| IE02 |  |  |  |  | 97.917 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE30 |  |  |  | 24.167 | 51.083 | 24.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL34 |  |  |  | 7.3333 | 82.417 | 10.25 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE25 |  |  |  | 18.833 | 50.833 | 29.417 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKE2 |  | 5.0833 | 13.5 | 18.5 | 18.417 | 20.5 | 10.917 | 8.5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE13 |  |  |  | 5.5833 | 53.75 | 39.417 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE14 |  |  |  | 27.25 | 28.167 | 34.583 | 6.5833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DK05 |  |  |  | 19.5 | 32.75 | 38 | 5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKH3 |  |  |  |  | 76.583 | 23.417 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKF2 |  |  |  |  | 43.917 | 45.75 | 5.9167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKD3 |  |  |  | 5.6667 | 29.25 | 51.417 | 11.667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKG1 |  |  |  |  | 10 | 90 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| BE25 |  |  |  |  | 6 | 91.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| ES30 |  |  |  |  | 20.833 | 40.917 | 21.5 | 11.917 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKJ4 |  | 6 | 9 | 13.583 | 21.167 | 16.5 | 10.667 | 11.833 | 8.25 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DEB3 |  |  |  |  |  | 57.917 | 34.25 | 5.9167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL12 |  |  |  |  |  | 77.25 | 22.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKM5 |  |  |  |  |  | 58.25 | 41.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKF1 |  |  |  |  |  | 46.75 | 30.25 | 19.25 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKE4 |  |  |  |  |  | 29.917 | 69.167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| SK01 |  |  |  |  |  | 22.75 | 70.333 | 6.9167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DEA3 |  |  |  |  |  | 30.833 | 37.75 | 27.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| FR71 |  |  |  |  |  |  | 91.5 | 6.8333 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| AT31 |  |  |  |  |  | 12.083 | 63.417 | 24.5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKK2 |  |  |  |  |  | 13.167 | 53.083 | 33.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE26 |  |  |  |  |  |  | 76.75 | 23.167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| NL13 |  |  |  |  |  |  | 60.75 | 38 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKG3 |  |  |  |  |  |  | 46.833 | 50.667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKL2 |  |  |  |  |  |  | 40 | 45.583 | 9.5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE92 |  |  |  |  |  | 18.667 | 27.833 | 30.417 | 19.417 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| F13 |  |  |  |  |  |  | 26.667 | 66.5 | 6.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKG2 |  |  |  |  |  |  | 15.583 | 70.917 | 12.75 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE72 |  |  |  |  |  | 7.8333 | 25.667 | 32.667 | 28.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| BE22 |  |  |  |  |  |  | 22.75 | 48.833 | 26.833 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE23 |  |  |  |  |  | 19.167 | 16.667 | 17.917 | 25.25 | 10.667 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DEA5 |  |  |  |  |  |  | 6.6667 | 72.75 | 20.167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE27 |  |  |  |  |  |  | ${ }^{5} 10.167$ | ${ }^{73.25}$ | ${ }^{21.75}$ |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKM3 |  |  |  |  |  |  |  | 53.75 | 43.333 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE50 |  |  |  |  |  |  |  | 63.917 | 36.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| AT33 |  |  |  |  |  |  |  | 47.583 | 48.333 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| AT32 |  |  |  |  |  | 6.8333 | 17.917 | 21.167 | 29.75 | 19.167 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKD4 |  |  |  |  |  |  |  | 8.9167 | 59.583 | 26.5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DEA4 |  |  |  |  |  |  |  |  | 87.333 | 12.417 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| AT22 <br> S102 |  |  |  |  |  |  |  |  | $\frac{84.5}{48}$ | 15.5 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| UKD5 |  |  |  |  |  |  |  |  | 598.833 | ${ }^{48.583}$ |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| DE91 |  |  |  |  |  |  |  |  | 45.333 | 54.083 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |




We next present the median performance of the regions with the $90 \%$ confidence interval computed across all the 1200 scenarios for each region (Figure 6-16). Regions are reordered from best to worst performers according to their median rank (in red). Error bars represent the $5^{\text {th }}$ and $95^{\text {th }}$ percentiles of the rank distribution for each region. Regions for which the width of the estimated $90 \%$ confidence interval (computed as difference between $95 \%$ and $5 \%$ percentiles across 1200 simulations) is higher than 30 , meaning an oscillation of the region rank of thirty positions wide, are highlighted in the Figure. Overall only eight regions belong to this class. The analysis of the picture
highlights, in agreement with all other UA results, that the RCI is rather robust and stable with respect to the selected sources of uncertainties. The narrow confidence interval estimated for all the regions suggests that there are no hotspots in the graph, in terms of volatile ranks. The difference between the median rank and the reference rank (computed with all parameters set to their reference value) goes from a minimum of -7 to a maximum of +3 .


Figure 6-16: Median and 90\% confidence intervals (across 1200 simulations) for the RCI ranks (displayed regions are those for which the estimated $\mathbf{9 0 \%} \mathbf{C I}$ is higher than 30 positions wide)

Finally, the distribution of the shift in rank for all the countries and all the simulations is shown in Figure 6-17. It provides an overall glance of the RCI robustness with respect to the sources of uncertainty under investigation and shows a clear pick around zero. A closer look at the distribution highlights that in more than $80 \%$ of the cases the shift in rank is at most of 5 positions.


Figure 6-17: Histogram of the overall shift in ranks.

## The effect of discarding one pillar at a time

To evaluate the balance among the pillars included in the RCI framework, it is interesting to quantify the effect of discarding one pillar at a time on final scores. To this aim, all the uncertain parameters are set back to their reference values and we compute regional scores discarding one pillar at a time. Eleven simulations are run each discarding one pillar at a time.


Figure 6-18: Effect of discarding one pillar at a time on RCI reference ranks

Figure 6-18 summarizes results of this analysis. Boxplots refer to the different simulations discarding one pillar at a time and display the interquartile range of the distribution of the difference between the modified rank, obtained without one pillar, and the reference rank, computed on the basis of the reference RCI score (see Table 99). Vertical lines show the
entire range of variation of the rank difference distribution for each simulation. All the interquartile ranges are between the band -10 and +10 , meaning that, for all the simulations, $75 \%$ of the times the maximum shift of the region rank is up to 10 positions wide. This indicates a very balanced role of the pillars. The most influencing pillars are Higher Education/ Training and Lifelong Learning, Labor Market Efficiency and Market Size. These results are strictly related to the fact that these three pillars are featuring medium-stage economies which are assigned, on average across the three development stages, the highest weights (see Figure 6-1).

## Compensability effects at a glance

As most composite indicators, RCI is an aggregation of several indicators describing related but different factors. In this kinD of setting the aggregation always implies taking a position on the key issue of compensability. 'Compensability' is here understood as the:
existence of trade-off, i.e. the possibility of offsetting a disadvantage on some criteria by a sufficiently large advantage on another criterion (Munda, 2008, pg. 71).

RCI has the mathematical form of a linear aggregation. It intrinsically entails compensability at all its computational levels: from the 'micro' level of sub-sub-pillars to the 'macro' level of sub-indices (basic, efficiency and innovation).

RCI is then affected by compensability, but to what extent? Various approaches may be used to assess the level of compensability of composite indicators, most of them are based on fully compensatory or fully non compensatory multi-criteria methods (see Munda, 2008 for a review). Our approach is here to provide a quick glance of compensability issues by means of the Ordered Weighted Averaging (OWA), originally proposed by Yager (Yager, 1988 and 1996). The OWA method consists of a family of operators which, for any given object (country, region, individual, ....), map a set of (k) real values $\left\{x_{1}, x_{2}, \ldots . ., x_{k}\right\}$, indicators observed for that object, into a single index depending on a set of weights $\left\{w_{1}, w_{2}, \ldots . ., w_{k}\right\}$ :

$$
f_{\mathrm{owA}}\left(x_{1}, x_{2}, \ldots, x_{k}\right)=\sum_{i=1}^{k} w_{i} x_{(i)} \quad w_{i} \in[0,1] \quad \sum_{i=1}^{k} w_{i}=1
$$

where $x_{(i)}$ is the $i$-th largest $x_{i}$, that is $\left\{x_{(1)}, x_{(2)}, \ldots . ., x_{(k)}\right\}$ is the series of $x_{i}$ values reordered in descending order. Operators $f_{\text {owA }}$ are not a weighted average since the set of weights depends only on the $i$-th ordered position without considering the original set of indicators. The interesting feature of the OWA operators is that they embed many different types of aggregations depending on the set of weights $w_{i}$. If the need is to emphasize higher (lower) values of $x_{i}$ 's then the first weights should be assigned higher (lower) values. A number of special cases can be defined for the OWA operators. Among these, the following three have a special role:
a. Purely optimistic operator: $\quad f_{\text {owA }}^{o}\left(x_{1}, x_{2}, \ldots, x_{k}\right)=\sum_{i=1}^{k} w_{i}^{o} x_{(i)} \quad w_{i}^{o}=\{1,0, \ldots ., 0\}$
b. Purely pessimistic operator: $f_{\mathrm{owA}}^{\mathrm{P}}\left(x_{1}, x_{2}, \ldots, x_{k}\right)=\sum_{i=1}^{k} w_{i}^{P} x_{(i)} \quad w_{i}^{P}=\{0,0, \ldots, 1\}$
c. Average operator:

$$
f_{\mathrm{owA}}^{\mathrm{A}}\left(x_{1}, x_{2}, \ldots, x_{k}\right)=\sum_{i=1}^{k} w_{i}^{A} x_{(i)} \quad w_{i}^{A}=\left\{\frac{1}{k}, \frac{1}{k}, \ldots ., \frac{1}{k}\right\}
$$

The optimistic operator $f^{0}{ }_{\text {OWA }}$ includes in the computation only the highest value of the $x_{i}$ thus meaning full compensability among indicators. This is implicitly equivalent to an 'or' multiple criteria condition, where the satisfaction of at least one criterion is enough.

On the other hand, the pessimistic operator $f^{P}{ }_{\text {owA }}$ takes into account only the lowest value of the indicators, thus meaning no compensation at all across indicators. The worst case is taken as representative and this is equivalent to an 'and condition: all criteria must be satisfied.
in many cases the type of aggregation operator lies somewhere between these two extremes, as the $f^{A}{ }_{\text {owA }}$ operator which is the simple arithmetic mean with equal weights.

In the case of RCI the three scenarios are computed at the sub_index level: for each pillar group and for each region the corresponding sub_index is computed using both $f^{0}{ }_{\text {owA }}$ and $f^{P}{ }_{\text {owa }}$ (the average OWA is equal to the sub_indices shown in Table 94, Section 6.1). These values are then compared to the reference RCI score, computed with the set of weights at their reference value according to the region development stage.

Figure 6-19 shows the reference RCI value, computed using average OWA operator for all the three sub_indices, and the 'optimistic' (blue line) and 'pessimistic' (red line) RCI scores. As expected, the two lines are always located respectively above and below the reference line, with the space between the two slightly increasing going form left to right of the picture (that is from best to worst regions). The important piece of information that can be deduced from this figure is the range of variability of each region. Indeed, regions with very low pessimistic RCI scores are also those with very high optimistic RCI scores. These regions (highlighted in Figure 6-19) are mostly influenced by compensability effects so that a change in the weighting scheme highly affects their final score. Their wide range of variability, associated to the different OWA operators, indicates high levels of heterogeneity of the subscores across each pillar group. In total about 15 regions seem to have a high range of variation. Further, as the distance between the average trend of the blue and the red line tends to increase going from left to right, low performing regions are more affected by compensability issues than the others.

Overall, given that the two OWA operators $f^{0}{ }_{\text {OwA }}$ and $f^{p}{ }_{\text {owa }}$ are at the extreme ends of the aggregation decision-making process, OWA results can be considered rather satisfactory.


Figure 6-19: RCI scores computed with OWA operators

In conclusion, the uncertainty analysis detailed in this section supports the robustness of RCI. The index provides a synthetic picture of the level of competitiveness of Europe at the NUTS2 level representing a well balanced plurality of different fundamental aspects.

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## Appendix A - Literature Review

Table A_1: Detailed list of variables included in the GCI (Schwab and Porter, 2007, pp. 269-270)



Table A_2: Detailed list of variables included in the WCY (IMD World Competitiveness Yearbook, 2008, pp. 294, 336, 376, 416)

## Pillar 1. Economic Performance

## I.I Domestic Economy

Size
1.1.01 Gross domestic product (GDP)
1.1.02 GDP (PPP)
I.I. 03 Household consumption expenditure
I.I. 04 Household consumption expenditure
I.I. 05 Government consumption expenditure
I.I.06 Government consumption expenditure
1.1.07 Gross fixed capital formation
I.I.08 Gross fixed capital formation
1.1.09 Gross domestic savings
I.I.10 Gross domestic savings
I.I.II Economic sectors
1.1.12 Diversification of the economy

Growth
I.I.I3 Real GDP Growth
I.I.I4 Real GDP Growth per capita
I.I.15 Household consumption expenditure - real growth
I.I.I6 Government consumption expenditure - real growth
I.I.I7 Gross fixed capital formation - real growth
I.I.I8 Resilience of the economy

Wealth
1.1.19 GDP per capita
1.1.20 GDP (PPP) per capita
1.I.21 Household consumption expenditure per capita
1.1.22 Government consumption expenditure per capita
1.1.23 Gross fixed capital formation per capita
1.I. 24 Gross domestic savings per capita

## Forecasts

I.I. 25 Forecast: Real GDP Growth
I.I. 26 Forecast: Inflation
I.1.27 Forecast: Unemployment
I.I. 28 Forecast: Current account balance

### 1.2 International Trade

1.2.01 Current account balance
1.2.02 Current account balance
1.2.03 Balance of trade
1.2.04 Balance of trade
I.2.05 Balance of commercial services
1.2.06 Balance of commercial services
1.2.07 Exports of goods
1.2.08 Exports of goods
1.2.14 Imports of goods \& commercial services
1.2.15 Imports of goods \& commercial services
1.2.16 Imports of goods \& commercial services - growth
1.2.17 Imports breakdown by economic sector
1.2.18 Trade to GDP ratio
1.2.19 Terms of trade index
1.2.20 Tourism receipts
1.2.21 Exchange rates

## I. 3 International Investment Investment <br> I.3.01 Direct investment flows abroad <br> 1.3.02 Direct investment flows abroad <br> 1.3.03 Direct investment stocks abroad <br> I.3.04 Direct investment stocks abroad <br> 1.3.05 Direct investment stocks abroad - growth <br> I.3.06 Direct investment flows inward <br> 1.3.07 Direct investment flows inward <br> 1.3.08 Direct investment stocks inward <br> 1.3.09 Direct investment stocks inward <br> 1.3.10 Direct investment stocks inward - growth <br> 1.3.1I Balance of direct investment flows <br> 1.3.12 Balance of direct investment flows <br> 1.3.13 Net position in direct investment stocks <br> 1.3.14 Net position in direct investment stocks <br> 1.3.15 Relocation threats of production <br> 1.3.16 Relocation threats of R\&D facilities <br> 1.3.17 Relocation threats of services <br> Finance <br> 1.3.18 Portfolio investment assets . <br> 1.3.19 Portfolio investment liabilities

### 1.4 Employment

1.4.01 Employment
1.4.02 Employment
1.4.03 Employment - growth
1.4.04 Employment by sector
1.4.05 Employment in the public sector
1.4.06 Unemployment rate
1.4.07 Long-term unemployment
1.4.08 Youth unemployment
1.2.09 Exports of goods - growth
1.2.10 Exports of commercial services
1.2.11 Exports of commercial services
1.2.12 Exports of commercial services - growth
1.2.13 Exports breakdown by economic sector
1.5 Prices
1.5.01 Consumer price inflation
1.5.02 Cost-of-living index
I.5.03 Apartment rent
1.5.04 Office rent

## Pillar 2. Government efficiency

## 2.I Public Finance

2.1.01 Government budget surplus/deficit
2.1.02 Government budget surplus/deficit
2.I.03 Total general government debt
2.1.04 Total general government debt
2.1.05 Total general government debt - real growth
2.1.06 Central government domestic debt
2.1.07 Central government foreign debt
2.1.08 Interest payment
2.1.09 Management of public finances
2.1.10 Pension funding
2.1.11 Total reserves
2.1.12 General government expenditure

### 2.2 Fiscal Policy

2.2.01 Collected total tax revenues
2.2.02 Collected personal income tax
2.2.03 Collected corporate taxes
2.2.04 Collected indirect tax revenues
2.2.05 Collected capital and property taxes
2.2.06 Collected social security contribution
2.2.07 Effective personal income tax rate
2.2.08 Corporate tax rate on profit
2.2.09 Consumption tax rate
2.2.10 Employee's social security contribution rate
2.2.11 Employer's social security contribution rate
2.2.12 Real personal taxes
2.2.13 Real corporate taxes
2.2.14 Tax evasion

### 2.3 Institutional Framework

 Central Bank2.3.01 Real short-term interest rate
2.3.02 Cost of capital
2.3.03 Interest rate spread
2.3.04 Country credit rating
2.3.05 Central bank policy
2.3.06 Exchange rate stability

State Efficiency
2.3.07 Policy direction of the government
2.3.08 Legal and regulatory framework
2.3.09 Adaptability of government policy
2.3.10 Government decisions
2.3.11 Political parties
2.3.12 Transparency
2.3.13 Public service
2.3.14 Bureaucracy
2.3.15 Bribing and corruption

### 2.4 Business Legislation

## Openness

2.4.01 Customs' authorities
2.4.02 Protectionism
2.4.03 Public sector contracts
2.4.04 International transactions
2.4.05 Foreign investors
2.4.06 Capital markets
2.4.07 Investment incentives

Competition and Regulations
2.4.08 Government subsidies
2.4.09 Subsidies
2.4.10 State ownership of enterprises
2.4.II Competition legislation
2.4. 12 Product and service legislation
2.4.13 Price controls
2.4.14 Parallel economy
2.4.15 Regulation intensity
2.4.16 Ease of doing business
2.4.17 Creation of firms
2.4.18 Start-up days

Labor Regulations
2.4.19 Labor regulations
2.4.20 Unemployment legislation
2.4.21 Immigration laws

### 2.5 Societal Framework

2.5.01 Justice
2.5.02 Personal security and private property
2.5.03 Ageing of society
2.5.04 Risk of political instability
2.5.05 Social cohesion
2.5.06 Income distribution - lowest 20\%
2.5.07 Income distribution - highest 20\%
2.5.08 Equal opportunity
2.5.09 Females in parliament
2.5.IO Female positions
2.5. II Gender income ratio

## Pillar 3. Business efficiency

### 3.1 Productivity and Efficiency

3.1.01 Overall productivity (PPP)
3.1.02 Overall productivity
3.I.03 Overall productivity - real growth
3.1.04 Labor productivity (PPP)
3.1.05 Labor productivity (PPP) growth
3.1.06 Agricultural productivity (PPP)
3.1.07 Productivity in industry (PPP)
3.1.08 Productivity in services (PPP)
3.1.09 Large corporations
3.1.10 Small and medium-size enterprises
3.I.II Productivity of companies

### 3.2 Labor Market

Costs
3.2.01 Compensation levels
3.2.02 Unit labor costs in the manufacturing sector
3.2.03 Remuneration in services professions
3.2.04 Remuneration of management

## Relations

3.2.05 Working hours
3.2.06 Labor relations
3.2.07 Worker motivation
3.2.08 Industrial disputes
3.2.09 Employee training

Availability of Skills
3.2.10 Labor force
3.2.1I Labor force
3.2.12 Labor force growth
3.2.13 Part-time employment
3.2.14 Female labor force
3.2.15 Foreign labor force
3.2.16 Skilled labor
3.2. 17 Finance skills
3.2.18 Attracting and retaining talents
3.2.19 Brain drain
3.2.20 Foreign high-skilled people
3.2.21 International experience
3.2.22 Competent senior managers

### 3.3 Finance

## Bank Efficiency

3.3.0I Banking sector assets
3.3.02 Credit
3.3.03 Financial cards in circulation
3.3.04 Financial card transactions
3.3.05 Investment risk
3.3.06 Venture capital
3.3.07 Banking and financial services
3.3.08 Retail banking
3.3.09 Banking regulation
3.3.10 Financial risk factor

## Stock Market Efficiency

3.3.1| Stock markets
3.3.12 Stock market capitalization
3.3.13 Stock market capitalization
3.3.14 Value traded on stock markets
3.3.15 Listed domestic companies
3.3.16 Stock market index
3.3.17 Shareholders' rights
3.3.18 Financial institutions' transparency

Finance Management
3.3.19 Cash flow
3.3.20 Corporate debt
3.4 Management Practices
3.4.01 Adaptability of companies
3.4.02 Ethical practices
3.4.03 Credibility of managers
3.4.04 Corporate boards
3.4.05 Auditing and accounting practices
3.4.06 Shareholder value
3.4.07 Customer satisfaction
3.4.08 Entrepreneurship
3.4.09 Social responsibility
3.4. 10 Health, safety \& environmental concerns

### 3.5 Attitudes and Values

3.5.01 Attitudes toward globalization
3.5.02 Image abroad
3.5.03 National culture
3.5.04 Flexibility and adaptability
3.5.05 Need for economic and social reforms
3.5.06 Value system
3.5.07 Corporate values

## Pillar 4. Infrastructure

4.1 Basic Infrastructure<br>4.1.01 Land area<br>4. 1.02 Arable area<br>4.1.03 Access to water<br>4. I.04 Access to commodities<br>4.1. 05 Urbanization<br>4. 1.06 Population - market size<br>4. 1.07 Population under 15 years<br>4.I. 08 Population over 65 years<br>4. 1.09 Dependency ratio<br>4.1.10 Roads<br>4.1.II Railroads<br>4.I.I2 Air transportation<br>4.I.I3 Quality of air transportation<br>4.1.14 Distribution infrastructure<br>4.1.15 Water transportation<br>4.I.16 Maintenance and development<br>4.1.17 Energy infrastructure<br>4.1.18 Future energy supply<br>4.I.19 Total indigenous energy production<br>4.I. 20 Total indigenous energy production per capita<br>4.1.21 Total indigenous energy production<br>4.1.22 Total final energy consumption<br>4.1.23 Total final energy consumption per capita<br>4.1.24 Electricity costs for industrial clients<br>4.2 Technological Infrastructure<br>4.2.01 Investment in telecommunications<br>4.2.02 Fixed telephone lines<br>4.2.03 International fixed telephone costs<br>4.2.04 Mobile telephone subscribers<br>4.2.05 Mobile telephone costs<br>4.2.06 Communications technology<br>4.2.07 Computers in use<br>4.2.08 Computers per capita<br>4.2.09 Internet users<br>4.2.10 Internet costs<br>4.2.11 Broadband subscribers<br>4.2.12 Broadband costs<br>4.2.13 Information technology skills<br>4.2.14 Technological cooperation<br>4.2.15 Public and private sector ventures<br>4.2.16 Development and application of technology<br>4.2.17 Funding for technological development<br>4.2.18 Technological regulation<br>4.2.19 High-tech exports<br>4.2.20 High-tech exports<br>4.2.21 Cyber security<br>4.3 Scientific Infrastructure<br>4.3.01 Total expenditure on R\&D<br>4.3.02 Total expenditure on R\&D<br>4.3.03 Total expenditure on R\&D per capita<br>4.3.04 Business expenditure on R\&D<br>4.3.05 Business expenditure on R\&D<br>4.3.06 Total R\&D personnel nationwide<br>4.3.07 Total R\&D personnel nationwide per capita<br>4.3.08 Total R\&D personnel in business enterprise 4.3.09 Total R\&D personnel in business per capita

## Table A_3: Data sources (Huggins and Davies, 2006, pp. 36-37)

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Data Sources
Europe
Eurostat - nttp://epp.eurostat ec.europa.eu
EU Regional Pollcy - nttp://ec.europa_eu/regional_pollcy/ndex_en.ntm
Pax15 - http://www.cordl5.lu/paxis/src/
OECD - nttp://oecd.org/statistics
InternatIonal Reports - nttp///www. Internationalreports.net/europe/
Innovating Reglons in Europe - nttp:/www.Innovating-reglons.org/
UK Trade and investment Country Profles - nttp://www.tradepariners.gov.uk
Buslness Patrol Country Reports - nttp://www.businesspatrol.comv
Auatria
Austrla Buslness Agency - nttp://www.aba.gv.at/en/pages/
Belglum
Fedichem - nttp//www.fedichem.be/EN/homeen.htm
Invest in Belglum - http:/Investinbelglum.fgov.be/
nttp://www.Invest.belglum.be/
Agency for Forelgn Trade - http://www.obcebdoh.be/menu/menule.html?ver-en
The Belglum Portal for Research and Innovation - http://www.research.be/
Brussels Enterprise Agency - www.Investinbrussels.com
FInland
Invest in Finland - nttp://www.Investinfiland.f/
Assoclation of Blotecinology Industry in Finland - http://www.finblo.net
Vitual Finland - nttp://virtual.finland.fi/
Finnish Reglonal Counclis - nttp://www.reg.f/engllishengindex.ntml
Finnish Bloindustries - http//www.finblonet/nome.htm
Otanleml Sclence Park Website - nttp://www.otech.n/Veng/
Uusimaa Regional Councll - nttp///www.uudenmaanilitto.N/eng/
Councll of Oulu Region - http//wwww.ponjols-ponjanmaa.f///Ne/kehys.htm
Clty of Helsinkl - nt:p//www.hel.n
France
Tourlst website of Parls-lle de France Reglon - nttp://www.parls-lle-de-france.com/AN-worid/home/home asp?pays-UK
Germany
Elan - http://www.elanit.conv/mmedlacyimain.asp?page-1918
Invest In Germany - nt:p//www.Invest-In-germany.de/en/
Invest in Baden-Wurtemburg - www.bw-Investcom
Invest in Bavarla - www.Invest-In-bavaria.com
Ministy of State of Baden-Worttemberg - http://www.baden-wuerttemberg.de/english/IIvestleren/index.ntm|
The Hessen Website - htp://www.Invest-In-hessen.de/she/Index_Js.htm
Ireland
IDA Ireland - nt:p://www./daireland.com/home/Index.asp
Italy
Invest in Italy - hitp://www.Investinitaly.comv
Invest in Emilla-Romagna - nt:p://www.livestinemillaromagna.it/nvest_Inglese/ndex_eng.ntm
Unlon Camere Emilla-Romagna - ntip//www.rer.camcom.t/welcome/defaulthtm
Invest in South Italy - http://www.Italydat!.It/asp/default.asp
Itallan Blotech Directory - nttp//www.blodirectory.It/index.ntml
Luxembourg
Statistic Omce - nttp://www.statec.lu/html_en/ndex.html
Luxembourg Tourist omice in London - nttp://www.luxembourg.co.uk/
Luxembourg Board of Economic Development, Ministry of the Economy - nttp//www.bed.public.lu
Norway
Noway in the UK - nttp://www.norway.org.uk
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Sweden
Invest In Sweden Agency - www.lsa.se
Gateway to Sweden - ntp://www.sweden.5e/
swedish instrute - www.sl.se
Swlss Representations in Australla - nttp://www.eda.admin.ch/australla_alle/home.html
ntlp:/www.ltsweden.com
www.blosweden.org
Swedlish Sclence and Technology Park - www.swedepark.se
The Swedish Research Councll - nttp///www.v.se/english/hdex.asp
Education and Research in Sweden - ntip//ulblidning.regeringen.se/inenglish/educresearch/research.htm
Stockhoim County Councll - nttp//www.5ll.se/nternational/defaut.asp
Cliy of Stockhoim - www.stockholm.se
Buslness Arena Stockholm - http://www.bas.stockholm.se/
TIME-Stockholm - nttp://www:Ime.stockhoim.se/eng/ndex.html
Karollnska Instltutet - wWW.kl.se
Klsta Sclence Park - http://www.klsta.com/
Switzerland
Sclence and Innovation in Switzerland - http///www.5w/s5-sclence.org/
Location Switzeriand - nttp;/www.standortschwelz ch/seco/nternet/er/index.htm
United KIngdom
UK Trade and Investment - nttp://www.Invest.uk.com
Department of Trade and Industry - www.dtl.gov.uk
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## Appendix B - Indicator on the strength of regional clusters

We are proposing to use the 'European Cluster Observatory' database for measuring the development level of clusters at the regional level.

The European Cluster Observatory (www.clusterobservatory.eu) (ECO) is a new platform for information on clusters produced by the Center for Strategy and Competitiveness at the Stockholm School of Economics and financed by DG Enterprise and Industry, under the Europe INNOVA programme.

The Cluster Mapping part of the ECO considers regions, at the spatial level, and sectors, at the industrial level. By combining the two dimensions of geography and industry, it statistically traces regional agglomerations of employment ${ }^{18}$, defined as statistical regional clusters, across EU 27 at the NUTS 2 level. Exceptions are Belgium, Greece, and the Netherlands, where the NUTS 1 level is used in order to obtain comparability in terms of land area and employment, and for Ireland due to data availability.

The database evaluates the strength of regional clusters based on three criteria - size, focus and specialization, and consequently assigns each cluster from one (less strong) to three stars (very strong).

The rationale behind the size measure implies that if employment reaches a sufficient share in proportion to total European employment, it is more likely that meaningful economic effects will be present. Thus, if a cluster is in the top $10 \%$ of all clusters within the same cluster category in terms of the number of employees, it is evaluated as strong and receives a star.

The specialization measure compares the proportion of employment in a cluster category in a region over the total employment in the same region, to the proportion of total European employment in that cluster category over total European employment. If a cluster category receives a quotient of 2 or more, it is evaluated as strong and receives a star. The rationale is that if a region is more specialized in a specific cluster category than the overall economy across all regions, this is likely to be an indication that the economic effects of the regional

[^16]cluster have been strong enough to attract related economic activity from other regions to this location, and consequently spillovers and linkages will be stronger.

The focus measure shows the extent to which the regional economy is focused on the industries comprising the cluster category. It relates employment in the cluster to total employment in the region. The top $10 \%$ of clusters which account for the largest proportion of their region's total employment are evaluated as strong and receive a star

The general logic of the database is that the higher the number of stars, the larger and more specialized the regional cluster.

Below is an example of the information available from the Observatory:

## European Cluster Observatory



In order to evaluate the state of cluster development in a NUTS 2 region, we propose to use two indicators - the number of clusters and the relative strength (measured as the number of starts given to a cluster). Thus, we imply a relation where regions with more regional clusters and higher strength (given by the median number of stars per region) imply higher competitiveness.

Data limitation have led to the use of employment data for identifying and evaluating clusters in the ECO database, creating a bias towards employment-intensive clusters as both the size and focus measures are sensitive to the size of employment. Thus, in order to account for the importance of and emphasize the role of technology and knowledgeintensive clusters for the innovative capacity of regions and their competitiveness, we suggest complementing the overall evaluation of cluster development within a region (measured as the total number of clusters and their relative strength) by adding the number and relative strength of technology and knowledge-intensive clusters only.

Out of the 38 cluster categories ${ }^{19}$ used by the European Cluster Observatory, we have identified the following 14 as being technology and knowledge-intensive ${ }^{20}$ cluster categories:

- Aerospace
- Analytical Instruments
- Automotive
- Business Services
- Chemical Products
- Communications Equipment
- $\quad$ Education and Knowledge Creation
- Heavy Machinery
- $\quad$ Financial Services
- Information Technology
- Medical Devices
- Biopharmaceuticals
- Power Generation and Transmission
- Production Technology

Thus, we are proposing to consider four sub_indicators (number of clusters and their relative strength for all cluster categories and number of clusters and their relative strength for knowledge-intensive category) to be aggregated into a single indicator for the overall measure for the level of cluster development in regions and included in the Business sophistication pillar (Sect. 3.10, main text).

[^17]
## Appendix C - List of candidate indicators

| pillar |  | indicator <br> id | Indicators | source | geographical level | unit of measurement | periodicity | reference year taken | Notes | included (I)/ discarded (D) | reason for discarding |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Institutions | 1 | 1.1 | Corruption is a major problem in (OUR COUNTRY) | Special Eurobarometer 325 | country | survey data - \% of respondents | one time 2009 | 2009 |  | 1 |  |
| Institutions | 1 | 1.2 | There is corruption in regional institutions in (OUR COUNTRY) | Special Eurobarometer 325 | country | survey data - \% of respondents | one time 2009 | 2009 |  | 1 |  |
| Institutions | 1 | 1.3 | Perceived extent to which the state budget is defrauded | Flash Eurobarometer 2008 | country | survey data - \% of respondents | one time 2008 | 2008 |  | 1 |  |
| Institutions | 1 | 1.4 | Perceived extent of corruption or other wrongdoing in the national government institutions | Flash Eurobarometer 2008 | country | survey data - \% of respondents | one time 2008 | 2008 |  | 1 |  |
| Institutions | 1 | 1.5 | Voice and accountability | Worldbank Worldwide Governance Indicators | country | $\begin{array}{\|c\|} \text { score ranging from -2.5 to } \\ 2.5 \& \% \text { rank (0-100) } \end{array}$ | yearly | 2008 |  | 1 |  |
| Institutions | 1 | 1.6 | Political stability | Worldbank Worldwide Governance Indicators | country | $\left\|\begin{array}{c} \text { score ranging from -2.5 to } \\ 2.5 \& \% \text { rank }(0-100) \end{array}\right\|$ | yearly | 2008 |  | 1 |  |
| Institutions | 1 | 1.7 | Government effectiveness | Worldbank Worldwide Governance Indicators | country | $\left\lvert\, \begin{gathered} \text { score ranging from -2.5 to } \\ 2.5 \& \% \text { rank (0-100) } \end{gathered}\right.$ | yearly | 2008 |  | 1 |  |
| Institutions | 1 | 1.8 | Regulatory quality | Worldbank Worldwide Governance Indicators | country | score ranging from - -2.5 to 2.5 \& \% rank (0-100) | yearly | 2008 |  | 1 |  |
| Institutions | 1 | 1.9 | Rule of law | Worldbank Worldwide Governance Indicators | country | score ranging from - -2.5 to 2.5 \& \% rank (0-100) | yearly | 2008 |  | 1 |  |
| Institutions | 1 | 1.10 | Control of corruption | Worldbank Worldwide Governance Indicators | country | score ranging from - 2.5 to $2.5 \& \%$ rank (0-100) | yearly | 2008 |  | 1 |  |
| Institutions | 1 | 1.11 | Easy of doing business | Worldbank | country | rank out of 181 (better express as percentage out of 181) | yearly | $\begin{gathered} \text { June 2008-May } \\ 2009 \end{gathered}$ |  | 1 |  |
| Macroeconomic stability | 2 | 2.1 | General government deficit/surplus | Eurostat | country | \% of GDP | yearly | $\begin{gathered} \text { average 2006- } \\ 2008 \end{gathered}$ |  | 1 |  |
| Macroeconomic stability | 2 | 2.2 | Income, saving and net lending/borrowing | Eurostat | country | \% of GDP | yearly | $\begin{gathered} \text { average 2006- } \\ 2008 \end{gathered}$ |  | 1 |  |
| Macroeconomic stability | 2 | 2.3 | Inflation | Eurostat | country | \% annual change | yearly | $\begin{gathered} \text { average 2006- } \\ 2008 \end{gathered}$ |  | 1 |  |


| Macroeconomic stability | 2 | 2.4 | Long-term bond yields | Eurostat | country | annual average rate of change | yearly | average 2006- $2008$ | 1 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Macroeconomic stability | 2 | 2.5 | Government gross debt | Eurostat | country | \% of GDP | yearly | average 2006- $2008$ | D | multivariate analysis |
| Infrastructure | 3 | 3.1 | Motorway density | Eurostat/DG <br> TREN/EuroGeographics/National Statistical Institutes | NUTS2 | combined index (average pop/area), $\mathrm{EU27}=100$ | yearly | 2006 | 1 |  |
| Infrastructure | 3 | 3.2 | Railway density | Eurostat/DG <br> TREN/EuroGeographics/National Statistical Institutes | NUTS2 | combined index (average pop/area), EU27=100 | yearly | 2007 | 1 |  |
| Infrastructure | 3 | 3.3 | Number of passenger flights (accessible within 90' drive) | Eurostat/EuroGeographics/Natio nal Statistical Institutes | NUTS2 | daily no. of passenger flights | yearly (from 2006 onwards) | 2007 | 1 |  |
| Health | 4 | 4.1 | Hospital beds | Eurostat Regional Health Statistics | NUTS2 | number of hospital beds $/ 100,000$ inhabitants | yearly | 2007 | D | multivariate analysis |
| Health | 4 | 4.2 | Road fatalities | Eurostat, CARE, ITF, NSIs, DG Regional Policy | NUTS2 | number of deaths in road accidents per million inhabitants | yearly | $\begin{array}{\|c} \hline \text { average 2004- } \\ 2006 \end{array}$ | 1 |  |
| Health | 4 | 4.3 | Healthy life expectancy | Eurostat, DG Regional Policy | NUTS2 | number of years of healthy life expected | yearly | 2007 | 1 |  |
| Health | 4 | 4.4 | Infant mortality | Eurostat Regional Health Statistics | NUTS2 | number of deaths of children under 1 year of age during the year to the number of live births in that year | yearly | 2007 | 1 |  |
| Health | 4 | 4.5 | Cancer disease death rate | DG Regional Policy | NUTS2 | standardized cancer death rate for population under 65 (neoplasm C00-D48) | yearly | $\begin{gathered} \text { average 2006- } \\ 2008 \end{gathered}$ | 1 |  |
| Health | 4 | 4.6 | Heart disease death rate | Eurostat, DG Regional Policy | NUTS2 | standardized heart diseases death rate for population under 65 (diseases of the circulatory system I00-I99) | yearly | $\begin{gathered} \text { average 2006- } \\ 2008 \end{gathered}$ | 1 |  |
| Health | 4 | 4.7 | Suicide death rate | Eurostat, DG Regional Policy | NUTS2 | standardized death rate for suicide for population under 65 (intentional selfharm X60-X84) | yearly | $\begin{aligned} & \text { average 2006- } \\ & 2008 \end{aligned}$ | 1 |  |
| Quality of primary \& secondary education | 5 | 5.1 | Share of low-achieving 15 years olds in reading | OECD Programme for International Student Assessment (PISA) | country | $\%$ of students with reading proficiency level 1 or below | every three years | 2006 | 1 |  |
| Quality of primary \& secondary education | 5 | 5.2 | Share of low-achieving 15 years olds in math | OECD Programme for International Student Assessment (PISA) | country | \% of students with math proficiency level 1 or below | every three years | 2006 | 1 |  |
| Quality of primary \& secondary education | 5 | 5.3 | Share of low-achieving 15 years olds in science | OECD Programme for International Student Assessment (PISA) | country | \% of students with science proficiency level 1 or below | every three years | 2006 | 1 |  |


| Quality of primary \& secondary education | 5 | 5.4 | Teacher/pupil ratio | Eurostat Educational Statistics | country | ratio of students to teachers (ISCED 1-3) | yearly | 2007 |  | D | multivariate analysis |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Quality of primary \& secondary education | 5 | 5.5 | Financial aid to students ISCED 1-4 | Eurostat Educational Statistics | country | $\%$ of total public expenditure on education | yearly | 2006 |  | D | multivariate analysis |
| Quality of primary \& secondary education | 5 | 5.6 | Public expenditure ISCED 1 | Eurostat Educational Statistics | country | \% of GDP | yearly | 2006 |  | D | multivariate analysis |
| Quality of primary \& secondary education | 5 | 5.7 | Public expenditure ISCED 2-4 | Eurostat Educational Statistics | country | $\%$ of GDP | yearly | 2006 |  | D | multivariate analysis |
| Quality of primary \& secondary education | 5 | 5.8 | Participation in early childhood education | Eurostat Educational Statistics | country | \% of pupils between 4-years-olds and starting of compulsory primary | yearly | 2007 |  | D | multivariate analysis |
| Higher education \& training | 6 | 6.1 | Population aged 25-64 with higher educational attainment (ISCED 5-6) | Eurostat (LFS) | NUTS2 | $\%$ of total population of age group | yearly | 2007 |  | 1 |  |
| Higher education \& training | 6 | 6.2 | Lifelong learning | Eurostat Regional Education Statistics | NUTS 2 | $\%$ of population aged 25 64 participating in | yearly | 2007 |  | 1 |  |
| Higher education \& training | 6 | 6.3 | Early school leavers | Eurostat Structural Indicators | NUTS2 | $\%$ of the population aged 18-24 having attained at most lower secondary school and not going further | yearly | $\begin{gathered} \text { average } \\ \text { 2006/2007 } \end{gathered}$ |  | 1 |  |
| Higher education \& training | 6 | 6.4 | Accessibility to universities | Nordregio, EuroGeographics, GISCO, EEA ETC-TE | NUTS2 | $\left\lvert\, \begin{gathered}\% \text { of regional population at } \\ \text { more than } 60 \text { minutes } \\ \text { from the nearest university }\end{gathered}\right.$ | one time 2006 | 2006 |  | 1 |  |
| Higher education \& training | 6 | 6.5 | Higher education expenditure | Eurostat Educational Statistics | country | total public expenditure as $\%$ of GDP at levels ISCED 56 | yearly | 2006 | imputed at the NUTS 2 <br> level according to the imputation method described in section 4.2.1 | 1 |  |
| Labor market efficiency | 7 | 7.1 | Employment rate (excluding agriculture) | Eurostat Regional Labour Market Statistics ( LFS) | NUTS 2 | $\%$ of population 15-64 years | yearly | 2008 |  | 1 |  |
| Labor market efficiency | 7 | 7.2 | Long-term unemployment | Eurostat Regional Labour Market Statistics ( LFS) | NUTS 2 | \% of labor force unemployed for 12 months or more | yearly | 2008 |  | 1 |  |
| Labor market efficiency | 7 | 7.3 | Unemployment rate | Eurostat Regional Labour Market Statistics ( LFS) | NUTS 2 | \% of active population | yearly | 2008 |  | 1 |  |
| Labor market efficiency | 7 | 7.4 | Job mobility | Eurostat Regional Labour Market Statistics ( LFS) | NUTS 2 | \% of total employment (people who started to work for the current employer or as selfemployed in the last 2 years) | yearly | 2007 |  | D | multivariate analysis |
| Labor market efficiency | 7 | 7.5 | Labor productivity | Eurostat Regional Labour Market Statistics ( LFS) | NUTS 2 | GDP/person employed in industry and services ( $€$ ), Index, EU27 = 100 | yearly | 2007 |  | 1 |  |


| Labor market efficiency | 7 | 7.6 | Gender balance unemployment | Eurostat, DG Regional Policy | NUTS 2 | \% difference between female and male unemployed | yearly | 2008 |  | 1 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Labor market efficiency | 7 | 7.7 | Gender balance employment | Eurostat, DG Regional Policy | NUTS 2 | $\%$ difference between female and male unemployed | yearly | 2008 |  | 1 |  |
| Labor market efficiency | 7 | 7.8 | Female unemployment | Eurostat Regional Labour Market Statistics ( LFS) | NUTS 2 | \% of female unemployed | yearly | 2008 |  | 1 |  |
| Labor market efficiency | 7 | 7.9 | Labor market policies | Eurostat Labor Market Policy Statistics | country | $\%$ of GDP spent on public expenditure on labor market policies | yearly | 2007 | imputed at the NUTS 2 level according to the imputation method described in section 4.2.1 | D | multivariate analysis |
| Market size | 8 | 8.1 | GDP | Eurostat Regional Economic Accounts | NUTS2 | PPS index (EU27=100) | yearly | 2007 |  | 1 |  |
| Market size | 8 | 8.2 | Compensation of employees | Eurostat Regional Economic Accounts | NUTS2 | millions of euro | yearly | 2006 |  | 1 |  |
| Market size | 8 | 8.3 | Disposable income | Eurostat, DG Regional Policy estimates | NUTS2 | net adjusted disposable household income in millions of ppcs | yearly | 2006 |  | 1 |  |
| Market size | 8 | 8.4 | Potential market size expressed in GDP | Eurostat, DG Regional Policy estimates | NUTS2 | index GDP (pps) EU27=100 | yearly | 2007 |  | 1 |  |
| Market size | 8 | 8.5 | Potential market size expressed in population | Eurostat, DG Regional Policy estimates | NUTS2 | index population EU27=100 | one time 2000 | 2000 |  | 1 |  |
| Technological readiness | 9 | 9.1 | Households with access to broadband | Eurostat Regional Information Statistics | NUTS2 | \% of total households | yearly | 2009 |  | 1 |  |
| Technological readiness | 9 | 9.2 | Individuals who ordered goods or services over the Internet for private use | Eurostat Regional Information Statistics | NUTS2 | \% of individuals | yearly | 2009 |  | 1 |  |
| Technological readiness | 9 | 9.3 | Household with access to internet | Eurostat Regional Information Statistics | NUTS2 | \% of total households | yearly | 2009 |  | 1 |  |
| Technological readiness | 9 | 9.4 | Enterprises use of computers | Eurostat Community Survey on ICT usage and e-commerce | country | \% of enterprises | yearly | 2009 | $\left\|\begin{array}{c} \text { regional data available for } \\ \text { some countries but not } \\ \text { all, so country values have } \\ \text { been taken instead } \end{array}\right\|$ | 1 |  |
| Technological readiness | 9 | 9.5 | Enterprises having access to Internet | Eurostat Community Survey on ICT usage and e-commerce | country | \% of enterprises | yearly | 2009 | $\begin{gathered} \text { regional data available for } \\ \text { some countries but not } \\ \text { all, so country values have } \\ \text { been taken instead } \end{gathered}$ | 1 |  |
| Technological readiness | 9 | 9.6 | Enterprises having a website or a homepage | Eurostat Community Survey on ICT usage and e-commerce | country | \% of enterprises | yearly | 2009 | regional data available for some countries but not all, so country values have been taken instead | 1 |  |



| Innovation | 11 | 11.6 | Total intramural R\&D expenditure | Eurostat Regional Science and Technology Statistics | NUTS2 | \% of GDP | yearly | 2007 | 1 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Innovation | 11 | 11.7 | Human Resources in Science and Technology (HRST) | Eurostat Regional Science and Technology Statistics | NUTS2 | \% of labour force | yearly | 2008 | 1 |
| Innovation | 11 | 11.8 | Employment in technology and knowledge-intensive | Eurostat Regional Science and Technology Statistics | NUTS2 | \% of total employment | yearly | 2008 | 1 |
| Innovation | 11 | 11.9 | High-tech inventors | OECD REGPAT | NUTS2 | number of inventors (authors of high technology EPO patent applications) per million inhabitants | yearly | $\begin{aligned} & \text { average 2005- } \\ & 2006 \end{aligned}$ | 1 |
| Innovation | 11 | 11.10 | ICT inventors | OECD REGPAT | NUTS2 | (authors of ICT EPO patent applications) per million inhabitants | yearly | average 20052006 | 1 |
| Innovation | 11 | 11.11 | Biotechnology inventors | OECD REGPAT | NUTS2 | number of inventors authors of biotechnology EPO patent applications) per million inhabitants | yearly | $\begin{gathered} \text { average 2005- } \\ 2006 \end{gathered}$ | 1 |

Appendix D - NUTS 2 region description and population size

| Region Code | Region ID | geoltime | 2004 | 2005 | 2006 | 2007 | 2008 | mean_pop_04_08 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| BE21 | 1 | Prov. Antwerpen | 1668812 | 1676858 | 1688493 | 1700570 | 1715707 | 1690088 |
| BE22 | 2 | Prov. Limburg (B) | 805786 | 809942 | 814658 | 820272 | 826690 | 815470 |
| BE23 | 3 | Prov. Oost-Vlaanderen | 1373720 | 1380072 | 1389450 | 1398253 | 1408484 | 1389996 |
| BE25 | 4 | Prov. West-Vlaanderen | 1135802 | 1138503 | 1141866 | 1145878 | 1150487 | 1142507 |
| BE32 | 5 | Prov. Hainaut | 1283200 | 1286275 | 1290079 | 1294844 | 1300097 | 1290899 |
| BE33 | 6 | Prov. Liège | 1029605 | 1034024 | 1040297 | 1047414 | 1053722 | 1041012 |
| BE34 | 7 | Prov. Luxembourg (B) | 254120 | 256004 | 258547 | 261178 | 264084 | 258787 |
| BE35 | 8 | Prov. Namur | 452856 | 455863 | 458574 | 461983 | 465380 | 458931 |
| BE00 | 9 | Bruxelles Capital + Vlaams Brabant + Brabant Wallon | 2392520 | 2408311 | 2429418 | 2454142 | 2482215 | 2433321 |
| BG31 | 10 | Severozapaden | 991165 | 974704 | 957947 | 943664 | 929872 | 959470 |
| BG32 | 11 | Severen tsentralen | 967046 | 958755 | 949401 | 941240 | 931950 | 949678 |
| BG33 | 12 | Severoiztochen | 1005991 | 1001668 | 996831 | 993549 | 992081 | 998024 |
| BG34 | 13 | Yugoiztochen | 1146704 | 1139926 | 1134741 | 1129846 | 1125982 | 1135440 |
| BG41 | 14 | Yugozapaden | 2110036 | 2114815 | 2118855 | 2116791 | 2114568 | 2115013 |
| BG42 | 15 | Yuzhen tsentralen | 1580331 | 1571181 | 1560975 | 1554200 | 1545785 | 1562494 |
| CZ01 | 16 | Praha | 1165581 | 1170571 | 1181610 | 1188126 | 1212097 | 1183597 |
| CZ02 | 17 | Strední Cechy | 1135795 | 1144071 | 1158108 | 1175254 | 1201827 | 1163011 |
| CZ03 | 18 | Jihozápad | 1175654 | 1175330 | 1179294 | 1184543 | 1194338 | 1181832 |
| CZ04 | 19 | Severozápad | 1125117 | 1126721 | 1127447 | 1127867 | 1138629 | 1129156 |
| CZO5 | 20 | Severovýchod | 1480771 | 1480144 | 1483423 | 1488168 | 1497560 | 1486013 |
| CZ06 | 21 | Jihovýchod | 1640081 | 1640354 | 1641125 | 1644208 | 1654211 | 1643996 |
| CZ07 | 22 | Strední Morava | 1228179 | 1225832 | 1229303 | 1229733 | 1232571 | 1229124 |
| CZ08 | 23 | Moravskoslezsko | 1260277 | 1257554 | 1250769 | 1249290 | 1249897 | 1253557 |
| DK01 | 24 | Hovedstaden | : | : | : | 1636749 | 1645825 | 1641287 |
| DK02 | 25 | Sjælland | : | : | : | 816118 | 819427 | 817773 |
| DK03 | 26 | Syddanmark | : | : | : | 1189817 | 1194659 | 1192238 |
| DK04 | 27 | Midtijlland | : | : | : | 1227428 | 1237041 | 1232235 |
| DK05 | 28 | Nordjylland | : | : | : | 576972 | 578839 | 577906 |
| DE11 | 29 | Stuttgart | 3994612 | 4003172 | 4007373 | 4005380 | 4007095 | 4003526 |
| DE12 | 30 | Karlsruhe | 2722550 | 2727733 | 2732455 | 2734260 | 2739274 | 2731254 |
| DE13 | 31 | Freiburg | 2178813 | 2185027 | 2190727 | 2193178 | 2196410 | 2188831 |
| DE14 | 32 | Tübingen | 1796581 | 1801487 | 1805146 | 1805935 | 1806976 | 1803225 |
| DE21 | 33 | Oberbayern | 4195673 | 4211118 | 4238195 | 4279112 | 4313446 | 4247509 |
| DE22 | 34 | Niederbayern | 1194472 | 1196178 | 1196923 | 1193820 | 1194138 | 1195106 |
| DE23 | 35 | Oberpfalz | 1089826 | 1090289 | 1089543 | 1087939 | 1086684 | 1088856 |
| DE24 | 36 | Oberfranken | 1109674 | 1106541 | 1101390 | 1094525 | 1088845 | 1100195 |
| DE25 | 37 | Mittelfranken | 1706615 | 1708972 | 1712275 | 1712622 | 1714123 | 1710921 |
| DE26 | 38 | Unterfranken | 1344740 | 1344629 | 1341481 | 1337876 | 1334767 | 1340699 |
| DE27 | 39 | Schwaben | 1782386 | 1786166 | 1788919 | 1786764 | 1788329 | 1786513 |
| DE30 | 40 | Berlin | 3388477 | 3387828 | 3395189 | 3404037 | 3416255 | 3398357 |
| DE41 | 41 | Brandenburg - Nordost | 1167493 | 1163924 | 1159168 | 1153722 | 1147653 | 1158392 |
| DE42 | 42 | Brandenburg - Südwest | 1407028 | 1403780 | 1400315 | 1394050 | 1388084 | 1398651 |
| DE50 | 43 | Bremen | 663129 | 663213 | 663467 | 663979 | 663082 | 663374 |
| DE60 | 44 | Hamburg | 1734083 | 1734830 | 1743627 | 1754182 | 1770629 | 1747470 |
| DE71 | 45 | Darmstadt | 3762995 | 3775025 | 3778124 | 3772906 | 3780232 | 3773856 |
| DE72 | 46 | Gießen | 1065467 | 1064228 | 1061323 | 1057553 | 1053259 | 1060366 |
| DE73 | 47 | Kassel | 1260966 | 1258512 | 1252907 | 1244900 | 1239064 | 1251270 |
| DE80 | 48 | Mecklenburg-Vorpommern | 1732226 | 1719653 | 1707266 | 1693754 | 1679682 | 1706516 |
| DE91 | 49 | Braunschweig | 1662595 | 1658918 | 1650435 | 1641776 | 1633318 | 1649408 |
| DE92 | 50 | Hannover | 2167157 | 2166626 | 2163919 | 2160253 | 2156841 | 2162959 |
| DE93 | 51 | Lüneburg | 1698434 | 1702971 | 1704133 | 1702938 | 1701132 | 1701922 |
| DE94 | 52 | Weser-Ems | 2465229 | 2472394 | 2475459 | 2477718 | 2480393 | 2474239 |
| DEA1 | 53 | Düsseldorf | 5245132 | 5237855 | 5226648 | 5217129 | 5208288 | 5227010 |
| DEA2 | 54 | Köln | 4350368 | 4363797 | 4378622 | 4384669 | 4391062 | 4373704 |
| DEA3 | 55 | Münster | 2625745 | 2624489 | 2622623 | 2619372 | 2614361 | 2621318 |
| DEA4 | 56 | Detmold | 2071803 | 2072488 | 2069758 | 2065413 | 2059198 | 2067732 |
| DEA5 | 57 | Arnsberg | 3786638 | 3776723 | 3760454 | 3742162 | 3723712 | 3757938 |
| DEB1 | 58 | Koblenz | 1527919 | 1527507 | 1521494 | 1513939 | 1507919 | 1519756 |
| DEB2 | 59 | Trier | 513755 | 513861 | 513363 | 515819 | 515972 | 514554 |
| DEB3 | 60 | Rheinhessen-Pfalz | 2017008 | 2019737 | 2023986 | 2023102 | 2021752 | 2021117 |
| DECO | 61 | Saarland | 1061376 | 1056417 | 1050293 | 1043167 | 1036598 | 1049570 |
| DED1 | 62 | Chemnitz | 1568153 | 1553406 | 1537203 | 1520537 | 1503723 | 1536604 |
| DED2 | 63 | Dresden | 1674343 | 1667676 | 1662482 | 1657114 | 1646716 | 1661666 |
| DED3 | 64 | Leipzig | 1078941 | 1075202 | 1074069 | 1072123 | 1069761 | 1074019 |
| DEEO | 65 | Sachsen-Anhalt | 2522941 | 2494437 | 2469716 | 2441787 | 2412472 | 2468271 |
| DEF0 | 66 | Schleswig-Holstein | 2823171 | 2828760 | 2832950 | 2834254 | 2837373 | 2831302 |
| DEG0 | 67 | Thüringen | 2373157 | 2355280 | 2334575 | 2311140 | 2289219 | 2332674 |
| EE00 | 68 | Estonia | 1351069 | 1347510 | 1344684 | 1342409 | 1340935 | 1345321 |
| IE01 | 69 | Border, Midlands and Western | 1073820 | 1098144 | 1126474 | 1153796 | 1179280 | 1126303 |
| IE02 | 70 | Southern and Eastern | 2953912 | 3011029 | 3082545 | 3158730 | 3222055 | 3085654 |
| GR11 | 71 | Anatoliki Makedonia, Thraki | 605565 | 607847 | 607460 | 607205 | 606684 | 606952 |
| GR12 | 72 | Kentriki Makedonia | 1909297 | 1911508 | 1919401 | 1927823 | 1935660 | 1920738 |
| GR13 | 73 | Dytiki Makedonia | 294470 | 294508 | 294155 | 293864 | 293519 | 294103 |
| GR14 | 74 | Thessalia | 737340 | 737583 | 737144 | 737034 | 736079 | 737036 |
| GR21 | 75 | Ipeiros | 340854 | 341851 | 345100 | 348520 | 351786 | 345622 |
| GR22 | 76 | Ionia Nisia | 218594 | 220398 | 223149 | 225879 | 228572 | 223318 |
| GR23 | 77 | Dytiki Ellada | 730238 | 732292 | 734505 | 736899 | 738955 | 734578 |
| GR24 | 78 | Sterea Ellada | 559351 | 558503 | 557364 | 556441 | 555069 | 557346 |
| GR25 | 79 | Peloponnisos | 599199 | 598156 | 596621 | 595092 | 593378 | 596489 |
| GR30 | 80 | Attiki | 3940099 | 3973326 | 4001911 | 4032456 | 4061326 | 4001824 |
| GR41 | 81 | Voreio Aigaio | 203169 | 202402 | 201731 | 201083 | 200517 | 201780 |
| GR42 | 82 | Notio Aigaio | 302549 | 303114 | 303980 | 304975 | 305966 | 304117 |
| GR43 | 83 | Kriti | 599925 | 601263 | 602658 | 604469 | 606274 | 602918 |


| ES11 | 84 | Galicia | 2706126 | 2712162 | 2718490 | 2723915 | 2735078 | 2719154 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| ES12 | 85 | Principado de Asturias | 1060065 | 1059133 | 1058330 | 1058059 | 1059136 | 1058945 |
| ES13 | 86 | Cantabria | 545125 | 551085 | 557226 | 563611 | 570613 | 557532 |
| ES21 | 87 | Pais Vasco | 2094909 | 2103441 | 2113052 | 2124235 | 2138739 | 2114875 |
| ES22 | 88 | Comunidad Foral de Navarra | 573038 | 580616 | 588306 | 596236 | 606234 | 588886 |
| ES23 | 89 | La Rioja | 288384 | 294347 | 300821 | 306254 | 311773 | 300316 |
| ES24 | 90 | Aragón | 1228886 | 1243464 | 1258847 | 1275904 | 1297581 | 1260936 |
| ES30 | 91 | Comunidad de Madrid | 5705620 | 5821054 | 5938391 | 6052583 | 6189297 | 5941389 |
| ES41 | 92 | Castilla y León | 2462169 | 2469303 | 2477128 | 2486166 | 2501860 | 2479325 |
| ES42 | 93 | Castilla-la Mancha | 1823013 | 1856787 | 1892657 | 1929947 | 1977596 | 1896000 |
| ES43 | 94 | Extremadura | 1066149 | 1068799 | 1071339 | 1074419 | 1078908 | 1071923 |
| ES51 | 95 | Cataluña | 6637355 | 6784145 | 6936148 | 7085308 | 7238051 | 6936201 |
| ES52 | 96 | Comunidad Valenciana | 4400459 | 4518126 | 4641240 | 4759263 | 4892475 | 4642313 |
| ES53 | 97 | Illes Balears | 931831 | 957953 | 985620 | 1014405 | 1045008 | 986963 |
| ES61 | 98 | Andalucia | 7552978 | 7670365 | 7794121 | 7917397 | 8046131 | 7796198 |
| ES62 | 99 | Región de Murcia | 1265983 | 1300083 | 1335347 | 1370802 | 1411623 | 1336768 |
| ES63 | 100 | Ciudad Autónoma de Ceuta (ES) | 71456 | 71372 | 71414 | 71561 | 71989 | 71558 |
| ES64 | 101 | Ciudad Autónoma de Melilla (ES) | 66956 | 67102 | 66412 | 67556 | 69699 | 67545 |
| ES70 | 102 | Canarias (ES) | 1864840 | 1908698 | 1953361 | 1997010 | 2041468 | 1953075 |
| FR10 | 103 | Île de France | 11319972 | 11399319 | 11532398 | 11616500 | : | 11467047 |
| FR21 | 104 | Champagne-Ardenne | 1338759 | 1337672 | 1338850 | 1336000 | : | 1337820 |
| FR22 | 105 | Picardie | 1877194 | 1880890 | 1894355 | 1898000 | : | 1887610 |
| FR23 | 106 | Haute-Normandie | 1802229 | 1805955 | 1811055 | 1813000 | : | 1808060 |
| FR24 | 107 | Centre | 2487618 | 2496654 | 2519567 | 2529500 | : | 2508335 |
| FR25 | 108 | Basse-Normandie | 1442873 | 1445732 | 1456793 | 1460000 | : | 1451350 |
| FR26 | 109 | Bourgogne | 1621257 | 1622542 | 1628837 | 1630000 | : | 1625659 |
| FR30 | 110 | Nord - Pas-de-Calais | 4027031 | 4032135 | 4018644 | 4021500 | : | 4024828 |
| FR41 | 111 | Lorraine | 2331578 | 2334245 | 2335694 | 2336500 | : | 2334504 |
| FR42 | 112 | Alsace | 1794987 | 1806069 | 1815493 | 1826000 | : | 1810637 |
| FR43 | 113 | Franche-Comté | 1138410 | 1141861 | 1150624 | 1154500 | : | 1146349 |
| FR51 | 114 | Pays de la Loire | 3372044 | 3400745 | 3450329 | 3480500 | : | 3425905 |
| FR52 | 115 | Bretagne | 3037548 | 3062117 | 3094534 | 3118500 | : | 3078175 |
| FR53 | 116 | Poitou-Charentes | 1695885 | 1705347 | 1724123 | 1734000 | : | 1714839 |
| FR61 | 117 | Aquitaine | 3054252 | 3080091 | 3119778 | 3146500 | : | 3100155 |
| FR62 | 118 | Midi-Pyrénées | 2707262 | 2734954 | 2776822 | 2806000 | : | 2756260 |
| FR63 | 119 | Limousin | 722644 | 724243 | 730920 | 733000 | : | 727702 |
| FR71 | 120 | Rhône-Alpes | 5907972 | 5958320 | 6021293 | 6073500 | : | 5990271 |
| FR72 | 121 | Auvergne | 1328308 | 1331380 | 1335938 | 1339000 | : | 1333657 |
| FR81 | 122 | Languedoc-Roussillon | 2466221 | 2496871 | 2534144 | 2565000 | : | 2515559 |
| FR82 | 123 | Provence-Alpes-Côte d'Azur | 4713095 | 4750947 | 4815232 | 4855000 | : | 4783569 |
| FR83 | 124 | Corse | 274474 | 276911 | 294118 | 298500 | : | 286001 |
| FR91 | 125 | Guadeloupe (FR) | 439998 | 444002 | 436926 | 439000 | : | 439982 |
| FR92 | 126 | Martinique (FR) | 393005 | 396001 | 397732 | 400000 | : | 396685 |
| FR93 | 127 | Guyane (FR) | 193997 | 197997 | 205954 | 213500 | , | 202862 |
| FR94 | 128 | Reunion (FR) | 763204 | 774596 | 781962 | 790500 | : | 777566 |
| ITC1 | 129 | Piemonte | 4270215 | 4330172 | 4341733 | 4352828 | 4401266 | 4339243 |
| ITC2 | 130 | Valle d'Aosta/Vallée d'Aoste | 122040 | 122868 | 123978 | 124812 | 125979 | 123935 |
| ITC3 | 131 | Liguria | 1577474 | 1592309 | 1610134 | 1607878 | 1609822 | 1599523 |
| ITC4 | 132 | Lombardia | 9246796 | 9393092 | 9475202 | 9545441 | 9642406 | 9460587 |
| ITD1 | 133 | Provincia Autonoma Bolzano-Bozen | 471635 | 477067 | 482650 | 487673 | 493910 | 482587 |
| ITD2 | 134 | Provincia Autonoma Trento | 490829 | 497546 | 502478 | 507030 | 513357 | 502248 |
| ITD3 | 135 | Veneto | 4642899 | 4699950 | 4738313 | 4773554 | 4832340 | 4737411 |
| ITD4 | 136 | Friuli-Venezia Giulia | 1198187 | 1204718 | 1208278 | 1212602 | 1222061 | 1209169 |
| ITD5 | 137 | Emilia-Romagna | 4080479 | 4151369 | 4187557 | 4223264 | 4275802 | 4183694 |
| ITE1 | 138 | Toscana | 3566071 | 3598269 | 3619872 | 3638211 | 3677048 | 3619894 |
| ITE2 | 139 | Umbria | 848022 | 858938 | 867878 | 872967 | 884450 | 866451 |
| ITE3 | 140 | Marche | 1504827 | 1518780 | 1528809 | 1536098 | 1553063 | 1528315 |
| ITE4 | 141 | Lazio | 5205139 | 5269972 | 5304778 | 5493308 | 5561017 | 5366843 |
| ITF1 | 142 | Abruzzo | 1285896 | 1299272 | 1305307 | 1309797 | 1323987 | 1304852 |
| ITF2 | 143 | Molise | 321697 | 321953 | 320907 | 320074 | 320838 | 321094 |
| ITF3 | 144 | Campania | 5760353 | 5788986 | 5790929 | 5790187 | 5811390 | 5788369 |
| ITF4 | 145 | Puglia | 4040990 | 4068167 | 4071518 | 4069869 | 4076546 | 4065418 |
| ITF5 | 146 | Basilicata | 597000 | 596546 | 594086 | 591338 | 591001 | 593994 |
| ITF6 | 147 | Calabria | 2011338 | 2009268 | 2004415 | 1998052 | 2007707 | 2006156 |
| ITG1 | 148 | Sicilia | 5003262 | 5013081 | 5017212 | 5016861 | 5029683 | 5016020 |
| ITG2 | 149 | Sardegna | 1643096 | 1650052 | 1655677 | 1659443 | 1665617 | 1654777 |
| CYOO | 150 | Cyprus | 730367 | 749175 | 766414 | 778684 | 789258 | 762780 |
| LV00 | 151 | Latvia | 2319203 | 2306434 | 2294590 | 2281305 | 2270894 | 2294485 |
| LTOO | 152 | Lithuania | 3445857 | 3425324 | 3403284 | 3384879 | 3366357 | 3405140 |
| LU00 | 153 | Luxembourg (Grand-Duché) | 454960 | 461230 | 469086 | 476187 | 483799 | 469052 |
| HU10 | 154 | Közép-Magyarország | 2829704 | 2840972 | 2855670 | 2872678 | 2897317 | 2859268 |
| HU21 | 155 | Közép-Dunántúl | 1112984 | 1110897 | 1108124 | 1107453 | 1104841 | 1108860 |
| HU22 | 156 | Nyugat-Dunántúl | 1003185 | 1000348 | 1000142 | 999361 | 997939 | 1000195 |
| HU23 | 157 | Dél-Dunántúl | 983612 | 977465 | 970700 | 967677 | 960088 | 971908 |
| HU31 | 158 | Észak-Magyarország | 1280040 | 1271111 | 1261489 | 1251441 | 1236690 | 1260154 |
| HU32 | 159 | Eszak-Alföld | 1547003 | 1541818 | 1533162 | 1525317 | 1514020 | 1532264 |
| HU33 | 160 | Dél-Alföld | 1360214 | 1354938 | 1347294 | 1342231 | 1334506 | 1347837 |
| MT00 | 161 | Malta | 399867 | 402668 | 404346 | 407810 | 410290 | 404996 |
| NL11 | 162 | Groningen | 574384 | 575072 | 574042 | 573614 | 573459 | 574114 |
| NL12 | 163 | Friesland (NL) | 642066 | 642977 | 642230 | 642209 | 643189 | 642534 |
| NL13 | 164 | Drenthe | 482415 | 483369 | 484481 | 486197 | 488135 | 484919 |
| NL21 | 165 | Overijssel | 1105512 | 1109432 | 1113529 | 1116374 | 1119994 | 1112968 |
| NL22 | 166 | Gelderland | 1966929 | 1972010 | 1975704 | 1979059 | 1983869 | 1975514 |
| NL23 | 167 | Flevoland | 359904 | 365859 | 370656 | 374424 | 378688 | 369906 |
| NL31 | 168 | Utrecht | 1162258 | 1171291 | 1180039 | 1190604 | 1201350 | 1181108 |
| NL32 | 169 | Noord-Holland | 2587265 | 2599103 | 2606584 | 2613070 | 2626163 | 2606437 |
| NL33 | 170 | Zuid-Holland | 3451942 | 3458381 | 3458875 | 3455097 | 3461435 | 3457146 |
| NL34 | 171 | Zeeland | 379028 | 379978 | 380186 | 380497 | 380585 | 380055 |
| NL41 | 172 | Noord-Brabant | 2406994 | 2411359 | 2415946 | 2419042 | 2424827 | 2415634 |
| NL42 | 173 | Limburg (NL) | 1139335 | 1136695 | 1131938 | 1127805 | 1123705 | 1131896 |


| AT11 | 174 | Burgenland (A) | 276640 | 278215 | 279317 | 280257 | 281190 | 279124 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| AT12 | 175 | Niederösterreich | 1556956 | 1569596 | 1581422 | 1589580 | 1597240 | 1578959 |
| AT13 | 176 | Wien | 1598626 | 1626440 | 1651437 | 1664146 | 1677867 | 1643703 |
| AT21 | 177 | Kärnten | 559078 | 559891 | 560300 | 560407 | 561094 | 560154 |
| AT22 | 178 | Steiermark | 1192014 | 1197527 | 1202087 | 1203918 | 1205909 | 1200291 |
| AT31 | 179 | Oberösterreich | 1389170 | 1396228 | 1402050 | 1405674 | 1408165 | 1400257 |
| AT32 | 180 | Salzburg | 523185 | 526017 | 528351 | 529574 | 530576 | 527541 |
| AT33 | 181 | Tirol | 686410 | 691783 | 697435 | 700427 | 703512 | 695913 |
| AT34 | 182 | Vorarlberg | 358043 | 360827 | 363526 | 364940 | 366377 | 362743 |
| PL11 | 183 | Lódzkie | 2597094 | 2587702 | 2577465 | 2566198 | 2555898 | 2576871 |
| PL12 | 184 | Mazowieckie | 5135732 | 5145997 | 5157729 | 5171702 | 5188488 | 5159930 |
| PL21 | 185 | Malopolskie | 3252949 | 3260201 | 3266187 | 3271206 | 3279036 | 3265916 |
| PL22 | 186 | Slaskie | 4714982 | 4700771 | 4685775 | 4669137 | 4654115 | 4684956 |
| PL31 | 187 | Lubelskie | 2191172 | 2185156 | 2179611 | 2172766 | 2166213 | 2178984 |
| PL32 | 188 | Podkarpackie | 2097248 | 2097975 | 2098263 | 2097564 | 2097338 | 2097678 |
| PL33 | 189 | Swietokrzyskie | 1291598 | 1288693 | 1285007 | 1279838 | 1275550 | 1284137 |
| PL34 | 190 | Podlaskie | 1205117 | 1202425 | 1199689 | 1196101 | 1192660 | 1199198 |
| PL41 | 191 | Wielkopolskie | 3359932 | 3365283 | 3372417 | 3378502 | 3386882 | 3372603 |
| PL42 | 192 | Zachodniopomorskie | 1696073 | 1694865 | 1694178 | 1692838 | 1692271 | 1694045 |
| PL43 | 193 | Lubuskie | 1008786 | 1009168 | 1009198 | 1008520 | 1008481 | 1008831 |
| PL51 | 194 | Dolnoslaskie | 2898313 | 2893055 | 2888232 | 2882317 | 2878410 | 2888065 |
| PL52 | 195 | Opolskie | 1055667 | 1051531 | 1047407 | 1041941 | 1037088 | 1046727 |
| PL61 | 196 | Kujawsko-Pomorskie | 2068142 | 2068258 | 2068253 | 2066371 | 2066136 | 2067432 |
| PL62 | 197 | Warminsko-Mazurskie | 1428885 | 1428714 | 1428601 | 1426883 | 1426155 | 1427848 |
| PL63 | 198 | Pomorskie | 2188918 | 2194041 | 2199043 | 2203595 | 2210920 | 2199303 |
| PT11 | 199 | Norte | 3711797 | 3727310 | 3737791 | 3744341 | 3745236 | 3733295 |
| PT15 | 200 | Algarve | 405380 | 411468 | 416847 | 421528 | 426386 | 416322 |
| PT16 | 201 | Centro (PT) | 2366691 | 2376609 | 2382448 | 2385891 | 2385911 | 2379510 |
| PT17 | 202 | Lisboa | 2740237 | 2760697 | 2779097 | 2794226 | 2808414 | 2776534 |
| PT18 | 203 | Alentejo | 767549 | 767679 | 765971 | 764285 | 760933 | 765283 |
| PT20 | 204 | Região Autónoma dos Açores (PT) | 240024 | 241206 | 242241 | 243018 | 244006 | 242099 |
| PT30 | 205 | Região Autónoma da Madeira (PT) | 243007 | 244286 | 245197 | 245806 | 246689 | 244997 |
| RO11 | 206 | Nord-Vest | 2743281 | 2742676 | 2729181 | 2729256 | 2724176 | 2733714 |
| RO12 | 207 | Centru | 2543512 | 2533421 | 2534378 | 2524176 | 2524628 | 2532023 |
| RO21 | 208 | Nord-Est | 3742868 | 3735512 | 3734946 | 3727910 | 3722553 | 3732758 |
| RO22 | 209 | Sud-Est | 2855044 | 2849959 | 2843624 | 2834335 | 2825756 | 2841744 |
| RO31 | 210 | Sud - Muntenia | 3350248 | 3338195 | 3321392 | 3304840 | 3292036 | 3321342 |
| RO32 | 211 | Bucuresti - Ilfov | 2208254 | 2209768 | 2215701 | 2232162 | 2242002 | 2221577 |
| RO41 | 212 | Sud-Vest Oltenia | 2325020 | 2313903 | 2301833 | 2285733 | 2270776 | 2299453 |
| RO42 | 213 | Vest | 1943025 | 1935094 | 1929158 | 1926707 | 1926700 | 1932137 |
| S101 | 214 | Vzhodna Slovenija | 1078747 | 1077922 | 1078992 | 1080901 | 1087771 | 1080867 |
| S102 | 215 | Zahodna Slovenija | 917686 | 919668 | 924366 | 929476 | 938095 | 925858 |
| SK01 | 216 | Bratislavský kraj | 599787 | 601132 | 603699 | 606753 | 610850 | 604444 |
| SK02 | 217 | Západné Slovensko | 1863932 | 1863940 | 1863056 | 1862227 | 1863740 | 1863379 |
| SK03 | 218 | Stredné Slovensko | 1352452 | 1352497 | 1351882 | 1351088 | 1350366 | 1351657 |
| SK04 | 219 | Východné Slovensko | 1563882 | 1567253 | 1570543 | 1573569 | 1576042 | 1570258 |
| Fl13 | 220 | Itä-Suomi | 669354 | 667056 | 664196 | 660859 | 657257 | 663744 |
| F118 | 221 | Etelä-Suomi | 2569358 | 2580801 | 2595823 | 2613925 | 2632744 | 2598530 |
| FI19 | 222 | Länsi-Suomi | 1325241 | 1330371 | 1334293 | 1338973 | 1344565 | 1334689 |
| FIIA | 223 | Pohjois-Suomi | 629432 | 631853 | 634502 | 636275 | 638765 | 634165 |
| F120 | 224 | Aland | 26347 | 26530 | 26766 | 26923 | 27153 | 26744 |
| SE11 | 225 | Stockholm | 1860872 | 1872900 | 1889945 | 1918104 | 1949516 | 1898267 |
| SE12 | 226 | Östra Mellansverige | 1509841 | 1514549 | 1518077 | 1524509 | 1534529 | 1520301 |
| SE21 | 227 | Småland med öarna | 798528 | 799739 | 800054 | 802247 | 805353 | 801184 |
| SE22 | 228 | Sydsverige | 1302586 | 1311254 | 1320160 | 1335936 | 1351257 | 1324239 |
| SE23 | 229 | Västsverige | 1796314 | 1805683 | 1814323 | 1827143 | 1838691 | 1816431 |
| SE31 | 230 | Norra Mellansverige | 826949 | 826188 | 825037 | 824853 | 825000 | 825605 |
| SE32 | 231 | Mellersta Norrland | 371750 | 371619 | 370764 | 370998 | 370386 | 371103 |
| SE33 | 232 | Övre Norrland | 508830 | 509460 | 509392 | 509467 | 508195 | 509069 |
| UKC1 | 233 | Tees Valley and Durham | 1150800 | 1153900 | 1156100 | 1161400 | : | 1155550 |
| UKC2 | 234 | Northumberland, Tyne and Wear | 1393000 | 1394000 | 1396600 | 1398700 | : | 1395575 |
| UKD1 | 235 | Cumbria | 492800 | 495000 | 495900 | 496500 | : | 495050 |
| UKD2 | 236 | Cheshire | 991100 | 994900 | 998400 | 1001700 | : | 996525 |
| UKD3 | 237 | Greater Manchester | 2530700 | 2538400 | 2548600 | 2558000 | : | 2543925 |
| UKD4 | 238 | Lancashire | 1435200 | 1443000 | 1448100 | 1450600 | : | 1444225 |
| UKD5 | 239 | Merseyside | 1360400 | 1358300 | 1355500 | 1351900 | : | 1356525 |
| UKE1 | 240 | East Yorkshire and Northern Lincolnshire | 892000 | 898500 | 903000 | 906300 | : | 899950 |
| UKE2 | 241 | North Yorkshire | 766300 | 773600 | 780200 | 786100 | : | 776550 |
| UKE3 | 242 | South Yorkshire | 1276300 | 1283500 | 1290300 | 1296200 | : | 1286575 |
| UKE4 | 243 | West Yorkshire | 2111100 | 2130200 | 2151500 | 2171200 | : | 2141000 |
| UKF1 | 244 | Derbyshire and Nottinghamshire | 2014200 | 2027900 | 2040300 | 2051200 | : | 2033400 |
| UKF2 | 245 | Leicestershire, Rutland and Northants | 1588100 | 1603600 | 1622200 | 1641200 | : | 1613775 |
| UKF3 | 246 | Lincolnshire | 670500 | 677900 | 683400 | 689500 | : | 680325 |
| UKG1 | 247 | Herefordshire, Worcestershire and Warks | 1237100 | 1243300 | 1250000 | 1256800 | : | 1246800 |
| UKG2 | 248 | Shropshire and Staffordshire | 1501900 | 1507300 | 1511700 | 1515500 | : | 1509100 |
| UKG3 | 249 | West Midlands | 2580200 | 2588100 | 2597000 | 2602000 | : | 2591825 |
| UKH1 | 250 | East Anglia | 2231000 | 2254900 | 2277800 | 2299000 | : | 2265675 |
| UKH2 | 251 | Bedfordshire, Hertfordshire | 1623200 | 1631600 | 1643400 | 1655600 | : | 1638450 |
| UKH3 | 252 | Essex | 1638700 | 1650500 | 1663600 | 1679200 | : | 1658000 |
| UK100 | 253 | Inner London + Outer London | 7376600 | 7422600 | 7484200 | 7534600 | : | 7454500 |
| UKJ1 | 254 | Berkshire, Bucks and Oxfordshire | 2119900 | 2134100 | 2151800 | 2170100 | : | 2143975 |
| UKJ2 | 255 | Surrey, East and West Sussex | 2577400 | 2589500 | 2605200 | 2625000 | : | 2599275 |
| UKJ3 | 256 | Hampshire and Isle of Wight | 1801900 | 1812300 | 1824400 | 1837300 | : | 1818975 |
| UKJ4 | 257 | Kent | 1606800 | 1618900 | 1629700 | 1640800 | : | 1624050 |
| UKK1 | 258 | Gloucestershire, Wiltshire and Bristol/Bath area | 2207000 | 2227500 | 2247500 | 2268200 | : | 2237550 |
| UKK2 | 259 | Dorset and Somerset | 1206600 | 1211600 | 1217100 | 1225300 | : | 1215150 |
| UKK3 | 260 | Cornwall and Isles of Scilly | 514900 | 519300 | 524000 | 529000 | : | 521800 |
| UKK4 | 261 | Devon | 1094900 | 1105900 | 1116800 | 1128500 | : | 1111525 |
| UKL1 | 262 | West Wales and The Valleys | 1871700 | 1877600 | 1881900 | 1888500 | : | 1879925 |
| UKL2 | 263 | East Wales | 1067100 | 1072400 | 1077800 | 1084400 | : | 1075425 |
| UKM2 | 264 | Eastern Scotland | 1914335 | 1927555 | 1941045 | 1956630 | : | 1934891 |
| UKM3 | 265 | South Western Scotland | 2281495 | 2282733 | 2283402 | 2285807 | : | 2283359 |
| UKM5 | 266 | North Eastern Scotland | 436775 | 438310 | 441240 | 445780 | : | 440526 |
| UKM6 | 267 | Highlands and Islands | 435296 | 438003 | 440164 | 442333 | : | 438949 |
| UKNO | 268 | Northern Ireland | 1706475 | 1717365 | 1733013 | 1750384 | : | 1726809 |

## Appendix E -- Definition of Potential Market Size in terms of GDP

The indicator on "potential market size", denoted 'potential GDP' in the MARKET SIZE pillar, provides an estimate of the GDP available within a pre-defined neighborhood, and taking into account the distance within this neighborhood.

Basic data necessary for the computation:
a) GDP/head in PPS, expressed as index of EU27 average, at NUTS2 level (source: Eurostat);
b) population distribution grid, at $1 \mathrm{~km}^{2}$ resolution (= POPL_01) (sources: JRC population disaggregation grid, national statistical institutes, REGIO-GIS);
c) NUTS2 polygon geometry (and a derived $1 \mathrm{~km}^{2}$ grid version of the NUTS2 geometry) (sources: Eurostat-GISCO and REGIO-GIS).

The computation of Potential Market Size expressed in GDP consists of the following steps:

1. To estimate GDP at the level of raster cells, values of regional GDP/head are transformed into a grid with $1 \mathrm{~km}^{2}$ resolution: this grid (= GDPPC_01) is the raster version of the NUTS2 GDP/head map. The GDP/head grid is then multiplied by the population grid, to obtain an estimate of GDP per raster cell. This estimate assumes a uniform distribution of GDP/head throughout the NUTS2 region.

GDP_01 = GDPPC_01 * POPL_01
Further steps in the analysis are carried out at the level of $10 * 10 \mathrm{~km}$ raster cells. Therefore, the $1 \mathrm{~km}^{2}$ GDP grid is aggregated to $100 \mathrm{~km}^{2}$ grid cells, by summing the GDP over the $1 \mathrm{~km}^{2}$ cells (result $=$ GDP_10).
2. Around each $100 \mathrm{~km}^{2}$ cell, a circular neighborhood with a radius of 100 km is defined. In this neighborhood, each cell obtains a weight varying between 100 in the centre of the neighborhood, and 0 at the outer limits of the neighborhood. For each cell of the territory, the focal sum of GDP in the neighborhood is calculated, weighted by the cell weights (i.e. inverse distance weighted). Finally, this sum is divided by 100 (because the maximum cell weight is 100 ).
3. To obtain regional and EU averages, the results at cell level is averaged at the level of NUTS2 regions or countries. In this way the cell values and the regional averages can be expressed as index of the European average. This transformation allows for an easier interpretation of the results: the index figure expresses how the GDP available in the neighborhood relates to the average GDP available in any neighborhood of the same size throughout the Union.

## Appendix F - Stages of development of EU NUTS 2 regions

| region_code | region | GDP (PPP per inhabitant in \% of EU average) 2007 | development stage |
| :---: | :---: | :---: | :---: |
| BEOO | Bruxelles Capital+Vlaams Brabant+Brabant Wallon | 154.6 | HIGH |
| BE21 | Prov. Antwerpen | 135.7 | HIGH |
| BE22 | Prov. Limburg (B) | 96.2 | INTERMEDIATE |
| BE23 | Prov.Oost Vlaanderen | 104.6 | HIGH |
| BE25 | Prov. West-Vlaanderen | 110.1 | HIGH |
| BE32 | Prov. Hainaut | 75.3 | INTERMEDIATE |
| BE33 | Prov. Liège | 85.3 | INTERMEDIATE |
| BE34 | Prov. Luxembourg (B) | 78.1 | INTERMEDIATE |
| BE35 | Prov. Namur | 79.7 | INTERMEDIATE |
| BG31 | Severozapaden | 25.6 | MEDIUM |
| BG32 | Severen ts entralen | 26.7 | MEDIUM |
| BG33 | Severoiztochen | 32.4 | MEDIUM |
| BG34 | Yugoiztochen | 30.7 | MEDIUM |
| BG41 | Yugozapaden | 62.0 | MEDIUM |
| BG42 | Yuzhen tsentralen | 27.2 | MEDIUM |
| CZO1 | Praha | 171.8 | HIGH |
| CZO2 | Strední Cechy | 75.2 | INTERMEDIATE |
| CZ03 | Jihozápad | 71.1 | MEDIUM |
| CZO4 | Severozápad | 61.7 | MEDIUM |
| CZ05 | Severovýchod | 65.9 | MEDIUM |
| CZ06 | Jihovýchod | 71.7 | MEDIUM |
| CZO7 | Strední Morava | 62.3 | MEDIUM |
| CZ08 | Moravskoslezsko | 67.5 | MEDIUM |
| DK01 | Hovedstaden | 150.3 | HIGH |
| DK02 | Sjælland | 91.4 | INTERMEDIATE |
| DK03 | Syddanmark | 113.3 | HIGH |
| DK04 | Midtjylland | 115.4 | HIGH |
| DK05 | Nordjylland | 110.0 | HIGH |
| DE11 | Stuttgart | 141.4 | HIGH |
| DE12 | Karlsruhe | 132.2 | HIGH |
| DE13 | Freiburg | 114.2 | HIGH |
| DE14 | Tübingen | 125.3 | HIGH |
| DE21 | Oberbayern | 164.7 | HIGH |
| DE22 | Niederbayern | 115.8 | HIGH |
| DE23 | Oberpfalz | 122.1 | HIGH |
| DE24 | Oberfranken | 113.1 | HIGH |
| DE25 | Mittel franken | 132.5 | HIGH |
| DE26 | Unterfranken | 117.5 | HIGH |
| DE27 | Schwaben | 120.9 | HIGH |
| DE30 | Berlin | 97.8 | INTERMEDIATE |
| DE41 | Brandenburg - Nordost | 76.1 | INTERMEDIATE |
| DE42 | Brandenburg - Südwest | 87.3 | INTERMEDIATE |
| DE50 | Bremen | 158.6 | HIGH |
| DE60 | Hamburg | 192.0 | HIGH |
| DE71 | Darmstadt | 156.1 | HIGH |
| DE72 | Gießen | 107.5 | HIGH |
| DE73 | Kassel | 115.2 | HIGH |
| DE80 | Mecklenburg-Vorpommern | 81.1 | INTERMEDIATE |
| DE91 | Braunschweig | 111.4 | HIGH |


| DE92 | Hannover | 110.8 | HIGH |
| :---: | :---: | :---: | :---: |
| DE93 | Lüneburg | 83.7 | INTERMEDIATE |
| DE94 | Weser-Ems | 101.0 | HIGH |
| DEA1 | Düsseldorf | 127.6 | HIGH |
| DEA2 | Köln | 118.0 | HIGH |
| DEA3 | Münster | 98.3 | INTERMEDIATE |
| DEA4 | Detmold | 109.4 | HIGH |
| DEA5 | Arnsberg | 106.3 | HIGH |
| DEB1 | Koblenz | 97.5 | INTERMEDIATE |
| DEB2 | Trier | 94.2 | INTERMEDIATE |
| DEB3 | Rheinhessen-Pfalz | 106.3 | HIGH |
| DECO | Saarland | 114.5 | HIGH |
| DED1 | Chemnitz | 82.6 | INTERMEDIATE |
| DED2 | Dresden | 87.7 | INTERMEDIATE |
| DED3 | Leipzig | 88.6 | INTERMEDIATE |
| DEEO | Sachsen-Anhalt | 83.6 | INTERMEDIATE |
| DEFO | Schleswig-Holstein | 99.5 | INTERMEDIATE |
| DEGO | Thüringen | 83.0 | INTERMEDIATE |
| EEOO | Estonia | 68.8 | MEDIUM |
| IE01 | Border, Midlands and Western | 99.2 | INTERMEDIATE |
| IE02 | Southern and Eastern | 166.1 | HIGH |
| GR11 | Anatoliki Makedonia, Thraki | 62.1 | MEDIUM |
| GR12 | Kentriki Makedonia | 72.5 | MEDIUM |
| GR13 | Dytiki Makedonia | 75.8 | INTERMEDIATE |
| GR14 | Thessalia | 68.2 | MEDIUM |
| GR21 | Ipeiros | 68.3 | MEDIUM |
| GR22 | Ionia Nisia | 74.0 | MEDIUM |
| GR23 | Dytiki Ellada | 59.8 | MEDIUM |
| GR24 | Sterea Ellada | 83.9 | INTERMEDIATE |
| GR25 | Peloponnisos | 75.7 | INTERMEDIATE |
| GR30 | Attiki | 128.1 | HIGH |
| GR41 | Voreio Aigaio | 66.6 | MEDIUM |
| GR42 | Notio Aigaio | 96.2 | INTERMEDIATE |
| GR43 | Kriti | 83.7 | INTERMEDIATE |
| ES11 | Galicia | 88.8 | INTERMEDIATE |
| ES12 | Principado de Asturias | 96.9 | INTERMEDIATE |
| ES13 | Cantabria | 105.4 | HIGH |
| ES21 | Pais Vasco | 136.8 | HIGH |
| ES22 | Comunidad Foral de Navarra | 132.2 | HIGH |
| ES23 | La Rioja | 112.0 | HIGH |
| ES24 | Aragón | 114.4 | HIGH |
| ES30 | Comunidad de Madrid | 136.8 | HIGH |
| ES41 | Castilla y León | 101.4 | HIGH |
| ES42 | Castilla-la Mancha | 81.5 | INTERMEDIATE |
| ES43 | Extremadura | 72.4 | MEDIUM |
| ES51 | Cataluña | 123.3 | HIGH |
| ES52 | Comunidad Valenciana | 95.3 | INTERMEDIATE |
| ES53 | Illes Balears | 113.8 | HIGH |
| ES61 | Andalucia | 81.2 | INTERMEDIATE |
| ES62 | Región de Murcia | 86.9 | INTERMEDIATE |
| ES63 | Ciudad Autónoma de Ceuta (ES) | 97.3 | INTERMEDIATE |
| ES64 | Ciudad Autónoma de Melilla (ES) | 94.5 | INTERMEDIATE |


| ES70 | Canarias (ES) | 92.8 | INTERMEDIATE |
| :---: | :---: | :---: | :---: |
| FR10 | Île de France | 168.7 | HIGH |
| FR21 | Champagne-Ardenne | 99.7 | INTERMEDIATE |
| FR22 | Picardie | 85.7 | INTERMEDIATE |
| FR23 | Haute-Normandie | 98.4 | INTERMEDIATE |
| FR24 | Centre | 95.3 | INTERMEDIATE |
| FR25 | Basse-Normandie | 88.3 | INTERMEDIATE |
| FR26 | Bourgogne | 94.5 | INTERMEDIATE |
| FR30 | Nord - Pas-de-Calais | 88.2 | INTERMEDIATE |
| FR41 | Lorraine | 88.7 | INTERMEDIATE |
| FR42 | Alsace | 102.2 | HIGH |
| FR43 | Franche-Comté | 90.1 | INTERMEDIATE |
| FR51 | Pays de la Loire | 97.7 | INTERMEDIATE |
| FR52 | Bretagne | 94.7 | INTERMEDIATE |
| FR53 | Poitou-Charentes | 90.4 | INTERMEDIATE |
| FR61 | Aquitaine | 98.2 | INTERMEDIATE |
| FR62 | Midi-Pyrénées | 97.3 | INTERMEDIATE |
| FR63 | Limousin | 87.7 | INTERMEDIATE |
| FR71 | Rhône-Alpes | 109.5 | HIGH |
| FR72 | Auvergne | 91.4 | INTERMEDIATE |
| FR81 | Languedoc-Roussillon | 85.6 | INTERMEDIATE |
| FR82 | Provence-Alpes-Côte d'Azur | 102.2 | HIGH |
| FR83 | Corse | 84.5 | INTERMEDIATE |
| FR91 | Guadeloupe (FR) | 76.4 | INTERMEDIATE |
| FR92 | Martinique (FR) | 75.1 | INTERMEDIATE |
| FR93 | Guyane (FR) | 48.7 | MEDIUM |
| FR94 | Reunion (FR) | 62.5 | MEDIUM |
| ITC1 | Piemonte | 113.6 | HIGH |
| ITC2 | Valle d'Aosta/Vallée d'Aoste | 118.6 | HIGH |
| ITC3 | Liguria | 106.8 | HIGH |
| ITC4 | Lombardia | 134.8 | HIGH |
| ITD1 | Provincia Autonoma Bolzano-Bozen | 134.5 | HIGH |
| ITD2 | Provincia Autonoma Trento | 122.0 | HIGH |
| ITD3 | Veneto | 121.6 | HIGH |
| ITD4 | Friuli-Venezia Giulia | 116.6 | HIGH |
| ITD5 | Emilia-Romagna | 128.0 | HIGH |
| ITE1 | Toscana | 112.8 | HIGH |
| ITE2 | Umbria | 96.9 | INTERMEDIATE |
| ITE3 | Marche | 105.5 | HIGH |
| ITE4 | Lazio | 122.3 | HIGH |
| ITF1 | Abruzzo | 85.3 | INTERMEDIATE |
| ITF2 | Molise | 77.9 | INTERMEDIATE |
| ITF3 | Campania | 65.9 | MEDIUM |
| ITF4 | Puglia | 66.8 | MEDIUM |
| ITF5 | Basilicata | 75.0 | INTERMEDIATE |
| ITF6 | Calabria | 65.8 | MEDIUM |
| ITG1 | Sicilia | 66.0 | MEDIUM |
| ITG2 | Sardegna | 78.4 | INTERMEDIATE |
| CYOO | Cyprus | 93.6 | INTERMEDIATE |
| LV00 | Latvia | 55.7 | MEDIUM |
| LTOO | Lithuania | 59.3 | MEDIUM |
| LU00 | Luxembourg (Grand-Duché) | 275.2 | HIGH |


| HU10 | Közép-Magyarország | 102.9 | HIGH |
| :---: | :---: | :---: | :---: |
| HU21 | Közép-Dunántúl | 58.2 | MEDIUM |
| HU22 | Nyugat-Dunántúl | 61.5 | MEDIUM |
| HU23 | Dél-Dunántúl | 42.7 | MEDIUM |
| HU31 | Észak-Magyarország | 40.1 | MEDIUM |
| HU32 | Észak-Alföld | 39.4 | MEDIUM |
| HU33 | Dél-Alföld | 41.8 | MEDIUM |
| MT00 | Malta | 76.4 | INTERMEDIATE |
| NL11 | Groningen | 164.9 | HIGH |
| NL12 | Friesland (NL) | 107.5 | HIGH |
| NL13 | Drenthe | 103.6 | HIGH |
| NL21 | Overijssel | 114.7 | HIGH |
| NL22 | Gelderland | 113.5 | HIGH |
| NL23 | Flevoland | 107.3 | HIGH |
| NL31 | Utrecht | 155.4 | HIGH |
| NL32 | Noord-Holland | 150.1 | HIGH |
| NL33 | Zuid-Holland | 136.6 | HIGH |
| NL34 | Zeeland | 121.6 | HIGH |
| NL41 | Noord-Brabant | 134.4 | HIGH |
| NL42 | Limburg (NL) | 119.4 | HIGH |
| AT11 | Burgenland (A) | 81.3 | INTERMEDIATE |
| AT12 | Niederösterreich | 100.1 | HIGH |
| AT13 | Wien | 163.1 | HIGH |
| AT21 | Kärnten | 104.6 | HIGH |
| AT22 | Steiermark | 106.1 | HIGH |
| AT31 | Oberösterreich | 119.9 | HIGH |
| AT32 | Salzburg | 139.5 | HIGH |
| AT33 | Tirol | 128.2 | HIGH |
| AT34 | Vorarlberg | 128.1 | HIGH |
| PL11 | Lódzkie | 50.0 | MEDIUM |
| PL12 | Mazowieckie | 87.1 | INTERMEDIATE |
| PL21 | Malopolskie | 46.7 | MEDIUM |
| PL22 | Slaskie | 57.8 | MEDIUM |
| PL31 | Lubelskie | 36.9 | MEDIUM |
| PL32 | Podkarpackie | 36.7 | MEDIUM |
| PL33 | Swietokrzyskie | 41.9 | MEDIUM |
| PL34 | Podlaskie | 40.4 | MEDIUM |
| PL41 | Wielkopolskie | 56.9 | MEDIUM |
| PL42 | Zachodniopomorskie | 48.9 | MEDIUM |
| PL43 | Lubuskie | 48.2 | MEDIUM |
| PL51 | Dolnoslaskie | 59.2 | MEDIUM |
| PL52 | Opolskie | 45.2 | MEDIUM |
| PL61 | Kujawsko-Pomorskie | 47.3 | MEDIUM |
| PL62 | Warminsko-Mazurskie | 40.5 | MEDIUM |
| PL63 | Pomorskie | 53.6 | MEDIUM |
| PT11 | Norte | 60.3 | MEDIUM |
| PT15 | Algarve | 79.6 | INTERMEDIATE |
| PT16 | Centro (PT) | 64.4 | MEDIUM |
| PT17 | Lisboa | 104.7 | HIGH |
| PT18 | Alentejo | 71.9 | MEDIUM |
| PT20 | Região Autónoma dos Açores (PT) | 67.6 | MEDIUM |
| PT30 | Região Autónoma da Madeira (PT) | 96.3 | INTERMEDIATE |


| RO11 | Nord-Vest | 40.2 | MEDIUM |
| :---: | :---: | :---: | :---: |
| RO12 | Centru | 42.2 | MEDIUM |
| RO21 | Nord-Est | 26.6 | MEDIUM |
| RO22 | Sud-Est | 33.8 | MEDIUM |
| RO31 | Sud - Muntenia | 34.2 | MEDIUM |
| RO32 | Bucuresti - Ilfov | 92.2 | INTERMEDIATE |
| RO41 | Sud-Vest Oltenia | 32.7 | MEDIUM |
| RO42 | Vest | 48.2 | MEDIUM |
| SIO1 | Vzhodna Slovenija | 73.1 | MEDIUM |
| SIO2 | Zahodna Slovenija | 106.7 | HIGH |
| SKO1 | Bratislavský kraj | 160.3 | HIGH |
| SK02 | Západné Slovensko | 66.1 | MEDIUM |
| SK03 | Stredné Slovensko | 53.3 | MEDIUM |
| SK04 | Východné Slovensko | 46.0 | MEDIUM |
| FI13 | Itä-Suomi | 88.8 | INTERMEDIATE |
| FI18 | Etelä-Suomi | 135.6 | HIGH |
| FI19 | Länsi-Suomi | 104.9 | HIGH |
| FI1A | Pohjois-Suomi | 102.3 | HIGH |
| FI20 | Åland | 143.2 | HIGH |
| SE11 | Stockholm | 164.6 | HIGH |
| SE12 | Östra Mellansverige | 106.2 | HIGH |
| SE21 | Småland med öarna | 110.0 | HIGH |
| SE22 | Sydsverige | 110.1 | HIGH |
| SE23 | Västsverige | 119.1 | HIGH |
| SE31 | Norra Mellansverige | 108.1 | HIGH |
| SE32 | Mellersta Norrland | 108.3 | HIGH |
| SE33 | Övre Norrland | 115.1 | HIGH |
| UKC1 | Tees Valley and Durham | 81.5 | INTERMEDIATE |
| UKC2 | Northumberland, Tyne and Wear | 97.8 | INTERMEDIATE |
| UKD1 | Cumbria | 89.7 | INTERMEDIATE |
| UKD2 | Cheshire | 123.7 | HIGH |
| UKD3 | Greater Manchester | 105.3 | HIGH |
| UKD4 | Lancashire | 89.9 | INTERMEDIATE |
| UKD5 | Merseyside | 83.2 | INTERMEDIATE |
| UKE1 | East Yorkshire and Northern Lincolnshire | 90.5 | INTERMEDIATE |
| UKE2 | North Yorkshire | 101.2 | HIGH |
| UKE3 | South Yorkshire | 90.2 | INTERMEDIATE |
| UKE4 | West Yorkshire | 103.5 | HIGH |
| UKF1 | Derbyshire and Nottinghamshire | 100.6 | HIGH |
| UKF2 | Leicestershire, Rutland and Northants | 114.4 | HIGH |
| UKF3 | Lincolnshire | 83.3 | INTERMEDIATE |
| UKG1 | Herefordshire, Worcestershire and Warks | 100.6 | HIGH |
| UKG2 | Shropshire and Staffordshire | 89.0 | INTERMEDIATE |
| UKG3 | West Midlands | 105.3 | HIGH |
| UKH1 | East Anglia | 110.4 | HIGH |
| UKH2 | Bedfordshire, Hertfordshire | 127.0 | HIGH |
| UKH3 | Essex | 98.0 | INTERMEDIATE |
| UKI | Inner London + Outer London | 225.6 | HIGH |
| UKJ1 | Berkshire, Bucks and Oxfordshire | 156.1 | HIGH |
| UKJ2 | Surrey, East and West Sussex | 122.4 | HIGH |
| UKJ3 | Hampshire and Isle of Wight | 116.9 | HIGH |
| UKJ4 | Kent | 93.4 | INTERMEDIATE |


| UKK1 | Gloucestershire, Wiltshire and Bristol/Bath area | 128.3 | HIGH |
| :--- | :--- | :---: | :---: |
| UKK2 | Dorset and Somerset | 97.3 | INTERMEDIATE |
| UKK3 | Cornwall and Isles of Scilly | 75.2 | INTERMEDIATE |
| UKK4 | Devon | 88.6 | INTERMEDIATE |
| UKL1 | West Wales and The Valleys | 73.4 | MEDIUM |
| UKL2 | East Wales | 110.3 | HIGH |
| UKM2 | Eastern Scotland | 119.9 | HIGH |
| UKM3 | South Western Scotland | 103.6 | HIGH |
| UKM5 | North Eastern Scotland | 152.9 | HIGH |
| UKM6 | Highlands and Islands | 87.2 | INTERMEDIATE |
| UKNO | Northern Ireland | 92.8 | INTERMEDIATE |

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#### Abstract

The joint project between DG Joint Research Centre and DG Regional Policy on the construction of the EU Regional Competitiveness Index (RCI) aims at producing a composite indicator which measures the competitiveness of European regions at the NUTS 2 level for all EU Member States. The concept of competitiveness has been largely discussed over the last decades. A broad notion of competitiveness refers to the inclination and skills to compete, to win and retain position in the market, increasing market share and profitability, thus, being commercially successful. The concept of regional competitiveness which has gained more and more attention in recent years, mostly due to the increased attention given to regions as key in the organization and governance of economic growth and the creation of wealth. An important example is the special issue of Regional Studies, published in 2004, fully devoted to the concept of competitiveness of regions. Regional competitiveness is not only an issue of academic interest but of increasing policy deliberation and action. This is reflected in the interest devoted in the recent years by the European Commission to define and evaluate competitiveness of European regions, an objective closely related to the realization of the Lisbon Strategy on Growth and Jobs. Why measuring regional competitiveness is so important? Because "if you can not measure it, you can not improve it" (Lord Kelvin). A quantitative score of competitiveness will help Member States in identifying possible regional weaknesses together with factors mainly driving these weaknesses. This in turn will assist regions in the catching up process. Given the multidimensional nature of the competitiveness concept, the structure of RCl is made of eleven pillars which describe the concept, taking into account its regional dimension, with particular focus on a region's potential. The long-term perspective is, in fact, essential for European policy and people's skills are understood to play a key role for EU future, as also underlined by the president of the Lisbon Council in his recent policy brief. For this reason the RCI includes aspects related to short and long-term capabilities of regions, with a special focus on innovation, higher education, lifelong learning and technological availability and use, both at the individual and at the enterprise level. A number of indicators have been selected to describe these dimensions with criteria based on coverage and comparability as well as within pillar statistical coherence. Most indicators come from Eurostat but where data was not available, alternative source were considered. A detailed univariate and multivariate statistical analyses have been carried out on the set of candidate indicators for the setting-up and refinement of the composite. Each choice with a certain degree of uncertainty has been submitted to a full robustness analysis to evaluate the level of variability of regions final score and ranking.


The final RCI shows a heterogeneous situation across EU regions with Eastern and Southern European regions showing lower performance while more competitive regions are observed in Northern Europe and parts of Continental Europe.

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[^0]:    ${ }^{1}$ Qualitative indicators from the Executive Opinion Survey are treated as quantitative as such.

[^1]:    ${ }^{2}$ The Global Competitiveness Report does not detail the computation of the average score within each pillar. It was deduced from the context that simple means are computed from (1-7) scaled indicators which describe the pillar.

[^2]:    ${ }^{3}$ The total knowledge employment index is computed by aggregating employment per capita across all

[^3]:    knowledge sectors.
    ${ }^{4}$ Higher education degrees are levels 5 and 6 according to the ISCED classification.

[^4]:    ${ }^{5}$ Since patenting varies strongly between years, the average across 6 years is considered to smooth variation.
    ${ }^{6}$ This is a very specific indicator developed by the authors (Alanen et al., 2000)

[^5]:    ${ }^{7}$ The numbering of the pillars follows their numbering in the text.

[^6]:    ${ }^{8}$ For more information on statistics on LMP, see
    http://epp.eurostat.ec.europa.eu/portal/page/portal/labour_market/labour_market_policy

[^7]:    ${ }^{9}$ The second criterion is related to the extent to which countries are factor-driven, i.e. they compete based on their factor endowments, primarily unskilled labor and natural resources. The proxy used is the share of exports of primary goods in total exports. As EU economies are not factor-driven in the GCI definition, we consider this criterion not to be relevant for the definition of the stages of development at the EU regional level.

[^8]:    ${ }^{10}$ For sake of simplicity the reference to the country is omitted in the notation.

[^9]:    ${ }^{11}$ Regions in between best and worst $10 \%$ are marked in grey. The white color has been used for regions which either have missing observations on the indicator in question or represent regions which do not belong to the

[^10]:    ${ }^{12}$ If the sign of the correlation of the indicators with the main factors is the same, it means that the set of indicators have all the same orientation with respect to the level of competitiveness.

[^11]:    ${ }^{13}$ In some cases the worst performers include more than $10 \%$ of all regions in order to accommodate the fact that they all have the same value for the indicator.

[^12]:    ${ }^{14}$ In the case of university accessibility indicator, the top performers include more than $10 \%$ of all regions in order to accommodate the fact that they all have the same value for the indicator.

[^13]:    ${ }^{15}$ Labor force, or active population, is the sum of employed and unemployed people, a synonymous is economically active population.

[^14]:    ${ }^{16}$ In the case of Biotech patents, the worst performing regions are more than $10 \%$ because more regions have the same value for the indicator.

[^15]:    ${ }^{17}$ http://www.weforum.org/en/initiatives/gcp/Global\%20Competitiveness\%20Report/index.htm

[^16]:    ${ }^{18}$ EU employment data is collected from the Labour Force Survey (LFS) and from the Structural Business Statistics (SBS), administrated by Eurostat and has been integrated with data from National Statistical Offices. A detailed list of all sources is available at: http://www.clusterobservatory.eu/index.php?id=47\&nid.

[^17]:    ${ }^{19}$ The full list of cluster categories is available at http://www.clusterobservatory.eu/index.php?id=46\&nid.
    ${ }^{20}$ We have used the identification of Technology and Knowledge-intensive sectors used by Eurostat and available at http://europa.eu.int/estatref/info/sdds/en/hrst/hrst sectors.pdf

