

# Income inequality and opportunity inequality in Europe

Daniele Checchi, University of Milan  
Vito Peragine, University of Bari  
Laura Serlenga, University of Bari

## Abstract

This paper proposes an estimate of the extent of opportunity inequality and of its determinants in a sample of European countries, based on the Survey on Income and Living Conditions conducted in 2005. Although the ranking among Northern European and Mediterranean countries is generally respected, our measures of equality of opportunity provide a different ranking with respect to the one offered by the measures of overall income inequality. Our figures show that equality of opportunity has correlation with both institutional measures of schooling (pre-primary education, de-tracked secondary school) as well as with labour market institutions (union density, but not employment protection).

JEL Classification: D31, D63, J62

Keywords: Inequality of opportunity, income inequality, intergenerational mobility.

## 1 Introduction

This paper proposes an estimation of the extent of opportunity inequality in Italy and in other 23 European countries. In Italy the theme is quite important, given the geographical differences in the measure inequality of opportunities between North and South of the country. The evidence we propose, in addition to be interesting *per se*, if one believes that equality of opportunity is the "right" theory of distributive justice, may also have an instrumental value: it might help to understand the genesis of standard income inequality; it may help to identify the priorities of a redistributive public intervention; finally, it could help identifying cases of inequality traps

(World Bank 2006) which, by preventing social groups from their full participation into economic and social life, might be partially responsible for the poor performance of some economies.

In this paper we refer to the conceptualization of EOp proposed by philosophers such as Dworkin (1981a,b), Arneson (1989), Cohen (1989) and explored within the economic literature - initiated by Roemer (1993, 1998) and Fleurbaey (1995) - that has flourished in the last two decades and that has analyzed different ways in which the concept of equality of opportunity may be translated in formal economic models (see Fleurbaey 2008 for a survey). More specifically, our contribution, which is both theoretical and empirical, is devoted to the issue of measuring opportunity inequality.

The contribution of this paper is twofold. First, we first provide a methodology to both measure opportunity inequality and decompose overall income inequality into two components: income inequality due to initial inequality in opportunities and income inequality due to individual responsibility. Secondly, we apply this methodology to measure opportunity inequality in Italy and in other 23 European countries and make an attempt to correlate the extent of opportunity inequality to institutional measures in such countries.

A common feature of the EOp literature is the basic idea that individuals' outcomes arise from two classes of variable: variables for which they should not be held responsible for (circumstances) and variables which belong to the sphere of individuals' responsibility (effort). Once this basic partition has been made, the concept of EOp can be decomposed into two distinct ethical principles: the Compensation Principle, which states that differences in outcomes due to circumstances are ethically unacceptable and should be compensated; the Reward Principle, which takes the view that differences due to effort are to be considered ethically acceptable and do not need any intervention.

The existing literature has developed different approaches to measuring opportunity inequality. The analysis start with partitioning the population in circumstance classes (types), where each class is formed by individuals endowed with the same set of circumstances: the income distribution within a circumstance class is interpreted as the opportunity set open to individuals in that class. Hence, in order to measure opportunity inequality, one focuses on the inequality between types inequality.

In some cases the existence of EOp in a given distribution is tested by checking for stochastic dominance between the types distributions, as in the studies by Lefranc et al. (2006a; 2006b) and Peragine and Serlenga (2008). Other studies propose opportunity-egalitarian social welfare functions to obtain partial rankings of income distributions (see Van de Gaer, 2003 and Peragine, 1998, 2004 on the theoretical side; Peragine and Serlenga 2008 for

an empirical application). Finally, some authors use inequality indices to obtain complete rankings of income distributions (see Bourguignon et al. 2003; Checchi and Peragine, 2009; Dardanoni et al, 2005; Ferreira and Gignoux, 2008; Pistoiesi, 2007). Studies which use this third methodology can be further distinguished depending on the empirical estimators they use. Bourguignon et al. (2003), Dardanoni et al.(2005) and Pistoiesi (2007) estimate EOp by using parametric models, while Checchi and Peragine (2009) use a non parametric method for their estimations. Recently, Ferreira and Gignoux (2008) compare parametric and non parametric methodologies, following the model proposed by Bourguignon et al. (2007). Each approach has pros and cons: the non parametric models avoids the arbitrary choice of a specific functional form on the relationship between outcome, circumstances and effort; on the other hand, parametric models allow to study partial effects of circumstances on outcome, other things constant (i.e. they make it easier building and studying counterfactuals).

In this paper we focus on complete rankings of opportunity inequality and explore both the parametric and the non parametric methodologies<sup>1</sup>. Furthermore, once an estimation of EOp measures is derived, we study the correlation between institutional characteristics and the opportunity inequality ranking for the countries under consideration.

The empirical application is therefore divided in two parts. First, along with the standard measures of inequality, we provide estimates of income inequality and opportunity inequality in 24 European countries available in the EU-SILC database. The purpose here is to rank European countries with respect to EOp by using both parametric and non parametric measures. Second, we focus on institutional characteristics that might influence the degree of opportunity inequality in the countries under analysis. Although the intuitive ranking among Northern European and Mediterranean countries is generally respected (with the formers exhibiting more EOp than the latter), our measures of equality of opportunities provide a different ranking with respect to the one offered by the measures of overall income inequality. Our figures also show that equality of opportunity is positively correlated with pre-primary education and de-tracked secondary school systems. Equality of opportunity is also positively associated to labour market regulation, to union density and to wage centralization whereas is positively related to fiscal redistribution. The paper is organized as follows. Section 2 introduces our methodology for measuring opportunity inequality and decomposing overall

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<sup>1</sup>In a companion paper (Checchi, Peragine, Serlenga 2009) we explore an alternative approach to measuring opportunity inequality, the *ex post* approach as opposed to the present *ex ante* approach, and we compare the results obtained with the two methodologies.

income inequality. Section 3 contains our empirical analysis: the data description, the estimating procedure and the discussion of the results. Section 4 concludes with some final remarks and some directions for future research.

## 2 Measuring opportunity inequality: a simple model

Each individual in our society is completely described by a list of traits, which can be partitioned into two different classes: the first class includes traits beyond the individual responsibility, represented by a person's set of circumstances  $\mathbf{c}$ . The individual sets of circumstances belong to a finite set  $\Omega = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}$ . Circumstances include a vast list of income generating inputs that are out of control of the individual, like gender, age, ethnicity, region of residence or parental background: various notions of equality of opportunity correspond to different choices of which of these variables are to be regarded as circumstances. In the sequel, on the basis of the data available, we will treat only gender and parental education as circumstances. If the only circumstances are gender, which can only take values in the set  $\{\text{male}, \text{female}\}$ , and parental education, that only takes values in the set  $\{\text{graduate parents}, \text{non graduate parents}\}$ ; in this case the set  $\Omega$  would be the following:

$$\Omega = \left( \begin{array}{cc} \{\text{female}, \text{non graduate parents}\}, \{\text{female}, \text{graduate parents}\} \\ \{\text{male}, \text{non graduate parents}\}, \{\text{male}, \text{graduate parents}\} \end{array} \right).$$

The second class includes factors for which the individual is fully responsible and is represented by a scalar variable, *effort*,  $e \in \Theta$ . We assume that effort is one-dimensional. It is important to stress that by effort in this paper we mean not only the extent to which a person exerts himself, but all the other background traits of the individual that might affect his success, but that are excluded from the list of circumstances. This amounts to say that any other factors, as native ability, talent, luck, and so on, are implicitly classified as within the sphere of individual responsibility. This assumption may lead us to overestimate the portion of inequality which is ethically acceptable<sup>2</sup>.

Income is generated by a function  $g : \Omega \times \Theta \rightarrow \mathbb{R}_+$ , that assigns individual incomes to combinations of effort and circumstances:

$$x = g(\mathbf{c}, e)$$

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<sup>2</sup>On the effect of partial observability of the circumstances on the estimates of opportunity inequality see Fleurbaey, Luongo and Peragine (2009).

To save notation, we may also write  $x(\mathbf{c}, e)$  and refer to it as both the individual income and the relevant income distribution.

Hence, this is a pure deterministic model, where for any given existing circumstances all variations in individual income are attributed to personal effort. We therefore deviate from standard Mincerian models of income generation, where incomes are explained by circumstances, proxies for effort and a random component which is typically assumed to be i.i.d.. In our analysis, the individual is held responsible for any random component that may affect his/her income (included native ability or talent, as long as they are not included in the set of circumstances).

Effort is unobservable. Unobservable is also the function  $g$ , hence we do not make any assumption about the degree of substitutability or complementarity among the circumstances in order to keep the approach as general as possible. We assume, however, that the function  $g$  is fixed and identical for all individuals. Moreover, we introduce two basic assumptions:

**Assumption 1** *The function  $g$  is monotonically increasing in effort  $e$*

**Assumption 2** *The distribution of effort  $e$  is independent of the circumstances.*

Assumption 1 is fairly reasonable. Assumption 2 appears to be more problematic, given the non observability of effort. From a theoretical point of view it would be hardly sustainable to hold people accountable for the factor  $e$ , were it dependent on external circumstances. However, from the empirical point of view, there are income determinants that are clearly the joint outcomes of effort and circumstances. Typical is the case of acquired education, which is the result of parental background (educated parents are typically richer in monetary and cultural resources) but also requires personal effort (in order to afford the psychological costs of studying). Since income is correlated with education, this would violate our Assumption 2. In such a case, we would be forced to extend the requirement of orthogonality between circumstances and effort to all these “intermediate” variables (where we could add labour market participation, fertility choices, migration, and similar). For this reason, we consider Assumption 2 to be the simplest property compatible with the empirical non parametric application that we adopt in the next section.

For  $\mathbf{c}_i \in \Omega$ , we call *type  $i$*  the set of individuals whose set of circumstances is  $\mathbf{c}_i$ . The type income distribution represents the set of outcome levels which can be achieved - by exerting different degrees of effort - starting from the same circumstance  $\mathbf{c}_i$ . That is to say, the type distribution is a representation of the *opportunity set* - expressed in outcome terms - open to any individual endowed with circumstances  $\mathbf{c}_i$ .

We propose the following definition of equality of opportunity.

**Definition** *There is EOp if the set of opportunities is the same, regardless of the circumstances. Inequality of opportunity decreases if inequality between individual opportunity sets decreases.*

Thus, this definition puts special emphasis on the differences in the outcome prospects for classes of individuals with identical circumstances. Accordingly, it focuses on inequality between types, and is instead neutral with respect to inequality within types.

Hence, in the model introduced the measure of opportunity inequality in a distribution is given by the degree of inequality between types. To capture such inequality we may construct an hypothetical *smoothed distribution* obtained after the following transformation:

$$x(\mathbf{c}, e) \rightarrow x(\mathbf{c}, \bar{e})$$

where  $x(\mathbf{c}, \bar{e})$  is the artificial distribution obtained by using a constant reference value of effort  $\bar{e}$ .

Hence, inequality of opportunity is given by a (scale invariant) inequality index  $I$  applied to the distribution  $x(c_i, \bar{e})$ :

$$OI = I(x(\mathbf{c}, \bar{e}))$$

or, in relative terms:

$$OI_{relative} = \frac{I(x(\mathbf{c}, \bar{e}))}{I(x(\mathbf{c}, e))}.$$

The meaning of  $x(c_i, \bar{e})$  depends on the specific measurement approach one decides to adopt.

In a non parametric descriptive approach  $x(c, \bar{e})$  can be represented by the average income<sup>3</sup> of a given type identified by  $\mathbf{c}$  (call it  $\mu_{\mathbf{c}}$ ). If we opt for a non parametric approach, then for any *path independent* measure of inequality<sup>4</sup> (Foster and Shneyrov, 2000) we have that

<sup>3</sup>Other interpretations are possible: for instance, in a normative approach,  $x(\bar{\mathbf{c}}, e)$  could be represented by the equally distributed equivalent income of a given tranche identified by  $e$ .

<sup>4</sup>In particular, we need to use the *mean logarithmic deviation (MLD)*, which is the only index which has a path-independent decomposition using the arithmetic mean as the representative income. For a distribution  $X = (x_1, \dots, x_N)$  with mean  $\mu_X$  the *MLD* is defined as:

$$MLD(X) = \frac{1}{N} \sum_{i=1}^N \ln \frac{\mu_X}{x_i}$$

$$I(x(\mathbf{c}, e)) = I\left(\frac{x(\mathbf{c}, e)}{x(\mathbf{c}, \bar{e})}\right) + I(x(\mathbf{c}, \bar{e}))$$

$$I(x(c, e)) = I\left(\frac{x(\mathbf{c}, e)}{\mu_c}\right) + I(\mu_c).$$

The interpretation is as follows: by measuring the inequality in the artificial vector  $\mu_c$ , obtained by replacing each income with its type mean income, we only capture the between-types inequality, which, in turn, reflects the opportunity inequality. On the other hand, by rescaling all type distributions until all types have the same mean income, hence obtaining the distribution  $\frac{x(\mathbf{c}, e)}{\mu_c}$ , we are left with an income vector in which the only inequality present is the within-types inequality, to be interpreted as inequality due to individual responsibility. Hence the decomposition above can be interpreted as:

total inequality = within types + between types

total inequality = effort inequality + opportunity inequality

Thus, we have a measure of opportunity inequality and a decomposition of overall inequality into an ethically acceptable and an ethically offensive part.

Inequality of opportunity by the ex ante approach can also be computed parametrically<sup>5</sup>. In this case we need to estimate the following relationship for the whole population such that

$$\ln x_i = \beta c_i + \varepsilon_i,$$

derive the following counterfactual distribution  $\hat{x}_i = \exp(\hat{\beta}c_i)$ . Hence, inequality of opportunity in the parametric case will be given by

$$I(\hat{x}_i).$$

In the following empirical analysis we will compare our estimates  $OI$  (parametric and non parametric) and use them for our an analysis of relationship between the extent of opportunity inequality and some relevant policy or institutional variables.

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<sup>5</sup>This is the approach followed by Ferreira and Guignoux (2008).

## **3 The empirical analysis: income inequality and opportunity inequality in Europe**

### **3.1 Data description**

We use data from the 2005 wave of the European Survey on Income and Living Conditions (EUSILC) which is annually conducted by the national Central Statistics Offices (CSOs) in order to obtain information on the income and living conditions of different household types. The survey contains information on a large number of individual and household characteristics as well as specific information on poverty and social exclusion. Representative random samples of households throughout a large number of European countries are approached to provide the required information. We consider 24 countries in our analysis, namely Austria, Belgium, Germany, Denmark, Estonia, Spain, Finland, France, Greece, Hungary, Ireland, Iceland, Italy, Luxemburg, Latvia, Lithuania, the Netherlands, Norway, Poland, Portugal, Sweden, Slovakia, Slovenia and the United Kingdom.

Differently from other sources of data, EUSILC provides a common data source with comparable individual and household level micro-data on income and living conditions in the EU countries, allowing for significant improvements in the comparability of country-specific measures. EUSILC is expected to become the EU reference source for comparative statistics on income distribution and social exclusion at European level. Indeed, our study has become possible because the 2005 EUSILC includes a special data module which provides information for attributes of each respondent's parents during her childhood period in the age 14-16. This additional module reports family composition, number of siblings, the educational attainment, occupational as well as the labour market activity status of each respondent's mother and father and the presence of financial problems in household. In what follows parental education is measured by the highest educational attainment in the couple of parents. Individuals are therefore divided in three groups: group 1 refers to individuals having both parents with no education; 2 corresponds to individuals who have at least one of the parents with primary or secondary (lower and upper) school degree, while group 3 corresponds to individuals who have at least one of the parents with post-secondary or tertiary degree. Parental occupation is also divided in three categories: category 1 corresponds to individuals having both parents occupied as plant and machine operator and assembler or in elementary occupations (groups 8000 and 9000 according to the ISCO88 classification); category 2 refers to individuals who have at least one of the parents occupied as service worker,



shop and market sales worker, skilled agricultural and fishery worker or as craft and related trades workers (groups from 5000 to 7000 of the ISCO88 classification); finally, category 3 refers to individuals who have at least one of the parents occupied as legislator, senior official, manager, professional, technician, associate professional or clerk (groups from 1000 to 4000 of the ISCO88 classification).

In the empirical analysis we also consider some additional individual characteristics as circumstances. This set comprehends gender, nationality (distinguishing those who declare the country of birth to be the same of the country of residence) and geographical location (distinguishing people living in densely populated area from others).

We restrict the sample to individuals working full-time or part-time, unemployed and those fulfilling domestic tasks and care responsibilities aged between 30 and 60.<sup>6</sup> Our reference variable is post-tax individual income which is available for 17 out of 24 countries under analysis, for the remaining ones we derived net income information from gross income by imputing the tax rate in 2004.<sup>7</sup> Being aware of the fact that welfare indicators estimated from micro-data can be very sensitive to the presence of extreme incomes (Cowell and Victoria-Feser, 1996a, 1996b, 2002) we censored the countries' income distributions by dropping the very extreme values.<sup>8</sup> Tables 9 and 10 show summary statistics of both individual and parental characteristics.

### **3.2 Income and opportunity inequality ranking in Europe**

In this section we aim to rank European countries with respect to EOp using both the parametric and non parametric approaches.

Starting with the estimates of overall income inequalities, we notice that the ranking based on calculation of Gini index from our data is quite consistent with the ranking provided by OECD and Eurostat (see Table 1 and Figure 1).<sup>9</sup> In particular our evidence shows that formerly planned economies

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<sup>6</sup>We exclude pupils, students, those in an unpaid work experience, those in retirement or in early retirement, permanently disabled or/and unfit to work, those in compulsory military community or service and other inactive person.

<sup>7</sup>Tables 4, 5, 6, 7 and 8 in the Appendix show the progressive tax rate used for the conversion. As for Slovakia we imputed a flat tax rate of 19% (source: <http://www.finance.gov.sk/>) whereas for Iceland a tax rate of 37.7% has been imputed for income higher than 1.191.000 ISK (source: <http://www.ministryoffinance.is>).

<sup>8</sup>Van Kerm (2007) discusses how ordinal comparisons of countries are found to be robust to variants of data adjustment procedures such as trimming and winsorizing.

<sup>9</sup>Spearman rank correlation between EUSILC Gini and the ones calculated by OECD and Eurostat are 0.90 and 0.84, respectively.

(Poland, Latvia, Lithuania, Estonia plus Portugal) obtain the highest values of both Gini and MLD. They are followed by the UK, Ireland and Mediterranean countries like Greece, Italy and Spain whereas Northern countries like Denmark, Finland, Norway and Sweden lead the ranking with low values of both Gini and Mean Logarithmic Deviation (MLD) (see Figure 2). Turning to the measurement of inequality of opportunity, our attention is confined to the MLD which is the only index that allows for a perfect decomposition of total income inequality in effort inequality and opportunity inequality. As a first insight we notice that inequality of opportunity generally accounts for a substantial share of income inequality in the EU countries under analysis, see Table (2). Notice that both the Netherlands and Norway have very few observations on parental socio-economic background, hence results in those cases might not be fully interpreted (see Tables 9 and 10). Summary statistics on the characteristics of the sample also show that we could not take into consideration information on degree of urbanization for the Netherlands and Slovenia whereas in the case of Sweden, given few observations available on parental occupation, we choose not to consider this characteristic.<sup>10</sup> Inequality of opportunity explains from the 2% to the 22% of income inequality. As mentioned in the previous section, given the partial observability of circumstances, those values can only be considered as lower bound estimates. Table (2) shows the ranking obtained by absolute measures, also reporting the incidence on total inequality).

It is interesting to note that country ranking based on inequality of opportunities does not coincide with the picture emerging when looking at total inequality (see Figure 3). Countries that were characterised by high levels of total inequality (Poland, Estonia, Lithuania and Latvia) show that only a small fraction of it was due to the effect of circumstances, indicating that most of it may be due to individual unobservable components (within which individual effort). A second pattern is observed among Nordic country (notably Sweden, Denmark and Finland, with Norway partially apart), which report low level of both total inequality and inequality of opportunities. Finally, most of the continental Europe and the Mediterranean area gather in an intermediate situation, with average inequality but high impact of circumstances. For these reasons we wonder whether these patterns could be associated to institutional features existing in these countries, which is the argument of the next section.

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<sup>10</sup>Few observations on parental occupation are also present in the UK case. However, in this case we considered both parental characteristics in the calculation as the results on the restricted sample are qualitative similar to those obtained using the larger sample with only education as parental characteristic.

### 3.3 Accounting for opportunity inequality

While searching for institutional features which may attenuate the impact of circumstances on income formation, we were induced to look at two passages which are crucial in individual careers: schooling and entrance in the labour market. As far as the first dimension, the literature suggests that early schooling may contribute to reducing the role of parental background in competence formation (Cunha and Heckman 2007). In addition, the stratification of the educational system may reinforce the impact of parents' education, since low educated parents may prevent their kids from aspiring to more academic oriented careers (Brunello and Checchi 2007). The quality of education may also play a role, since it may compensate the disadvantage of students coming from poor environment. Unfortunately, data on school quality are not easily available (unless one is ready to consider students achievements as a proxy for "revealed" quality). More modestly, we have considered economic resources publicly invested in the educational system as proxies for quality of education. In Figure 4 and Figure 5 we report the scatter plots of inequality of opportunities against some of these measures. In the left panel of Figure 4 we show that the ratio of enrolment in pre-primary education over primary enrolment (a rough indicator of pre-primary participation, in the absence of good quality data on the relevant age cohort for pre-primary education<sup>11</sup>) is negatively correlated with IO. Similarly, the right panel of the same Figure shows that a larger fraction of students enrolled in vocational education (another proxy for the extent of school stratification<sup>12</sup>) enhances the OI. When considering the resources available to the educational system, we have included in the left panel of Figure 5 the ratio of students to teachers in primary school (a negative proxy for resources - we were forced to use primary education, as it was the only variable which was non missing<sup>13</sup>), while in the right panel we have considered the incidence of

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<sup>11</sup>Source: OECD online database (<http://www.oecd.org/education/database>). Data are referred to 2002, and are obtained from Brunello and Checchi (2007). The enrolment in pre-primary schooling of 4-year-old children made available from Eurostat with reference to 2005 was discarded because it was reporting values at 100% for some countries, contradicting the figures contained in Education at a glance from OECD.

<sup>12</sup>Data from Eurostat 2005 ([http://epp.eurostat.ec.europa.eu/portal/page/portal/statistics/search\\_database](http://epp.eurostat.ec.europa.eu/portal/page/portal/statistics/search_database)). This indicator provides information on the percentage of boys and girls in upper secondary education who are enrolled in the vocational stream. It is indicative of the importance of initial vocational education and training in a country, taking into account also the gender dimension.

<sup>13</sup>Also from Eurostat 2005. The pupil-teacher ratio is calculated by dividing the number of full-time equivalent pupils by the number of full-time equivalent teachers teaching at ISCED level 1. Only teachers in service (including special education teachers) are taken into account.

public expenditure on gross domestic product.<sup>14</sup> In both cases, we show that greater resources invested in education reduce IO. However, we must recall that these are just simple correlations, which may be reflecting spurious correlation. If we use OLS regressions, including the absolute level of inequality as a sort of country control, we find that among this first group of variables related to schooling, only the share of students in vocational education comes out slightly significant (see Table 3).

Then let us move to labour market institutions. Here data availability, especially for new entrants in the EU, is scarce. We have considered two indirect measures of the degree of institutionalisation: the presence of union (proxied by union membership over dependent employment<sup>15</sup>) and the degree of employment protection, computed by OECD<sup>16</sup>. In accordance with the literature, we expected that when the labour market is heavily regulated, wages are less related to individual features, since unions press for job-related pay scales. In addition, employment protection reduces labour turnover, reducing individual income variability (and therefore aggregate wage inequality). Both measures have been proved to reduce total income inequality in the aggregate (Checchi and García Peñalosa 2008). However labour market institutions are only weakly and negatively correlated with IO, as it can also be seen in OLS regressions (see again Table 3).

Eventually, we have considered the role of welfare provisions. In general we do not have apriori theoretical expectation on their impact onto IO, since taxes and subsidies aim to contain income inequality (through taxation) and to provide income insurance against unforeseeable events (through subsidies), but in no case they include compensatory measures which attenuate the impact of circumstances. However, as long as fiscal redistribution sustains low incomes (that may be correlated to disadvantaged conditions), we could find some negative correlation with IO. We have selected two proxies for the welfare state, which are shown in Figure 7. In the left panel we have computed the ratio between the Gini index computed over gross incomes and the Gini index computed over disposable incomes: the larger is the

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<sup>14</sup>Also from Eurostat 2005. This indicator is defined as total public expenditure on education, expressed as a percentage of GDP. Generally, the public sector funds education either by bearing directly the current and capital expenses of educational institutions or by supporting students and their families with scholarships and public loans as well as by transferring public subsidies for educational activities to private firms or non-profit organisations. Both types of transactions together are reported as total public expenditure on education.

<sup>15</sup>Data were kindly made available by Jelle Visser (University of Amsterdam).

<sup>16</sup>It is the index of overall Employment Protection Legislation (version 2), referred to 2003 (OECD 2004).

ratio, the stronger is the redistributive role of the state.<sup>17</sup> We find a slightly negative correlation, which however does not reach statistical significance. More surprising is the right panel, which exhibits the scatter plot of IO against social expenditure.<sup>18</sup> In such a case we find a positive correlation with IO, as if a targeted expenditure may reinforce the income generating impact of circumstances. In both cases however the coefficients of these variables in multivariate regressions do not achieve statistical significance.

## 4 Concluding remarks

In this paper we have presented an approach to measuring inequality of opportunities based on the identification of circumstances, which are out of responsibility of the individual. We have made explicit assumptions, which are necessary to validate our proposed measure, both parametrically and not. We have then applied these methods to European countries, providing alternative country rankings. We show that standard income inequality and inequality of opportunities do not necessarily offer the same type of rankings. We have also searched for potential correlation with the is the most favourable institutional environment with respect to maximising EOp. Our results suggest that equality of opportunity is positively correlated with pre-primary education and de-tracked secondary school systems. Some correlation is also found for labour market regulation (namely with union density and EPL), while we find some counterintuitive result when considering social expenditure.

We would like to search for more robust results with respect to the role of institutions. But this require to move to structural model estimation, where through interactions with circumstances we may exploit temporal and cross-country variations in order to be able to assess the contribution of the institutional set-up to income formation. But this is left to future research.

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<sup>17</sup>Data are from the OECD database (<http://stats.oecd.org/Index.aspx?DataSetCode=INEQUALITY>) and are referred to mid-2000.

<sup>18</sup>>From Eurostat 2005: Expenditure on social protection contains: social benefits, which consist of transfers, in cash or in kind, to households and individuals to relieve them of the burden of a defined set of risks or needs; administration costs, which represent the costs charged to the scheme for its management and administration; other expenditure, which consists of miscellaneous expenditure by social protection schemes (payment of property income and other).

Table 1: Inequality of income and Comparable Gini Calculations

country	OECD Gini	EUROSTAT Gini	EUSILC Gini	EUSILC Mld
AT	0.27	0.26	0.275	0.164
BE	0.27	0.28	0.266	0.145
DE	0.3	0.26	0.29	0.185
DK	0.23	0.24	0.217	0.083
EE			0.344	0.243
ES	0.31	0.32	0.314	0.216
FI	0.27	0.26	0.271	0.136
FR	0.28	0.28	0.285	0.163
GR	0.31	0.33	0.316	0.2
HU	0.3	0.28	0.305	0.161
IE	0.32	0.32	0.296	0.187
IS	0.29	0.25	0.279	0.188
IT	0.35	0.33	0.309	0.197
LT			0.356	0.228
LU	0.26	0.26	0.276	0.148
LV			0.357	0.229
NL	0.27	0.27	0.27	0.184
NO	0.28	0.28	0.262	0.145
PL	0.38	0.36	0.364	0.271
PT	0.38	0.38	0.354	0.247
SE	0.24	0.23	0.231	0.106
SI			0.239	0.104
SK	0.27	0.26	0.278	0.132
UK	0.34	0.34	0.319	0.204

Notes: EUSILC Gini and MLD are given by authors' calculations; OECD Gini on working age population and Eurostat Gini are taken from <http://stats.oecd.org> and OECD (2008).

Table 2: Inequality of opportunity in absolute and as percentage of total inequality

country	ex ante absolute	ex ante relative
AT	0.037	0.396
BE	0.023	0.372
DE	0.03	0.27
DK	0.01	0.241
EE	0.021	0.272
ES	0.038	0.329
FI	0.012	0.169
FR	0.017	0.258
GR	0.026	0.27
HU	0.007	0.211
IE	0.032	0.337
IS	0.023	0.372
IT	0.024	0.305
LT	0.016	0.215
LU	0.026	0.311
LV	0.02	0.214
NL	0.033	0.38
NO	0.025	0.276
PL	0.017	0.207
PT	0.022	0.312
SE	0.011	0.189
SI	0.002	0.135
SK	0.014	0.242
UK	0.037	0.314

Notes: The columns refer to MLD in absolute terms and as percentage of total inequality as measured by MLD.

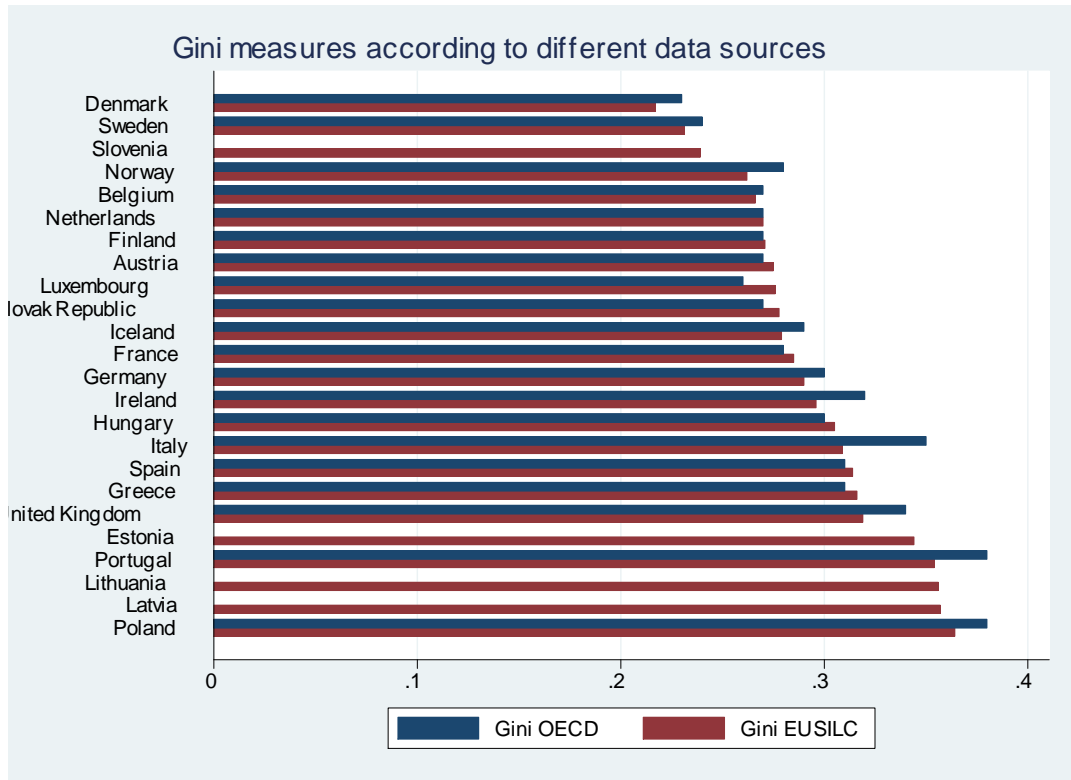
Table 3: Multivariate analysis of ex ante nequality of opportunity - OLS

enrolment in preprimary over primary	-0.0511			
	[0.78]			
	0.0014			
	[1.77]*			
expenditure in education/gdp	-0.0082			
	[0.90]			
pupil/teacher ratio in primary school	0.0039			
	[1.38]			
union density rate			-0.0003	
			[0.22]	
OECD employment protection legislation			-0.0086	
			[0.41]	
gini before/after tax and transfer				-0.0889
				[1.14]
expenditure in social protection/gdp				0.0028
				[1.42]
Observations	20	23	17	17
R-squared	0.21	0.11	0.02	0.2
Log likelihood	35.94	37.61	27.86	31.44

Notes: Robust t statistics in brackets - \* significant at 10%; \*\* significant at 5%; \*\*\* significant at 1% - Total inequality and constant added as controls.

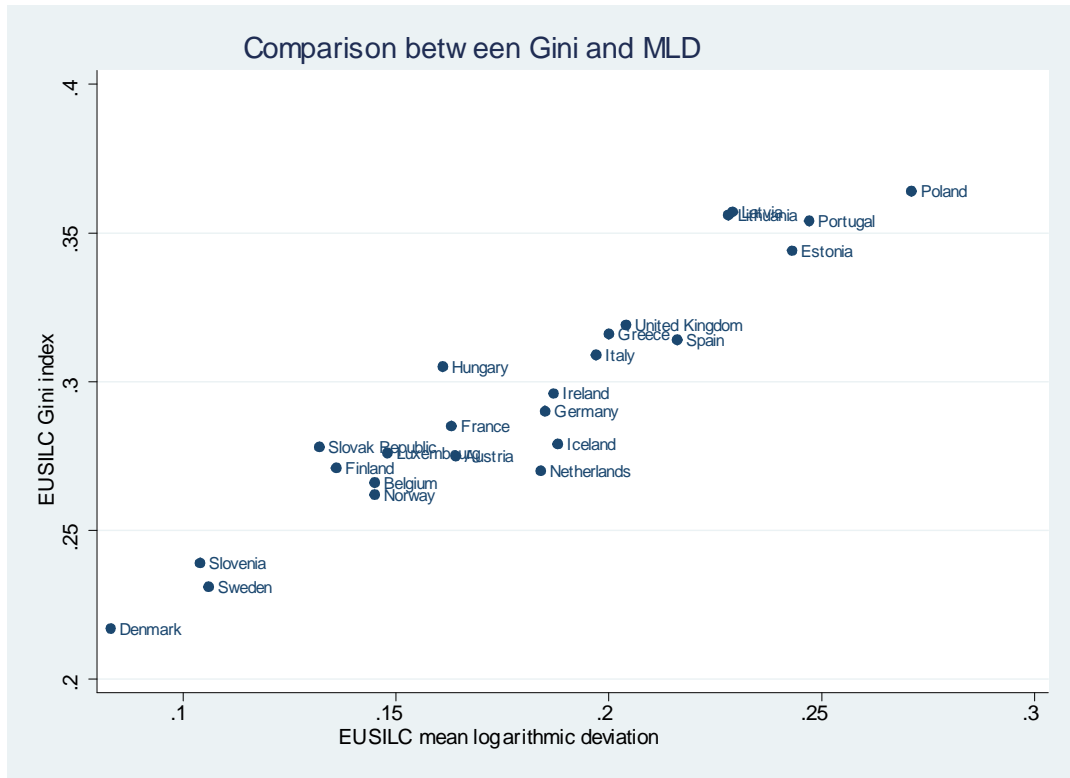


Figure 1 - Correlation between EUSILC Gini and OECD Gini on working age population



Sources: Authors's calculations and <http://stats.oecd.org>

Figure 2 - Correlation between EUSILC Gini and EUSILC MLD on working age population



Sources: Authors's calculations

Figure 3 - Opportunity inequality and total inequality



Sources: Authors's calculations

Figure 4 - Opportunity inequality and schooling

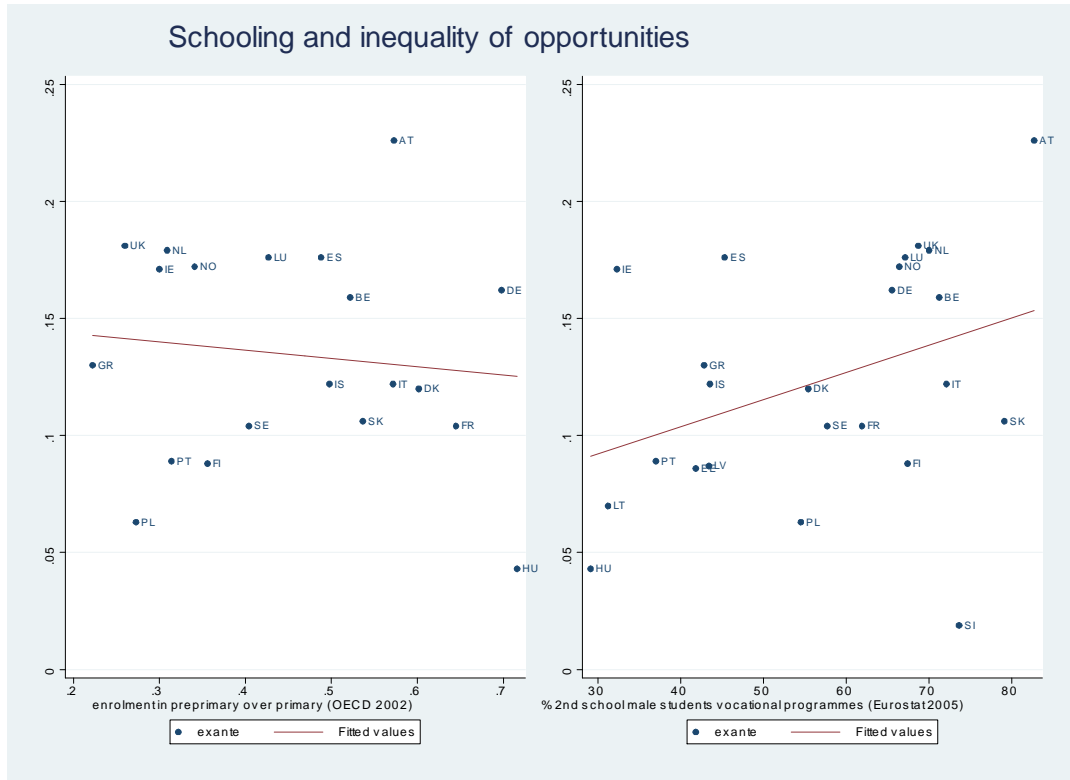


Figure 5 - Opportunity inequality and educational resources

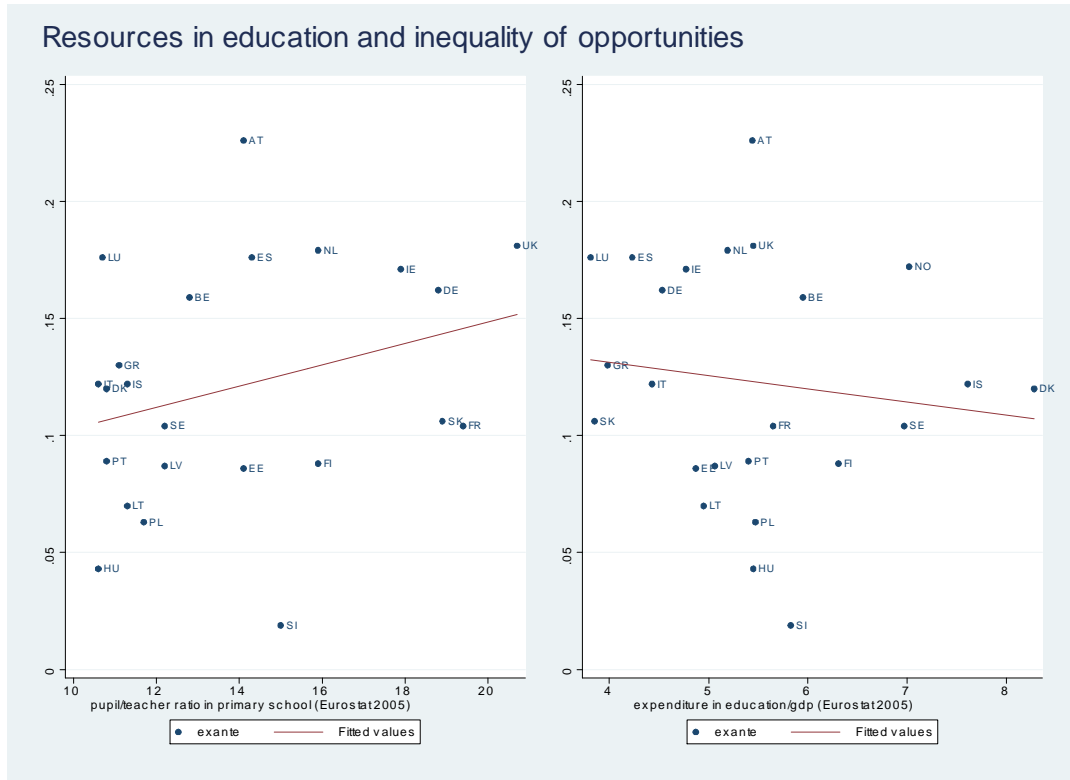


Figure 6 - Opportunity inequality and labour market institutions

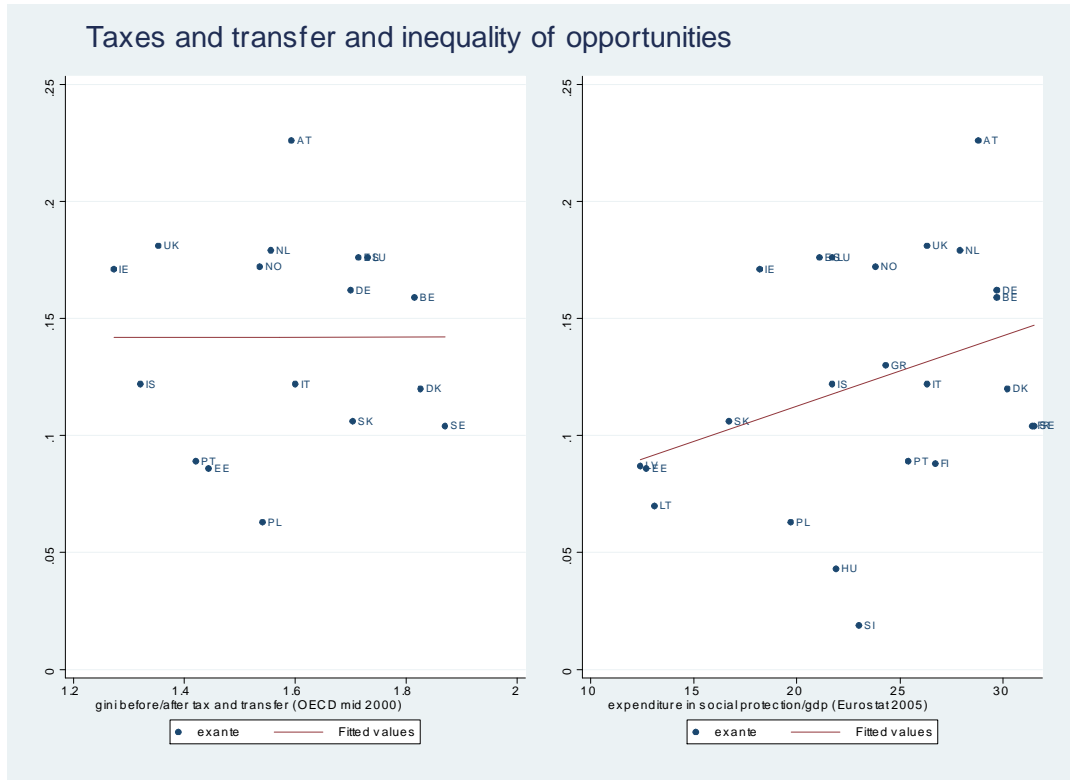
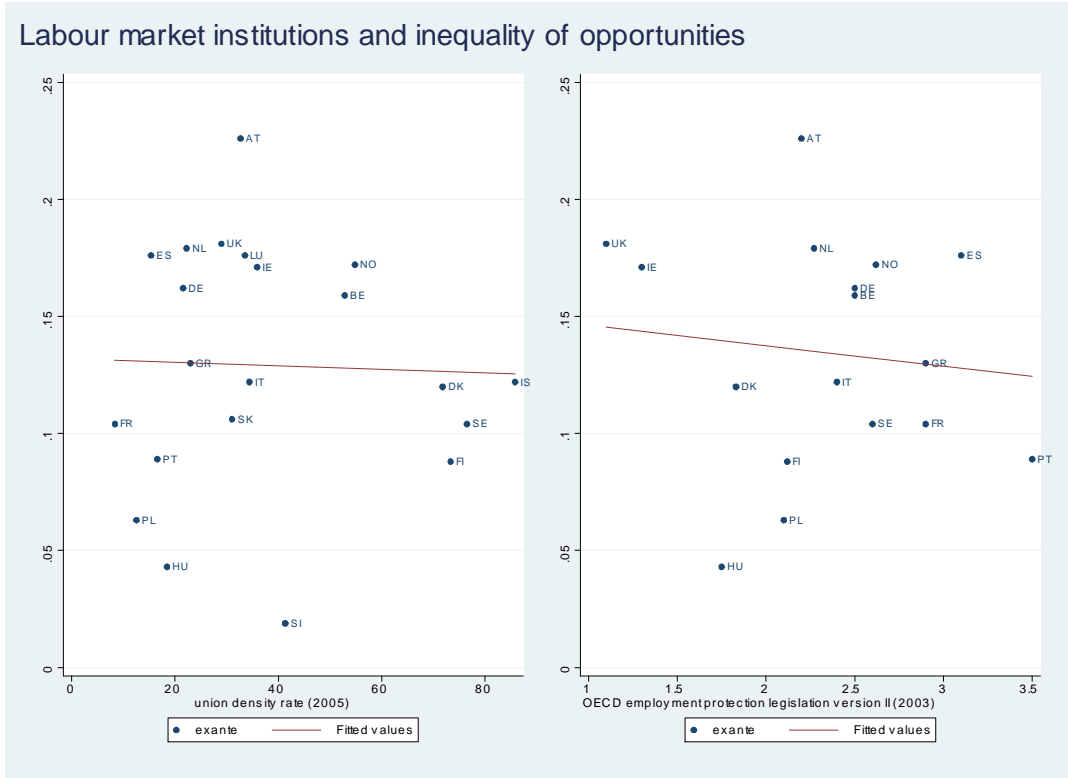


Figure 7 - Opportunity inequality and fiscal redistribution



## 5 Appendix

Table 4: Tax rate in Denmark

Taxable income in euro	tax rate
0 - 3250	0
32501 - 6500	13.3
65001 - 9750	19.2
97501 - 13000	24.1
130001 - 16250	27.6
162501 - 19500	28.7
195001 - 26000	30.4
260001 - 32500	32.7
325001 - 39000	34.5
390001 - 45500	36.2
455001 - 52000	38.9
520001 - 65000	42.3
650001 - 97500	48.4
975001 - 130000	53
130001	59.2

Source: [www.skm.dk](http://www.skm.dk).

Table 5: Tax rate in Finland

Taxable income in euro	tax rate
12200 - 17000	9
17001 - 20000	14
20001 - 32800	19.5
32801 - 58200	25
58201 -	32.5

Source: [www.vero.fi](http://www.vero.fi).



Table 6: Tax rate in Hungary

Taxable income in euro	tax rate
1 - 5960	18
5961 -	38

Source: [www.worldwide-tax.com](http://www.worldwide-tax.com).

Table 7: Tax rate in the Netherlands

Taxable income in euro	tax rate
1 - 16265	0
16266 - 29543	7.95
29544 - 50652	42
50653 -	52

Source: OECD.

Table 8: Tax rate in Norway

Taxable income in NOK	tax rate
lower limit	tax rate
0	0.0
29600 - 43022	25.0
43023 - 65999	7.8
66000 - 102580	35.8
102581 - 185160	27.1
185161 - 380999	35.8
381000 - 799999	47.8
800000	51.3

Source: [www.Taxnorway.no](http://www.Taxnorway.no).

Table 9: Summary statistics on main individual variables

country	m(x)	sd(x)	female	foreign	not densely pop	m(age)	sd(age)	education			N
								1	2	3	
AT	18237	8941	45.0	11.6	70.5	43.4	7.9		0.69	0.31	4256
BE	21310	12343	44.4	10.8	49.1	43.2	7.9	0.006	0.525	0.469	3601
DE	19508	9999	54.1	4.8	56.1	44.2	7.5		0.49	0.51	8296
DK	19601	8206	50.2	3.6	67.2	44.1	8.5		0.65	0.35	2796
EE	3807	2847	54.6	14.2	69.4	44.3	8.2		0.59	0.41	3730
ES	14062	8052	41.3	6.2	50.9	42.7	8.3		0.67	0.33	10012
FI	20915	10178	49.3	1.9	76.9	44.9	8.5		0.60	0.40	4535
FR	19368	10099	48.3	10.9	56.0	43.6	8.2		0.67	0.33	7164
GR	13870	8698	36.9	8.7	60.2	43.0	8.4		0.70	0.30	3980
HU	4353	4934	47.1	2.5	63.6	43.4	8.5	0.001	0.797	0.202	4591
IE	23938	12545	45.5	11.6	62.7	44.2	8.2		0.57	0.43	4072
IS	20466	10001	50.2	4.9	42.9	42.7	8.6		0.97	0.03	1248
IT	18654	13943	42.1	5.8	64.9	43.1	8.1	0.005	0.774	0.221	17151
LT	3022	2301	53.0	6.4	54.7	44.6	7.9		0.38	0.62	3659
LU	30034	14482	43.6	50.6	54.0	41.5	8.1		0.69	0.31	2498
LV	2871	2062	54.4	15.7	49.9	43.8	8.3	0.002	0.639	0.359	2706
NL	18198	8768	45.2	5.1		43.5	8.3		0.61	0.39	8441
NO	19809	9403	48.7	6.8	50.6	43.8	8.5	0.001	0.596	0.403	5674
PL	3690	2547	45.9	0.3	60.2	42.9	7.8	0.003	0.752	0.246	10347
PT	8997	6199	45.5	2.4	64.4	43.2	8.4		0.88	0.12	3375
SE	19453	8256	49.4	9.9	81.5	44.5	8.9		0.57	0.43	2435
SI	9557	4117	48.6	11.3		42.4	7.9		0.78	0.22	3011
SK	3748	3146	48.8	2.0	71.0	43.8	8.0		0.80	0.20	4461
UK	23861	13872	50.9	8.7	26.7	44.0	8.6		0.55	0.45	5421
Total	15129	11911	46.7	7.4	59.5	43.48	8.21	0.003	0.752	0.246	127460

The columns of this table show the following statistics: 1. average post tax individual income; 2 post tax individual income standard deviation; 3 percentage of females; 4. percentage of foreigners; 5. percentage of individuals not living in a densely populated area; 6. average age; 7 age standard deviation; 8 percentage of individuals with no education; 9 percentage of individuals with primary or secondary school degree ; 10 percentage of individuals with higher degree. 11 number of observations.

Table 10: Summary statistics on main parental variables

country	occupation			N	education			N
	1	2	3		1	2	3	
AT	0.59	0.32	0.09	3845	0	0.94	0.06	4025
BE	0.47	0.25	0.28	2995	0.11	0.67	0.22	3601
DE	0.52	0.23	0.25	7421	0	0.66	0.34	8296
DK	0.48	0.30	0.22	2649	0	0.78	0.22	2796
EE	0.75	0.10	0.15	3029	0.01	0.72	0.27	3730
ES	0.60	0.28	0.12	9496	0.20	0.69	0.11	10012
FI	0.52	0.32	0.17	3810	0.03	0.81	0.16	4535
FR	0.55	0.29	0.16	6562	0.04	0.86	0.10	7164
GR	0.35	0.51	0.14	3817	0.25	0.66	0.09	3980
HU	0.66	0.22	0.12	3986	0.00	0.84	0.16	4591
IE	0.47	0.14	0.39	2612	0.01	0.83	0.15	2792
IS	0.38	0.37	0.25	1083	0	0.76	0.24	1248
IT	0.58	0.24	0.19	15562	0.11	0.85	0.04	17151
LT	0.77	0.11	0.12	3170	0.05	0.69	0.26	3659
LU	0.55	0.23	0.22	2365	0.06	0.76	0.18	2498
LV	0.77	0.10	0.13	2104	0.02	0.77	0.21	2706
NL	0.39	0.20	0.41	3534	0	0.83	0.17	3866
NO	0.45	0.31	0.24	2519	0	0.58	0.42	2752
PL	0.53	0.36	0.11	9377	0.11	0.80	0.09	10347
PT	0.61	0.31	0.08	3057	0.35	0.61	0.03	3375
SE	0.50	0.29	0.21	535	0.01	0.79	0.20	2435
SI	0.61	0.27	0.11	2623	0.04	0.87	0.09	3011
SK	0.72	0.14	0.14	4070	0	0.90	0.10	4461
UK	0.62	0.24	0.13	2780	0.50	0.16	0.34	5421
Total	0.57	0.26	0.17	103001	0.10	0.75	0.15	118452

The columns of this table show the following statistics: 1. percentage of individuals who have at least one of the parents occupied in the categories 8000 and 9000 of the ISCO88 classification; 2. percentage of individuals who have at least one of the parents occupied in categories from 5000 to 7000 of the ISCO88 classification; 3. percentage of individuals who have at least one of the parents occupied in categories 1000 to 4000 of the ISCO88 classification (see the text for further details); 4. number of observations available for the parental occupation variable; 5. percentage of individuals who have both parents with no education; 6. percentage of individuals who have at least one of the parents with primary or secondary school degree; 7. percentage of individuals who have at least one of the parents with post-secondary or higher degree; 8. number of observations available for the parental education variable.

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