

Towards the Realization of the Europe 2020 Agenda for Economic Growth in the European Union: An Empirical Analysis based on Goal Programming

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Abstract

Goal Programming is a simple and widely used technique to solve policy-making problems involving different and conflicting objectives or criteria. It consists in minimizing for each criterion the distance between the achieved level and its respective goal. Weighted Goal Programming (WGP) is suitable to the analysis of macroeconomic policies involving several competing objectives with preferred weights of importance as it allows finding the optimal strategy for global sustainability which implies several coexisting goals. In this chapter we propose a WGP model that can be used to determine the optimal allocation of labor in each economic sector in order to minimize the deviations from the goals of four different criteria which model economic, environmental, energetic and social objectives. We apply our model to each country of the European Union and measure their performance with respect to the Europe 2020 agenda. Our model provides insights and policy recommendations, such as a better integration of the incoming workforce in a context of increasing immigration flows, development of renewable sources of energy, and green sustained transformation of national economic environments.

Keywords: Weighted Goal Programming, Mixed Integer Linear Programming, Multiple Criteria Decision Analysis, The Europe 2020 agenda, Sustainable Development.

1 Introduction

As awareness on environmental issue has spiked, with dissemination of social movements in favor of stronger environmental respect, and profusion of international treaties, optimal resource allocation and economic growth cannot be distinguished from sustainability and ecological considerations. Nowadays countries are subject to numerous binding frameworks for policy making (i.e. the 1992 United Nations Framework Convention on Climate Change, and the more recent Cop 21 conference organized in Paris in 2016), introducing greenhouse gas emission targets. As scientific reports (such as the Intergovernmental Panel on Climate Change) encounter large media coverage, the importance of these targets can be expected to increase, further adding constraints on economic policies design and implementation.

These added international and social incentives to more environmental-friendly measures yet encounter complex national and international economic situations. Indeed, most developed countries show low growth and budget recovery after the last global financial crisis, and improving, but still stagnant, labor markets, thus balancing these fundamental policy topics with the more recent environmental stakes is evidently a struggle. An illustrative example did arise in France, when important social unrest from the “yellow jackets” originally emerged from an increase in the taxation of most polluting fuels. These elements further demonstrate the importance of careful consideration of each of the goals in the design of public policy. While such arbitrages are difficult to make by decision makers, several tools can bring a substantial help and support in policy design. The problem of simultaneously conciliating economic, environmental, social and energy policy targets is well captured by the mathematical models of multiple criteria decision analysis (MCDA), which indeed allow to identify the optimal policy design accounting for all different objectives.

The European Union is a typical and interesting case study. After the 2008 economic crisis, in 2010 European institutions and notably the European Council, and European Commission, proposed the introduction of numerical multi-criteria policy targets, the so-called “Europe 2020” strategy. This highly varied set of national targets covers a large subset of national issues, such as economic growth, employment, education and research, reduction of greenhouse gas emissions, progress towards higher energy efficiency, and struggle against poverty and social exclusion. While these sets of actions and associated objectives (such as a decrease by 20% of GHG emissions, or an increase in energy efficiency by 20%) are formulated at the European Union scale, the objectives are nationally transposed, leaving European Union members a relative margin of freedom in their implementation. This feature opens the possibility of evaluating the satisfaction of Europe 2020 objectives by European Union states in MCDA prospective.

The application of MCDA techniques, and specifically Goal Programming (GP) and its variants and extensions, to macroeconomic policies is not novel. They have been widely used to tackle similar issues of decision making with competing objectives; as they allow to identify a Pareto optimal solution with respect to the goals involved. Nevertheless, the use of GP models has before emerged in engineering, where they have been implemented to analyse supply chain optimization problems facing imprecise assessment of both demand and information (Selim and Ozkarahan, 2008a; Tsai, 2009), vendor problems (Kumar et al., 2004; Zhou et al., 2008), production planning (Selim et al., 2008b), or also decision making in manufacturing (Sheikhalishahi and Torabi, 2014; Taghizadeh et al., 2011). The increasing popularity of GP models, already noticed by Ignizio (1982), gave birth for instance to financial portfolio selection applications (Watada 1997; Inuiguchi and Ramik, 2000), with consideration of manager preferences (Mansour et al., 2007), highlighting the flexibility of GP approaches to integrate these constraints as well, and also in the context of optimal allocation of renewable energies with quota constraints (Daim et al., 2010). In marketing, decision making problems seeing an application of GP methods are sales operation (Tyagi et al., 2011) and media planning (Charnes et al., 1968; Jha et al., 2011). In agricultural and environmental management, as the GP methodology constitutes an adequate tool to model environmental interactions (Linares and Romero, 2000), uses of MCDA approaches have dealt with agricultural land planning (Biswas

and Pal, 2005; Sharmar et al., 2007), outsourcing management (Araz et al., 2007), and crops selection (Mirkarimi et al., 2013).

The extant MCDA literature cover a large variety of models and applications reviewed by Colapinto, Jayaraman and Marsiglio (2017), and the GP formulation has been applied to macroeconomic policy design and evaluation. Some researchers (Jayaraman et al., 2015, 2017a, 2017b) have used Weighted Goal Programming (WGP) to study the global sustainability and development of the GCC countries. Other authors employ this methodology to study macroeconomic policies in Spain (André, 2009; San Cristóbal, 2012) through the joint analysis of optimal economic and environmental policies. More recently, Omrani, Valipour and Emrouznejad (2018) have applied a WGP model to more efficiently plan regional sustainable development and workforce allocation in Iran. Zografidou et al. (2017) have suggested an optimal design of renewable energy production in Greece by weighting social, financial and production goals. Nomani et al. (2017) implemented a Fuzzy Goal Programming approach to evaluate the satisfaction of sustainability policy targets in India. Finally, Bravo, Jones, Pla-Santamaria and Wall (2018) have studied the robustness of WGP models applied to offshore wind-farm site location determination. Indeed, GP and its variants offer the ability to balance different objectives and are relevant tools to study the satisfaction of several mutually conflicting objectives. Recent economic works provide multi-criteria models emphasizing the comparison of welfare criteria (Colapinto et al., 2017), and studying the economy/environment trade-off through externality modelling, notably establishing that “independently of the relative importance of economic and environmental factors, it is paradoxically optimal for the economy to asymptotically reach the maximum pollution level that the environment is able to bear” in the multi-criteria decision-making (Marsiglio and Privileggi, 2019).

Our chapter proposes a GP model to describe and measure the progress of each European country towards the satisfaction of the main Europe 2020 goals, following authors’ previous work with the use of a Fuzzy GP model (Vié et al., 2019). With respect to Vié et al. (2019) this chapter presents a more detailed analysis with updated and more complete data. The Europe 2020 strategy is used as a reference framework for activities in the EU at national and regional levels. Our model is linear and includes seven economic sectors and four different criteria. The data have been collected using information provided by the European Commission and the EU statistics office, Eurostat. Eurostat regularly publishes comprehensive progress reports for the targets. We referred to the targets for different sectors indicated in the Europe 2020 agenda , including Employment, Research and development (R&D), Climate change and energy, Education, and Poverty and social exclusion. In particular, the Climate change and energy targets are:

- a) greenhouse gas emissions 20% lower than 1990 levels
- b) 20% of energy coming from renewables
- c) 20% increase in energy efficiency

Our model mainly focuses on the analysis of a sustainable economic growth and combines the economic objective (GDP) together with GHG emissions and energy efficiency. Our model

does not distinguish skilled from unskilled labor force and supposes that population level is equal to the labor force.

The chapter is organized as follows. Section 2 reviews some basic notions in MCDA and GP. Section 3 presents the model formulation, including a detailed description of the economic sectors and the criteria. Section 4 describes how the data have been collected, while Section 5 illustrates the results. As usual, Section 6 concludes. In the appendices (A to J) we report data calculation and model results.

2 A Review of Multiple Criteria Decision Analysis and Goal Programming

Multiple Criteria Decision Analysis or Multiple Criteria Decision Making (MCDM) is a discipline that considers decision making situations involving multiple and conflicting criteria. Some examples of conflicting criteria that have been considered in literature are quality, cost, price, satisfaction, risk, and others. Considering multiple criteria with respect to a single criterion leads to more informed and better decisions. However, typically there does not exist a unique optimal solution and it is necessary to use a decision maker's preferences to differentiate between solutions. Many important advances have been developed in this field in the last sixty years including new approaches, innovative methods, and sophisticated computational algorithms. A classical MCDA model involves several criteria, objectives or attributes, to be considered simultaneously. These dimensions are usually conflicting and the decision-maker will look for the solution of the best compromise.

MCDA models are based on the notion of partial or Pareto order which can be summarized as follows: Given two vectors $a, b \in R^p$, we say that $a \leq b$ if and only if $a_i \leq b_i$ for all $i = 1 \dots p$. The general formulation of a MCDA model can be stated as follows (Sawaragi et al., 1985): Given a set of p criteria f_1, f_2, \dots, f_p , determine the optimal solution of the vector function $f(x) := [f_1(x), f_2(x), \dots, f_p(x)]$ under the condition that $x \in D \subseteq \mathfrak{R}^n$ where D designates the set of feasible solutions. The optimal solution has to be understood in Pareto sense: We say that a point $\hat{x} \in D$ is a global Pareto optimal solution or global Pareto efficient solution if $f(x) \in f(\hat{x}) + (-R_+^p \setminus \{0\})^c$ for all $x \in D$. Practically speaking, a Pareto optimal solution describes a state in which goods and resources are distributed in such a way that it is not possible to improve a single criterion without also causing at least one other criterion to become worse off than before the change. In other words, a state is not Pareto efficient if there exists a certain change in allocation of goods and resources that may result in some criteria being in a better position with no criterion being in a worse position than before the change. If a point $x \in D$ is not Pareto efficient, there is potential for a Pareto improvement and an increase in Pareto efficiency.

GP is a well-known technique to solve MCDA models. The GP model is based on a notion of distance and it seeks to minimize positive or negative deviations of the achievement levels with respect to the aspiration ones. It is an aggregating methodology that allows obtaining a solution representing the best compromise that can be achieved by the decision maker, as noted by Jayaraman et al. (2015).

The WGP model is one of the possible variants of the GP model that have been proposed in literature. In this context, the decision maker can show different appreciation of the positive and negative deviations based on the relative importance of the objective and this is expressed by introducing different weights w_i^+ and w_i^- . The mathematical formulation of the WGP model reads as:

$$\text{Minimize } \sum_{i=1}^p w_i^+ \delta_i^+ + w_i^- \delta_i^-$$

Subject to:

$$f_i(x) + \delta_i^- - \delta_i^+ = g_i \quad i = 1 \dots p$$

$$x \in F$$

$$\delta_i^-, \delta_i^+ \geq 0 \quad i = 1 \dots p$$

3 Model Formulation

As previously mentioned, MCDA usually deals with decision making with multiple and conflicting criteria, objectives or attributes, and considers decision makers' preferences to determine the best compromise among optimal solutions. GP was first introduced by Charnes and Cooper (1961). Given a set of n linear criteria $F_i(X_1, X_2, \dots, X_n) = \sum_{j=1}^n A_{ij} X_{ij}$ and a set of goals G_i the WGP model reads as

$$\sum_{i=1}^p \alpha_i^+ D_i^+ + \alpha_i^- D_i^- \quad (1)$$

Subject to:

$$\sum_{j=1}^n A_{ij} X_j + D_i^- + D_i^+ = G_i, \quad i = 1 \dots p$$

$$X \in \Omega$$

$$D_i^-, D_i^+ \geq 0, i = 1 \dots p$$

where Ω is a feasible set, α_i^+ and α_i^- are weights, X_j are the input variables representing the number of employees in each economic sector, the coefficient A_{ij} states the contribution of the j^{th} variable to the achievement of the i^{th} criterion, and D_i^- and D_i^+ are the positive and negative deviations with respect to the aspirational goal levels G_i , $i = 1, \dots, n$, respectively.

Previous researchers – see i.e. Andre' et al. (2009) and San Cristóbal (2012) in Spain, Jayaraman et al. (2017a, 2017b) in the GCC countries – have shown how the government can determine its optimal policy according to different criteria using the GP approach. Considering the Europe 2020 objectives established by the European Commission, our macroeconomic model simultaneously considers the following four criteria with their respective units:

- a) F_1 is the economic output (in million US\$)
- b) F_2 is the GHG emissions (in Gg of CO₂ equivalent kilo tonnes)

- c) F_3 is the electric consumption (in thousand tonnes of oil equivalent)
- d) F_4 is the number of employees (in thousands)

The decision variables in our GP model are all relevant economic sectors for the analysis. They are equivalent to the main activities identified by NACE Rev. 2 classification:

- X_1 : agriculture, forestry and fishing;
- X_2 : energy industry;
- X_3 : manufacturing industry;
- X_4 : construction and residential;
- X_5 : trade, transports, distribution and repairing;
- X_6 : commercial services (information, communication, financial and insurance activities);
- X_7 : general services (administrative, state, technical, scientific, education, health and social services).

Each criterion F_i is linear with respect to each decision variable X_j and takes the form:

$$F_1(X_1, X_2, \dots, X_7) = A_{11}X_1 + A_{12}X_2 + \dots + A_{17}X_7$$

$$F_2(X_1, X_2, \dots, X_7) = A_{21}X_1 + A_{22}X_2 + \dots + A_{27}X_7$$

$$F_3(X_1, X_2, \dots, X_7) = A_{31}X_1 + A_{32}X_2 + \dots + A_{37}X_7$$

$$F_4(X_1, X_2, \dots, X_7) = A_{41}X_1 + A_{42}X_2 + \dots + A_{47}X_7$$

The seven economic categories aim to represent – for each country – its economic, social, environmental, and energy characteristics. We also replicate the choice made in previous publications (San Cristóbal, 2012; Jayaraman et al 2017a), that fit well the NACE (second revision) classification of economic activities created within the European Union. Such a classification is complete, aiming to describe efficiently whole economic patterns. Our choice of categories is restricted either to specific global categories, such as agriculture combined with fishing and forestry, either to an aggregate category as commercial services, which includes both financial and insurance activities as well as information and communication economics.

The GP problem we intend to solve can then be written in the following form:

$$\text{Minimize } (\alpha_1^+ D_1^+ + \alpha_1^- D_1^-) + (\alpha_2^+ D_2^+ + \alpha_2^- D_2^-) + (\alpha_3^+ D_3^+ + \alpha_3^- D_3^-) + (\alpha_4^+ D_4^+ + \alpha_4^- D_4^-) \quad (2)$$

Subject to:

$$A_{11}X_1 + A_{12}X_2 + \dots + A_{17}X_7 - D_1^+ + D_1^- = G_1$$

$$A_{21}X_1 + A_{22}X_2 + \dots + A_{27}X_7 - D_2^+ + D_2^- = G_2$$

$$A_{31}X_1 + A_{32}X_2 + \dots + A_{37}X_7 - D_3^+ + D_3^- = G_3$$

$$X_1 + X_2 + \dots + X_7 - D_4^+ + D_4^- = G_4$$

$$X_1 \geq \Omega_1, X_2 \geq \Omega_2, X_3 \geq \Omega_3, X_4 \geq \Omega_4, X_5 \geq \Omega_5, X_6 \geq \Omega_6, X_7 \geq \Omega_7$$

$$X_j, j = 1, 2, \dots, 7 \text{ are positive and integer}$$

$$D_i^+, D_i^- \geq 0, i = 1, 2, 3, 4$$

The variables D_i^+, D_i^- describe the positive and the negative deviations. The input variables X_j take integer values and must be at least equal to the positive number Ω_j which is the number of employees in each sector of our analysis (see Appendix A). A_{1j} is the economic output per capita (worker) for the j^{th} economic sector (see Appendix B). A_{2j} describes the GHG emission per capita (worker) for the j^{th} economic sector (see Appendix C). A_{3j} models the energy consumption per capita (worker) for the j^{th} economic sector (see Appendix D).

For each country we solve the above model (2) which takes then the form of a Mixed Integer Linear Programming (MILP) model. We have implemented it using LINGO. So far, we have assumed equal weights for each objective. In other words, all the weights α_n^+ and α_n^- , $n = 1, \dots, 4$, are equal. Our choice is motivated by the fact that the Europe 2020 agenda does not provide any priority or ranking among the objectives and all of them must be jointly met. However, these weights might be modified at the regional level when the national preferences and the economic situation of each country have to be taken into account.

4 Data Collection and Computation

In this section we briefly discuss how we have estimated all parameters involved in the model. For each constraint we describe how the data has been collected and how the estimations have been computed. We also provide their interpretation.

3.1 The Gross Domestic Product constraint

The GDP per capita denoted A_{1i} (see Appendix B) is expressed in US\$ thousands per capita. It is computed for a given sector i in each country by taking the ratio of the economic output (GDP) for the selected sector i in US\$ millions in 2015 (Eurostat, 2014, 2015), and the number of employees (expressed in thousands) in the same year for the same economic sector i (OECD Statistics, 2015; Eurostat, 2015). Following the Europe 2020 recommendation that the economic output must be at least conserved with respect to the year of analysis, and following GDP growth rate projections of the International Monetary Fund (IMF) and the OECD up to 2020, we defined the economic GDP objective G_1 as the sum of the forecast GDP of all sectors. For instance, a country with a GDP of 100 in 1990 forecasted to grow by 10% to 2020 will have an economic output target of 110. The resulting GDP constraint can be expressed as follows: the economic output of the country needs to be at least as good as current projections.

3.2 The greenhouse gas constraint

The average sectorial greenhouse gas (GHG) emissions per capita A_{2i} (see Appendix C for the data) is expressed in tonnes of CO₂ equivalent per capita, for a given sector i in each country. It is computed by dividing sectorial GHG emissions (in thousands of tonnes) in the selected country in 2015, and the number of employees (expressed in thousands) in the same year for the same economic sector and country (OECD Statistics, 2015; Eurostat, 2014, 2015). This delivers the benchmark from which the environmental policy target G_2 is computed, accordingly to the Europe 2020 objective consisting in a decrease by 20% of GHG emissions at the scale of the national economy, an objective that we apply to all sectors separately. For instance, a country with a GHG emissions level of 100 in 1990 will have a GHG emission target of 80 at the horizon 2020.

3.3 The energy constraint

The average sectorial energy consumption per capita A_{i3} (see Appendix D for the data) is expressed in tonnes of oil equivalent per capita, for a given economic sector i in each country.

In order to obtain its numerical values, we took the ratio of the energy consumption for the selected sector i in the selected country in thousands of tonnes of oil equivalent in 2015 (International Energy Agency, 2014¹, and of the number of employees (expressed in thousands) in the same year for the same sector i and country, (OECD Statistics, 2015; Eurostat, 2014, 2015). The determination of the energy consumption targets G_3 proceeded by transforming the resulting consumption with the 20%² increase in energy efficiency imposed by the Europe 2020 strategy, with respect to the 1990 level constituting the benchmark of the policy set.

3.4 The employment constraint

This constraint is simple, as the model aims at maintaining employment at the benchmark level set in the year of analysis. This constraint is declined in each economic sector, based on the number of jobs recorded in the year of data collecting (OECD, 2015; Eurostat, 2015). Natural population growth (and estimated forecasts for later years) were considered to formulate the aggregate employment goal G_4 . Note that this constraint operates in the aggregate level, offering countries flexibility in the allocation of incoming workforce across economic sectors.

5 Model Implementation and Discussion

As said above, we implement our model using the software LINGO and the results are presented in Appendix I. We have ticked with “x” any time there is a significant deviation value (detailed results table can be found in Appendix J). In all simulations we have normalized the weights to 1. We would like to point out that confidential and missing data for Cyprus, Croatia and Malta, restrict us to make less precise conclusions than those done for other countries.

From what we observed, the current trend will allow all European countries to have $D_4^+ = 0$, which means that the entire available labor force will be used for creating a sustainable development. This result outlines the need for a better integration of unemployed workforce, possibly through intensified training policies and inclusive measures in the labor market. In the context of automation, improving the effectiveness of changes in qualification seems relevant to ensure optimal economic performance, employment and social inclusion.

In several countries we observe a significant nonzero value for the deviation D_4^- , meaning that the satisfaction of the objectives of the model requires additional workforce to be integrated within the economy in specific sectors. In a context of increasing immigration towards European countries, this study outlines the necessity and the benefits coming from a more efficient integration of new comers within the labor market in Eastern and Northern countries (for instance Germany).

Some countries present a significant deviation in the values taken by D_1^- , standing for a negative deviation from the economic goal. For these countries, the satisfaction of the simultaneous objectives leads to a slower economic output than targeted. This can be interpreted as a lack of productivity, that can be verified empirically comparing the average economic product of these countries, with economically well-performing countries. Improving competitiveness and

¹ Eurostat data in thousands of tonnes of oil equivalent (ktoe) on a net calorific value basis.

² Europe 2020 objectives in energy sector deal with an increasing of energy efficiency by 20%, which mathematically implies a reduction of energy consumption by 16.66% from the 1990 consumption to the horizon 2020.

production efficiency may result from organizational changes or evolutions in the length of work time permitted.

A significant deviation in the parameter D_1^+ , a positive deviation from the economic goal, shows a great economic performance of the country, for which the optimal allocation of workers to reach Europe 2020 goals leads to higher economic growth than expected (for instance, Eastern countries, Belgium, Luxemburg, Germany show this result).

Only two countries show a significant negative deviation from the environmental goal D_2^- . The optimal allocation of workers provides the achievement of the environmental goals in Czech and Slovak Republics. These results can be explained by the transformations of national economic patterns following the collapse of USSR during the early nineties and the dissolution of Czechoslovakia in 1993. It illustrates the significant progress made in switching to a cleaner energy production system after the soviet period. A large number of European countries include in their optimal allocation a positive deviation D_2^+ , stressing that GHG emissions at the optimal allocation are significantly higher than the objective (especially in the United Kingdom, France, and Spain). The importance and the wide diffusion of these positive deviations outlines the need for an increased transformation of European production systems leading to lower environmental impact sectors, especially in industry and agriculture, for which these countries show higher GHG emissions than average.

Five countries (Cyprus, Czech and Slovak Republics, Romania and Sweden) show significant values in the deviation D_3^- . This result stands for a negative deviation from the energy consumption reduction goal which means that their optimal allocation goes beyond the required reduction of energy consumption. This result for the Czech and Slovak republics and in Romania might be justified again by their national economic pattern transformation from former soviet-linked countries into Eastern European countries.

Most countries in our sample show important and significant positive deviation D_3^+ from the energy production goal, outlining emergency and significant needs for a better energy efficiency, and development of renewable sources of energy, able to satisfy the Europe 2020 energy objective while keeping the other parameters – economic output, number of jobs, environmental impact – at least as good with respect to the sustainability objectives. Different measures may be implemented to reach these goals: for instance development of renewable energies, intensified use of nuclear energy, or waste reduction.

6 Conclusion and policy implications

The GP model allows policy makers to identify the best combination of investments and choices that optimizes the multiple and competing objectives that usually co-exist within a national strategy plan. To ensure that each member state tailors the Europe 2020 strategy in the most effective way, the European Commission has proposed that goals are translated into national targets and trajectories. The purpose of the European plan relies on the interactions among interrelated targets: indeed, investing in cleaner, low carbon technologies could help the environment, contribute to fighting climate change and create new business and employment opportunities in Europe. However, some stakeholders will enjoy benefits only in the long term, and that is why these changes meet resistance and conflicting objectives occur. Our implementation of a GP model for European countries, with respect to the Europe 2020 goals in reduction of GHG emissions and increase in energy efficiency, combined with economic output and employment constraints, outlines the need for a better integration and training of the

incoming workforce in a context of increasing immigration flows, jointly with deep and sustained transformations of economic national systems towards a more energy-efficient production system, combined with the development of renewable energies in many UE countries.

Improving our model to represent more efficiently the reality and support public policies implies to identify the utility payoffs attached to each deviation that goes beyond the object of this chapter. For future research it would be interesting considering that the utility, or preferences of the policy maker, can be taken into account by switching to a more sophisticated GP variant, where the appropriate coefficients for each sector of deviation are identified.

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Appendix A – Employment by main sector in EU (in thousands)

Countries	Agriculture, forestry and fishing	Energy industry	Manufacturing industry	Construction and residential	Trade, transports, distribution, repairing	Commercial services (Information, communication, financial and insurance activities)	Public, administrative services, technical, scientific, education, health, and social services
Austria	176.4	57.8	631.5	297.2	1,171.6	241.4	1,510.8
Belgium	59.6	53.1	502	266.2	974.5	236.7	2,311
Bulgaria³	56.72	32.57	536.22	147.16	793.70	79.41	237.15
Croatia⁴	36.23	14.74	254.85	102.3	415,009	38.41	133.697
Cyprus⁵	0 ⁶	0 ⁷	27.68	18.51	116,087	9.13	33.22
Czech Rep.	158.9	128	1367.9	411.2	1246.4	226	1,458.1
Denmark	68	25	284	173	722	179	1,225
Estonia	24.3	13.6	117.9	51.9	151.9	35.7	197.1
Finland	108.1	31.6	339.3	189.8	525.8	144.4	1,022.8
France	766	318	2,678	1,789	6,238	1,618	12,572
Germany	620	571	7,516	2,450	9,961	2,301	17,011
Greece	471.7	50.3	337.9	181.4	1,300.8	163.2	1,227.6
Hungary	273.8	82.1	775.6	269.2	1,020.2	211.6	1,488.6
Ireland	110.2	22.6	202.9	139	564	152.2	716.9
Italy	910.4	312.4	3,833.2	1,552.1	6,164.6	1,260.8	7,756.6
Latvia	71.6	21.5	118.5	69.6	245.5	43.1	277.5
Lithuania	120.9	26.9	202.7	104.8	358.4	45.8	419.8
Luxembourg	4.6	4.4	32.3	41.5	93.5	62.4	149.3
Malta⁸	0 ⁹	0 ¹⁰	0 ¹¹	9.99	57.72	6.91	26.28
Netherlands	192	67	767	457	2,19	503	4,249
Poland¹²	130.54	300.11	2,425.35	831.23	3,154.54	309.27	1,230.23
Portugal¹³	451.8	59.3	714.2	274	1,143.6	165.5	1,492.9
Romania¹⁴	85.466	123.239	1,180.1	365.32	1,384.74	157.36	44.20
Slovak Rep.	73.4	47.2	490.7	163.2	609	107.6	527.37
Slovenia	75.4	21.1	190.9	62.6	200.1	50.4	304.7
Spain	732	244.7	2057.7	1005.7	5,632.7	821.6	6,394.9
Sweden	112.3	66	580	340.4	981.5	270.7	2,233.3
UK	359.3	384.9	2,563	2,056.9	8,270.4	2,391.6	13,599.1

Source: Eurostat, 2015 (if not otherwise specified)

³ Eurostat, Persons employed by NACE Rev. 2, 2014 data.

⁴ Eurostat source, cf. supra.

⁵ Eurostat source, cf. supra.

⁶ Data mentioned as confidential – Eurostat source.

⁷ Data mentioned as confidential – Eurostat source.

⁸ Eurostat source, cf. supra.

⁹ Data mentioned as confidential – Eurostat source.

¹⁰ Data mentioned as confidential – Eurostat source.

¹¹ Data mentioned as confidential – Eurostat source.

¹² Eurostat source, cf. supra.

¹³ Same source, OECD estimations.

¹⁴ Eurostat source, cf. supra. 2014

Appendix B - Economic output (GDP) per economic sector in EU countries (US\$ millions)

Country	Agriculture, forestry and fishing	Energy industry	Manufacturing industry	Construction and residential	Trade, transports, distribution, repairing	Commercial services (Information, communication, financial and insurance activities)	Public, administrative services, technical, scientific, education, health, and social services
Austria	3,905.4	1,182.7	57,220.9	19,411.5	53,535.7	33,770	81,120.9
Belgium	2,745	233.5	52,417.2	19,956.6	65,047.5	38,275.1	133,183.7
Bulgaria	1,873.4	963.8	6,168.8	1,698.7	7,693.7	4,810.7	8,105.2
Croatia ¹⁵	353	10	738.2	555.8	2,880.5	2,669.4	4,680.4
Cyprus	1,504.3	819.2	5,320.6	1,905.3	5,829.3	3,983.6	8,478.5
Czech Rep.	3,785.4	1,364.2	40,392.3	8,571.2	24,746.5	14,110.5	32,263
Denmark	2,900.8	3,477.8	34,516.9	11,031.9	43,738.1	25,541.2	74,420.8
Estonia	592.3	250.9	2,770.9	1,092.3	3,447.6	1,744	4,373.8
Finland	4,443	551	30,652	11,481	25,800	15,614	54,647
France	33,854	2,148	218,987	106,155	290,404	184,239	698,443
Germany	17,351	4,158	622,608	124,755	387,625	242,562	800,412
Greece	6,385.6	812.2	14,708.1	3,696.8	28,141.9	12,763	39,879.3
Hungary	3,800.8	152.9	22,618.7	3,816.2	15,443.7	7,980.1	24,123.4
Ireland	2,398	898.8	87,448.2	6,024.7	26,020.7	34,131.9	51,905.5
Italy	33,158.7	5,470	232,882	70,099	248,975.1	137,076.5	390,895.1
Latvia	723.9	105	2,652.4	1,381.7	5,057.1	1,919.5	4,924.3
Lithuania	1,220.5	103.7	6,493.9	2,440.4	10,027.6	1,859.4	6,957.9
Luxembourg	112.9	31.3	2,461.8	2,340.4	7,538.8	15,240.8	12,295.6
Malta	103.6	0	728.2	345.6	1,361.7	1,081.8	2,426.3
Netherlands	10,965	12,573	71,120	28,048	116,686	74,264	221,426
Poland	9,921.8	6,170.6	75,087.1	29,881	91,052.8	30,637.6	86,313.4
Portugal	3,654.2	497.9	21,555.2	6,363.9	30,258.1	13,827.4	41,977.9
Romania	6,651.3	1,421	30,948.1	9,272.9	24,074.5	14,289.6	26,834.4
Slovak Rep.	794.3	127	7,737.8	1,826.3	6,032.5	2,782.6	8,815.3
Slovenia	2,600.5	362	15,940.8	5,577.4	9,519.4	5,927.7	14,916.6
Spain	25,004	2,110	138,914	54,554	163,329	78,783	265,940
Sweden	5,211.8	1,637	67,240	23,272.6	64,369.1	41,214.6	124,932.1
UK	14,981.3	23,174.5	224,465.4	141,518.8	357,250.9	315,841.9	708,262.2

Source: Eurostat data, NACE Rev. 2 classification, 2014, 2015.

Appendix C - Emissions of GHG¹⁶ in economic sectors in EU countries¹⁷

Countries	Agriculture, forestry and fishing	Energy industry	Manufacturing industry	Construction and residential	Trade, transports, distribution, repairing	Commercial services (Information, communication,	Public, administrative services, technical,
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¹⁵ Due to missing data, Croatia data is extracted from Eurostat 2014 panel data.

¹⁶ CO₂, CH₄, N₂O, HFCs, PFCs, SF₆, NO₄, NO_x, NMVOC, PFC, PM (2, 5 & 10); summed and transformed in CO₂ equivalent and thousands of tonnes.

¹⁷ United Nation – Framework Convention on Climate Change – Submitted National Communications, sixth edition, due 1st January 2014, data from 2011 in principle, in equivalent CO₂ emissions (Gg), data based on national accounts // Data confirmed in European Environment Agency Data // Eurostat – Air emissions national accounts by NACE Rev. 2 activity, 2014 data, in thousands of tonnes of CO₂ equivalent.

						financial and insurance activities)	scientific, education, health, and social services
Austria	7,577.10	13,988.43	26,499.85	380.91	7,669.89	223.77	1,640.62
Belgium	9,365.88	21,862.84	33,771.89	439.74	9,999.41	1,183.62	6,221.01
Bulgaria	7,510.28	28,950.92	5,877.21	56.7	4,830.7	50.59	393.02
Croatia	3,442.21	6,275.44	4,173.59	37.08	1,358.21	116.72	521.48
Cyprus	633	3,722	1,732.52	33.76	493.40	35.97	110.95
Czech Rep.	8,065	58,424	18,556.53	187.47	9,215.65	180.71	2,859.96
Denmark	9,671.85	19,747.81	5,580.18	103.51	39,777.66	117.05	1,180.64
Estonia	1,270.52	14,875.63	2,514.08	27.22	1,694.24	14.86	396.05
Finland	5,881.11	24,628.42	13,845.4	8,914	10,471.12	254.99	1,863.34
France	90,340	53,015.88	95,234.84	3,228.13	52,662.67	2,244.39	20,766.18
Germany	63,936	353,793	160,692.01	2,995.36	97,468.43	4,665.26	26,064.59
Greece	8,965.84	53,840.83	13,348.27	57.2	6,718.45	39.01	847.56
Hungary	5,925	17,459	9,699.31	100.21	5,387.57	561.29	2,264.7
Ireland	17,730	11,935	6,325.86	144.57	4,249.55	255.5	866.24
Italy	31,486	132,413	89,590.47	1,739.74	49,530.1	517.48	4,842.16
Latvia	2,456	2,084	1,342.62	18.84	2,509.77	14.25	290.3
Lithuania	3,756	4,467	4,823.88	14.3	7,480.83	18.38	218.76
Luxembourg	663	1,003	1,458.00	17.2	3,879.1	111.98	219.8
Malta	89	1,914	43.96	6.30 ¹⁸	3,443.38	12.95	61.88
Netherlands	18,097	63,000	42,131.76	941.46	32,802.36	635.74	8,057.37
Poland	29,930	174,672	62,271.56	548.26	33,550.04	1,430.56	9,534.95
Portugal	6,958	16,506	15,795.2	664.9	7,167.8	133.8	1,417.9
Romania	17,093	35,558	23,034.06	153.4	8,817.3	464.34	2,215.37
Slovak Rep.	2,935	9,280	17,598.00	16.65	4,884.97	5.02	888.43
Slovenia	1,691	6,359	2,289.37	50.13	4,011.70	8.68	220.23
Spain	36,262	87,124	75,513.46	957.57	37,214.06	617.25	4,408.9
Sweden	7,338	10,974	14,001.58	63.07	13,820.01	180.91	1,545.78
UK	43,735	178,850	83,588.16	3,020.55	94,951.03	1,523.58	19,002.70

Appendix D - Energy consumption¹⁹ in economic sectors in EU countries

Country	Agriculture, forestry and fishing	Energy industry ²⁰	Manufacturing industry	Construction and residential	Trade, transports, distribution, repairing	Commercial services (Information, communication, financial and insurance activities)	Public, administrative services, technical, scientific, education, health, and social services
Austria	539	1,935	7,517	5,627	8,197	565.2	2,260.8
Belgium	645	8,393	10,638	7,390	8,699	851.6	3,406.4
Bulgaria	191	518	2,622	2,165	2,946	185.2	740.8
Croatia	233	563	1,101	2,183	1,882	142.2	568.8
Cyprus	39	24	220	288	596	39.6	158.4
Czech Rep.	593	2,908	6,829	5,673	5,881	559.6	2,238.4

¹⁸ Eurostat estimate.

¹⁹ in thousands of tonnes of oil equivalent (ktoe) on a net calorific value basis.

²⁰ Non-energy use (consumption of raw materials).

Denmark	727	251	2,090	3,956	4,014	364.8	1,459.2
Estonia	132	113	566	888	741	91.8	367.2
Finland	718	1,150	10,325	5,064	4,136	574.6	2,298.4
France	4,516	14,003	25,883	37,318	43,544	4,206.6	16,814.4
Germany	9,386 ²¹	22,120	54,882	51,287	54,998	6,581.4	26,325.6
Greece	281	705	3,088	3,786	5,640	342.4	1,369.6
Hungary	599	1,676	3,923	4,431	3,910	448.8	1,795.2
Ireland	221	216	2,233	2,561	3,677	246.4	985.6
Italy	2,776	7,187	25,280	29,541	37,009	2,933.2	11,732.8
Latvia	153	89	791	1,239	987	121.8	487.2
Lithuania	105	1,070	971	1,400	1,644	118	472
Luxembourg	25	36	613	478	2,096	78.8	315.2
Malta	5	4	47	72	185	21.5	102.5
Netherlands	3,546	14,186	13,181	9,120	10,280	1,265.2	5,060.8
Poland	3,401	5,323	14,166	18,945	15,639	1,559	6,236
Portugal	424	1,435	4,393	2,569	5,439	380.8	1,523.2
Romania	422	1,564	6,204	7,390	5,290	353.6	1,414.4
Slovak Rep.	138	919	3,296	1,952	2,199	261.4	1,045.6
Slovenia	74	147	1,230	1,040	1,797	85.4	341.6
Spain	2,766	4,106	19,229	14,698	28,098	1,767.6	7,070.4
Sweden	377	1,969	10,757	6,633	7,732	883.4	3,533.6
UK	973	6,892	23,124	35,140	39,784	3,177.8	12,711.2

Source: International Energy Agency, 2014 data, Eurostat.

Appendix E - Europe 2020 GP goals

Goals	Economy	Environment		Energy		Labor (In thousands)	
Country	GDP 2020 Output, G ₁ ²²	GHG 1990 adjusted emissions ²³	GHG 2020 adjusted emissions objective G ₂ ²⁴	1990 adjusted level of consumption ²⁵	Europe 2020 objective adjusted consumption, G ₃ ²⁶	Total number of jobs in 2015	Total number of jobs in 2020 (G ₄)
Austria	25,3207.24	44,783.98751	37,318.4968	19,257.3521	16,047.1515	4086.7	4,127.7308
Belgium	219,980.146	104,774.2633	87,308.3936	36,863.2895	30,718.1791	4403.1	4,627.70235
Bulgaria	20,124.1921	86,528.42057	72,104.1329	17,070.5778	14,224.9125	1882.951	1,367.26662
Croatia	24,608.6206	13,146.9679	10,955.3684	6,995.8871	5,829.67272	995.229	811.482554
Cyprus	99,634.8986	4,698.694239	3,915.42191	938.4375	781.999969	204.633	247.772675
Czech Rep.	119,073.993	153,336.5209	127,775.323	34,344.7404	28,619.4722	4996.5	4,996.5
Denmark	153,603.942	88,577.76917	73,811.855	12,862	10,717.9046	2676	2,812.50289

²¹ State of the Art on Energy Efficiency in Agriculture, Country data on energy consumption in different agro-production sectors in the European countries, agree, FP7 Program of the EU, 2012.

²² Based on Eurostat data previously mentioned, combined with growth rate forecasts from IMF, European Commission, OECD for each year, in millions of US\$.

²³ In thousands of tonnes of CO₂ equivalent, adjusted to remain consistent and consider the representability of the global economic national patterns in respect to the GHG emissions criteria, Eurostat data.

²⁴ In thousands of tonnes of CO₂ equivalent, adjusted in the same way, Eurostat data.

²⁵ Eurostat data for Europe 2020 Statistics, 1990, in million tonnes of oil equivalent; adjusted data to remain consistent and consider the representability of the global economic national patterns.

²⁶ Europe 2020 objectives in energy sector deal with an increasing of energy efficiency by 20%, which mathematically implies a reduction of energy consumption by 16.66..% from the 1990 consumption to the horizon 2020 : assume that $X = a*Y$ where X is the energy consumption, a the efficiency and Y the total product, the possible uses of energy ; if we increase a by 20% we obtain $X' = 1,2*a*Y$, so $a*Y = X'/1,2$ which implies $(5/6)*X=a.Y$, so $X'=(1-5/6)*X$.

Estonia	83,171.9581	11,192.33462	9,326.57244	5,901.53571	4,917.74971	592.4	563.366506
Finland	820,160.365	48,751.5178	40,624.6398	21,492.7429	17,909.9026	2361.8	2,421.43841
France	2,190,576.46	371,294.9451	309,400.078	141,807.95	118,168.565	25979	30,116.7812
Germany	1,239,551.34	965,454.0876	804,512.891	236,892.325	197,402.374	40430	35,989.5619
Greece	103,397.869	86,248.26403	71,870.6784	14,426.8645	12,021.9062	3732.9	3,255.44646
Hungary	173,794.38	67,836.52567	56,528.1768	21,972.4803	18,309.6678	4121.1	3,360.23242
Ireland	692,546.877	39,259.86781	32,715.2478	6,853.88889	5,711.34561	1907.8	2,764.46669
Italy	602,875.178	381,169.0226	317,628.147	110,702.862	92,248.6952	21790.1	19,003.055
Latvia	32,592.0349	19,742.82409	16,451.6953	6,347.48718	5,289.36107	847.3	716.423094
Lithuania	23,923.4809	51,061.97919	42,549.9473	11,442.0408	9,534.65261	1279.3	1,070.55496
Luxembourg	37,281.0115	8,118.19132	6,794.88883	3,004.65	2,503.77485	388	465.291916
Malta	269,440.916	3,692.682825	3,077.1126	262.2	218.49126	100.9	111.401753
Netherlands	544,074.804	189,398.4085	157,825.694	51,490	42,906.617	10148	10,878.5305
Poland	209,057.592	387,013.087	322,498.005	63,467.7451	52,887.672	8381.264	8,091.99793
Portugal	137,296.491	44,726.91459	37,270.9379	12,174.1519	10,144.7208	4301.3	3,848.52231
Romania	70,362.1455	199,171.7597	165,969.827	42,563.6129	35,468.2586	3340.422	2,752.17912
Slovak Rep.	51,076.038	65,111.064	54,257.0496	14,912.72	12,426.7696	2018.469	2,048.92824
Slovenia	482,344.805	16,395.04576	13,661.9916	3,792.5	3,160.29025	905.2	923.449412
Spain	530,605.115	205,972.288	171,636.708	30,622.8788	25,518.0449	16889.3	16,805.0222
Sweden	1,069,709.51	61,950.71469	51,623.5306	58,353.6378	48,626.0864	4584.2	5,161.35035
UK	1,107,326.7	619,593.5555	516,307.31	129,764.154	108,132.47	29625.2	33,846.48

Appendix F – A_{ij} coefficients for the economic constraint (In thousands of US\$ per capita)

Country	Agriculture, forestry and fishing	Energy industry	Manufacturing industry	Construction and residential	Trade, transports, distribution, repairing	Commercial services (Information, communication, financial and insurance activities)	Public, administrative services, technical, scientific, education, health, and social services
Austria	22.1394558	20.4619377	90.6110847	65.314603	45.6945203	139.892295	53.6940032
Belgium	46.057047	4.39736347	104.416733	74.9684448	66.7496152	161.703	57.6303332
Bulgaria	33.0277494	29,5889233	11,5043192	11,542983	9,69341215	60,57748	34,1762278
Croatia	9.74357559	0,67847208	2,89660585	5,43325252	6,94081333	69,4993361	35,007517
Cyprus²⁷	-	-	192,218208	102,91131	50,214925	436,224266	255,222757
Czech Rep.	23.8225299	10,6578125	29,5286936	20,844358	19,8543806	62,4358407	22,1267403
Denmark	42.6588235	139.112	121.53838	63.7682081	60.5790859	142.688268	60.7516735
Estonia	24.3744856	18.4485294	23.5021204	21.0462428	22.6965109	48.8515406	22.1907661
Finland	41.1008326	17.4367089	90.3389331	60.4899895	49.0680867	108.130194	53.4288228
France	44.1958225	6.75471698	81.7725915	59.3376188	46.5540237	113.868356	55.5554407
Germany	27.9854839	7.28196147	82.8376796	50.9204082	38.9142656	105.415906	47.052613
Greece	13.5374179	16.1471173	43.5279669	20.3792723	21.634302	78.2046569	32.4855816
Hungary	13.8816654	1.86236297	29.1628417	14.1760773	15.1379141	37.713138	16.2054279
Ireland	21.7604356	39.7699115	430.991621	43.3431655	46.1359929	224.256899	72.4027061
Italy	36.4221221	17.5096031	60.7539393	0.04516397	40.3878759	108.721843	50.3951603

²⁷ Due to missing or confidential data, some coefficients for these countries were undetermined.

Latvia	10.1103352	4.88372093	22.3831224	19.8520115	20.5991853	44.5359629	17.7452252
Lithuania	10.0951199	3.85501859	32.0370005	23.2862595	27.9787946	40.5982533	16.5743211
Lux.	24.5434783	7.11363636	76.2167183	56.3951807	80.628877	244.24359	82.35499
Malta²⁸	-	-	-	34.5876701	23.5898413	156.623715	92.3355025
Neth.	5.7109375	187.656716	92.7249022	61.3741794	53.4032037	147.642147	52.1124971
Poland	76.0040753	20.5614702	30.9592842	35.9481056	28.8640225	99.0645684	70.1604906
Portugal	8.08809208	8.39629005	30.1809017	23.2259124	26.4586394	83.5492447	28.1183602
Romania	77.82393	11.5304409	26.2250254	25.3829519	17.3855993	90.8071834	607.126858
Slovak Rep.	10.8215259	2.69067797	15.7689016	11.1905637	9.90558292	25.8605948	16.7156204
Slovenia	34.4893899	17.1563981	83.5034049	89.0958466	47.5732134	117.613095	48.9550377
Spain	34.1584699	8.62280343	67.5093551	54.2448046	28.9965736	95.8897274	41.5862641
Sweden	46.4096171	24.8030303	115.931034	68.3683901	65.5823739	152.251939	55.9405812
UK	41.6957974	60.2091452	87.579165	68.8019836	43.1963267	132.063012	52.0815495

Appendix G – A_{ij} coefficients for the environmental constraint (In tonnes of CO₂ equivalent per capita)

Countries	Agriculture, forestry and fishing	Energy industry	Manufacturing industry	Construction and residential	Trade, transports, distribution, repairing	Commercial services (Information, communication, financial and insurance activities)	Public, administrative services, technical, scientific, education, health, and social services
Austria	42.9540816	242.01436	41.9633001	1.28165882	6.54650905	0.92695112	1.08592799
Belgium	157.145638	411.729567	67.2746833	1.65191961	10.2610713	5.0005112	2.69191129
Bulgaria	132.405063	888.801154	10.9605196	0.38526668	6.08626768	0.63702874	1.65720044
Croatia	95.012559	425.771084	16.3766647	0.36249707	3.2727218	3.03874092	3.90046897
Cyprus²⁹	-	-	62.5911127	1.82332289	4.25026919	3.93933421	3.33982541
Czech Rep.	50.7551919	456.4375	13.5657073	0.45590467	7.39381659	0.79959735	1.96142651
Denmark	142.233088	789.9124	19.6485282	0.59830636	55.0937161	0.6539162	0.96378857
Estonia	52.2847737	1093.79632	21.3237998	0.52454721	11.1536471	0.41627451	2.00936073
Finland	54.4043478	779.38038	40.8057619	46.9652266	19.9146482	1.76585873	1.8218068
France	117.937337	166.716604	35.5619279	1.80443432	8.44223549	1.38713597	1.65178086
Germany	103.122581	619.602452	21.3799912	1.22259592	9.78500452	2.02749544	1.53221974
Greece	19.0075048	1070.39423	39.5035987	0.31532525	5.16486009	0.23903186	0.69042033
Hungary	21.6398831	212.655298	12.505557	0.37224368	5.28089492	2.65258507	1.52136101
Ireland	160.889292	528.097345	31.1772351	1.04006475	7.53465957	1.67869251	1.20831915
Italy	34.5847979	423.857234	23.3722407	1.12089621	8.03460079	0.41043702	0.62426334
Latvia	34.301676	96.9302326	11.3301181	0.27066092	10.2231039	0.33055684	1.04611532
Lithuania	31.0669975	166.05948	23.798145	0.13640267	20.8728432	0.40135371	0.52110291
Lux	144.130435	227.954545	45.1394118	0.41445783	41.4877005	1.79450321	1.47220362

²⁸ Due to missing or confidential data, some coefficients for these countries were undetermined.

²⁹ Due to missing or confidential data, some coefficients for these countries were undetermined.

Malta³⁰	-	-	-	-	59.6524149	1.87548863	2.35491114
Netherl.	9.42552083	940.298507	54.9305841	2.06008096	15.0125195	1.26389066	1.89629889
Poland	229.273113	582.036287	25.6752885	0.65958235	10.6354683	4.62561718	7.75055518
Portugal	15.4006197	278.347386	22.1159339	2.42664234	6.26775096	0.80845921	0.94976221
Romania	199.99766	288.528794	19.5187671	0.4199031	6.36748251	2.95076956	50.1226046
Slovak Rep.	39.986376	196.610169	35.8630589	0.10201593	8.02129228	0.04663569	1.68465344
Slovenia	22.4270557	301.374408	11.9924987	0.80079872	20.0484908	0.1722619	0.72278635
Spain	49..5382514	356.044136	36.6979929	0.95213781	6.60678946	0.75128286	0.6894394
Sweden	65.3428317	166.272727	24.1406586	0.18527615	14.0804982	0.6683044	0.69214973
UK	121.722794	464.666147	32.6134042	1.46849774	11.4808266	0.63705344	1.39734997

Appendix H – A_{ij} coefficients for the energy constraint³¹

Countries	Agriculture, forestry and fishing	Energy industry	Manufacturing industry	Construction and residential	Trade, transports, distribution, repairing	Commercial services (Information, communication, financial and insurance activities)	Public, administrative services, technical, scientific, education, health, and social services
Austria	3..05555556	33.4775087	11.9034046	18.9333782	6.99641516	2.34134217	1.49642573
Belgium	10.8221477	158.060264	21.1912351	27.7610819	8.92662904	3.59780313	1.47399394
Bulgaria	3..36730017	15.9027415	4.88982052	14.7115783	3.71171117	2.3320825	3.12364279
Croatia	6.43131193	38.1979782	4.32018835	21.3400328	4.53484141	3.70225728	4.25439613
Cyprus³²	-	-	7.94797688	15.5557956	5.13408047	4.33639947	4.76821192
Czech Rep.	3.73190686	22.71875	4.992324	13.7962062	4.71838896	2.47610619	1.53514848
Denmark	10.6911765	10.04	7.35915493	22.867052	5.55955679	2.03798883	1.19118367
Estonia	5.43209877	8.30882353	4.80067854	17.1098266	4.87820935	2.57142857	1.8630137
Finland	6.64199815	36.3924051	30.4302977	26.6807165	7.86610879	3.97922438	2.24716465
France	5.89556136	44.0345912	9.66504854	20.8596982	6.98044245	2.59987639	1.3374483
Germany	15.1387097	38.7390543	7.30202235	20.9334694	5.5213332	2.86023468	1.54756334
Greece	0.59571762	14.0159046	9.13879846	20.8710033	4.33579336	2.09803922	1.11567286
Hungary	2.18772827	20.4141291	5.0580196	16.4598811	3.83258185	2.12098299	1.20596534
Ireland	2.00544465	9.55752212	11.0054214	18.4244604	6.51950355	1.61892247	1.3748082
Italy	3.04920914	23.0057618	6.595012	19.0329231	6.00347143	2.32645939	1.51262151
Latvia	2.13687151	4.13953488	6.67510549	17.8017241	4.0203666	2.82598608	1.75567568
Lithuania	0.86848635	39.7769517	4.79033054	13.3587786	4.58705357	2.57641921	1.12434493
Lux	5.43478261	8.18181818	18.9783282	11.5180723	22.4171123	1.26282051	2.11118553

³⁰ Due to missing or confidential data, some coefficients for these countries were undetermined.

³¹ In tonnes of oil equivalent.

³² Due to missing or confidential data, some coefficients for these countries were undetermined.

Malta³³	-	-	-	7.20576461	3.2049061	3.11278413	3.90074971
Netherl.	1.846875	211.731343	17.1851369	19.9562363	4.70480549	2.51530815	1.19105672
Poland	26.0527183	17.7371253	5.84080648	22.7916355	4.95761193	5.04091907	5.06897908
Portugal	0.93846835	24.1989882	6.15093811	9.37591241	4.75603358	2.30090634	1.02029607
Romania	4.93763602	12.6907878	5.2571905	20.2288405	3.82021725	2.24704821	32.000724
Slovak Rep.	1.88010899	19.470339	6.71693499	11.9607843	3.61083744	2.42936803	1.98267247
Slovenia	0.98143236	6.96682464	6.44316396	16.6134185	8.98050975	1.69444444	1.12110272
Spain	3.77868852	16.7797303	9.34489965	14.6146962	4.98837147	2.15141188	1.10563105
Sweden	3.35707925	29.8333333	18.5465517	19.4858989	7.87773816	3.26339121	1.58223257
UK	2.70804342	17.9059496	9.02223956	17.0839613	4.8104082	1.3287339	0.93470891

Appendix I – Model results: Significant deviations in the results of the GP model

	D_1^-	D_1^+	D_2^-	D_2^+	D_3^-	D_3^+	D_4^-	D_4^+
Austria	x			x		x		
Belgium		x				x		
Bulgaria		x						x
Croatia				x		x		x
Cyprus	x				x			
Czech Rep.	x		X		x			
Denmark		x		x		x		
Estonia				x		x		x
Finland	x			x		x		
France	x			x		x		
Germany		x				x		x
Greece		x		x		x		x
Hungary						x		x
Ireland	x			x		x		
Italy		x				x		x
Latvia								x
Lithuania		x				x		x
Luxembourg		x		x		x		
Malta	x			x		x		
Netherlands						x		
Poland		x				x		x
Portugal				x		x		x
Romania		x			x			x
Slovak Rep.	x		X		x			
Slovenia	x			x		x		
Spain		x		x		x		
Sweden	x				x			

³³ Due to missing or confidential data, some coefficients for these countries were undetermined.

UK		x		x		x		
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Appendix J – Model results: Detailed results of the GP model

	D_1^-	D_1^+	D_2^-	D_2^+	D_3^-	D_3^+	D_4^-	D_4^+
Austria	4,460,1580	0	0	20,677,980	0	10,695,990	0	0
Belgium	0	104,524,100	0	0	0	9,862,652	0	0
Bulgaria	0	15,466,390	0	0	0	0	0	843,812.3
Croatia	0	0	0	5,525,223	0	1,517,922	0	36,6813
Cyprus	7,892,4460	0	0	0	779,739,100	0	0	0
Czech Rep.	128,010,800	0	30,208,690	0	3,858,174	0	0	0
Denmark	0	50,315,840	0	2,490,308	0	2,306,434	0	0
Estonia	0	0	0	12,036,040	0	160,5637	0	1,439,487
Finland	614,629,900	0	0	25,281,090	0	477,1897	0	0
France	185,143,400	0	0	13,856,240	0	38,919,930	0	0
Germany	0	960,975,200	0	0	0	34,115,800	0	4,593,585
Greece	0	2,990,281	0	11,953,400	0	3,203,648	0	477,454
Hungary	0	0	0	0	0	4,439,730	0	3,559,426
Ireland	114,504,200	0	0	35,501,810	0	13,858,420	0	0
Italy	0	446,004,600	0	0	0	24,601,410	0	2,804,939
Latvia	0		0	0	0	0	0	560,002.7
Lithuania	0	5,675,707	0	0	0	1,462,300	0	339,897.1
Luxembourg	0	7,099,267	0	573,367.9	0	2027462	0	0
Malta	262,580,900		0	460,428.6	0	194,785.3	0	0
Netherlands	0	0	0	0	0	10,347,080	0	0
Poland	0	120,342,200	0	0	0	12,580,790	0	307,363
Portugal	0	0	0	11,565,290	0	6,551,841	0	682,089.1
Romania	0	46,271,050	0	0	9,368,044	0	0	860,759.8
Slovak Rep.	22,878,340	0	500,978,300	0	2,023,232	0	0	0
Slovenia	425,353,700	0	0	970,167.7	0	1,611,453	0	0
Spain	0	198,577,300	0	706,137,600	0	52,337,540	0	0
Sweden	656,521,500	0	0	0	14,326,920	0	0	0
UK	0	898,065,800	0	396,765,400	0	1,751,931	0	0