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Article

On the Decomposition of the Esteban and Ray Index by Income Sources

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Abstract: This paper proposes a simple algorithm based on a matrix formulation to compute the Esteban and Ray (*ER*) polarization index. It then shows how the algorithm introduced leads to quite a simple decomposition of polarization by income sources. Such a breakdown was not available hitherto. The decomposition we propose will thus allow one to determine the sign, as well as the magnitude, of the impact of the various income sources on the *ER* polarization index. A simple empirical illustration based on EU data is provided.

Keywords: decomposition; Esteban and Ray polarization index; income sources; polarization

JEL Classification: D31; D63; J31

1. Introduction

During the past 25 years, many studies attempted to measure the extent of the middle class and stressed the link between the concept of bipolarization and the importance of the middle class. Another strand of the economic literature emphasized the concept of polarization (or multi-polarization). The basic contribution here is that of [Esteban and Ray \(1994\)](#) who linked the concept of polarization to the notions of identification, alienation, and potential social conflict. Identification refers to the idea that an individual feels some degree of identification with those who are ‘close’ to him/her. Identification is thus an increasing function of the number of individuals who are in the same income class as that individual. The alienation function on the contrary characterizes the antagonism caused by income differences so that an individual will feel alienated from those who are ‘far away’ from him/her. While [Esteban and Ray \(1994\)](#), as well as [Esteban et al. \(2007\)](#), assumed that the number of groups was determined ex ante, [Duclos et al. \(2004\)](#) extended the analysis of polarization to the continuous case, letting the data determine the number of relevant groups and poles.

The focus of most empirical studies of bi-polarization and polarization was on the distribution of total income. There have however been a few attempts to decompose bipolarization and polarization indices by income sources (e.g., [Araar 2008](#); [Deutsch and Silber 2010](#)) but the procedures are not very simple. More recently, [Bárcena-Martín et al. \(2017\)](#) proposed a simple matrix formulation to decompose the Foster and Wolfson bi-polarization index by income sources.

The main contribution of the present paper is to introduce a simple algorithm to compute the [Esteban and Ray \(1994\)](#) polarization index. We derive this algorithm from the simple matrix formulation suggested by [Silber \(1989\)](#) to compute the Gini index. We then show that, with such an

approach, it is easy to derive the contribution of various income sources (or explanatory variables in the case of an earnings function) to the degree of polarization of the distribution of total income.

Section 2 describes the algorithm allowing the simple computation of the ER index polarization index while Section 3 shows how such a formulation simplifies the decomposition of this index by income sources. Section 4 presents a simple empirical illustration and Section 5 concludes the paper.

2. Matrix Representation of the Esteban and Ray (1994) ER Polarization Index

The Esteban and Ray (1994) polarization index ER is expressed as

$$ER = \sum_{i=1}^n \sum_{j=1}^n v_i^\beta v_j | \mu_i - \mu_j | \quad (1)$$

where v_k is the relative population frequency of population subgroup k , μ_k the mean income¹ of group k and β a parameter which varies between 2 and 2.6 (see, Esteban and Ray 1994).

We can also write expression (1) as

$$ER = \sum_{i=1}^n \mu_i v_i \left(\sum_{j=1}^{i-1} v_j^\beta - \sum_{j=i+1}^n v_j^\beta \right) + \sum_{i=1}^n \mu_i v_i^\beta \left(\sum_{j=1}^{i-1} v_j - \sum_{j=i+1}^n v_j \right) \quad (2)$$

where the mean incomes μ_i are ranked by increasing values.

More generally, assuming n population subgroups, expression (2) becomes

$$ER = t'Gs + v'Gr = ER^A + ER^B \quad (3)$$

In (3), ER^A and ER^B are the two components of the ER index, t' is a (1 by n) row vector, written as $t' = [v_1^\beta v_2^\beta \dots v_n^\beta]$, s is a (n by 1) column vector which, as row vector, would be written as $s' = [\mu_1 v_1 \mu_2 v_2 \dots \mu_n v_n]$, v' is a (1 by n) row vector written as $v' = [v_1 v_2 \dots v_n]$ and r is a (n by 1) column vector which, as a row vector would be expressed as $r' = [(\mu_1 v_1^\beta) (\mu_2 v_2^\beta) \dots (\mu_n v_n^\beta)]$. G is a square n by n matrix, called G -matrix, whose typical element g_{ij} is equal to 0 if $i = j$, to -1 if $j > i$ and to $+1$ if $i > j$ (see, Silber 1989, for more details on this G -matrix²). It is important to stress that the elements $\mu_i v_i$ in vector s' and the elements $(\mu_i v_i^\beta)$ in vector r' have both to be ranked by decreasing values of the mean incomes μ_i .

Let τ' be a (1 by n) row vector, written as $\tau' = \left[\left(\frac{v_1^\beta}{\sum_{i=1}^n v_i^\beta} \right) \dots \left(\frac{v_i^\beta}{\sum_{i=1}^n v_i^\beta} \right) \dots \left(\frac{v_n^\beta}{\sum_{i=1}^n v_i^\beta} \right) \right]$. Let also θ be a (n by 1) column vector of the income shares $\left(\frac{\mu_i v_i}{\sum_{i=1}^n \mu_i v_i} \right)$. In other words, if we call $\left(\frac{v_i^\beta}{\sum_{i=1}^n v_i^\beta} \right)$ the 'identification modified population share' of population subgroup i , the expression $\tau'G\theta$ is a kind of Gini index comparing a priori shares which are the 'identification modified population shares' with a posteriori shares which are the actual income shares of the various population subgroups, the comparison being made via the linear operator G , the G -matrix.

Similarly, let η' be a (n by 1) row vector whose typical element η_i is written as $\eta_i = \left(\frac{\mu_i v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right)$. η_i will be labeled the 'identification modified income share' of population subgroup i . The expression $v'G\eta$ is then a kind of Gini index, comparing a priori shares, the actual population shares, with a posteriori shares, the 'identification modified income shares' of the various population subgroups. This comparison is made again via the linear operator G , the G -matrix.

¹ Esteban and Ray (1994) refer to the natural logarithm of income rather than to income. We will make a somehow similar assumption by stating that the mean income of a given group refers in fact to its mean income relative to the mean income in the whole population. To simplify the notations, we do not introduce the population mean income in the formulations.

² As stressed already in Silber (1989), the first matrix formulation of the Gini index was proposed by Pyatt (1976).

Expression (3) is then rewritten as

$$ER = \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) [\tau' G \theta] + \left(\sum_{i=1}^n \mu_i v_i^\beta\right) [v' G \eta] \tag{4}$$

In other words, the polarization index is equal to the corrected sum of two Gini-related indices. The first one compares the ‘identification modified population shares’ with the actual income shares of the different population subgroups. The second one compares the actual population shares with the ‘identification modified income shares’ of the different population subgroups. The first correction factor is equal to the product $\left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right)$ while the second correction factor is equal to the product $\left(\sum_{i=1}^n \mu_i v_i^\beta\right) \left(\sum_{i=1}^n v_i\right) = \left(\sum_{i=1}^n \mu_i v_i^\beta\right)$.

3. Decomposing the ER Index by Income Sources

Assume there are J income sources. The average income μ_i , in population subgroup i , may then be expressed as

$$\mu_i = \sum_{j=1}^J \mu_{ij} \tag{5}$$

so that expression (3) may also be written as

$$ER = t' G \left[\sum_{j=1}^J s_{.j}\right] + v' G \left[\sum_{j=1}^J r_{.j}\right] \tag{6}$$

where $s_{.j}$ is a $(n$ by $1)$ column vector whose typical element s_{ij} is equal to $v_i \mu_{ij}$ while $r_{.j}$ is a $(n$ by $1)$ column vector whose typical element r_{ij} is equal to $v_i^\beta \mu_{ij}$. Note that the elements s_{ij} in vector $s_{.j}$ and the elements r_{ij} in vector $r_{.j}$ have to be ranked by decreasing mean incomes μ_i .

We may then rewrite (6) as

$$ER = \sum_{j=1}^J D_j \tag{7}$$

where D_j , the contribution of income source j to the ER index, is expressed as

$$D_j = [t' G s_{.j} + v' G r_{.j}] \tag{8}$$

We could also express (8) as

$$D_j = \left[t' G \tilde{s}_{.j} \frac{t' G s_{.j}}{t' G \tilde{s}_{.j}} + v' G \tilde{r}_{.j} \frac{v' G r_{.j}}{v' G \tilde{r}_{.j}} \right] \tag{9}$$

where $\tilde{s}_{.j}$ is a $(n$ by $1)$ column vector whose typical elements \tilde{s}_{ij} , which are equal to $v_i \mu_{ij}$, are ranked in descending order of μ_{ij} , while $\tilde{r}_{.j}$ is a $(n$ by $1)$ column vector whose typical elements \tilde{r}_{ij} , which are equal to $v_i^\beta \mu_{ij}$, are ranked also in descending order of μ_{ij} .

Note however that

$$(t' G \tilde{s}_{.j} + v' G \tilde{r}_{.j}) = ER_j^A + ER_j^B = ER_j \tag{10}$$

where ER_j is the Esteban and Ray polarization index for income source j , ER_j^A and ER_j^B being its two components.

Let us also define two correlation measures, COR_j^A and COR_j^B , with

$$COR_j^A = \frac{t' G s_{.j}}{t' G \tilde{s}_{.j}} \tag{11}$$

$$COR_j^B = \frac{v' G r_{.j}}{v' G \tilde{r}_{.j}} \tag{12}$$

These correlation measures may evidently be positive or negative.

Combining expressions (7)–(12) we derive that

$$ER = \sum_{j=1}^J \left\{ \left[ER_j^A COR_j^A \right] + \left[ER_j^B COR_j^B \right] \right\} \quad (13)$$

We therefore conclude that, *ceteris paribus*,

- The higher ER_j^A , the higher the degree of polarization of the distribution of total income.
- The higher ER_j^B , the higher the degree of polarization of the distribution of total income.
- If COR_j^A is positive, the higher this correlation measure, the higher the degree of polarization of the distribution of total income. However, if it is negative, it will have a negative impact on the overall Esteban and Ray index ER .
- Similarly, if COR_j^B is positive, the higher this correlation measure, the higher the degree of polarization of the distribution of total income. However, if it is negative, it will have a negative impact on the overall Esteban and Ray index ER ³.

4. A Short Empirical Illustration

In this section, we present a simple empirical illustration, based on EU data from the European Union Statistics on Income and Living Conditions (EU-SILC) data set for the 2016 wave (EUROSTAT 2016). EU-SILC is an international database that consists of comparable, country-specific data. We analyze polarization in the 17 countries with data available for 2016: AT (Austria), BE (Belgium), BG (Bulgaria), EE (Estonia), EL (Greece), ES (Spain), FR (France), HR (Croatia), HU (Hungary), LT (Lithuania), LV (Latvia), PL (Poland), PT (Portugal), RO (Romania), RS (Serbia), SE (Sweden), and SI (Slovenia). The units of analysis are the individuals and the unit of measurement is the household. The measure of income is the total disposable household income. Since a given level of household income corresponds to a different standard of living, depending on the size and composition of the household, we adjust incomes for differences in household size and composition using the “modified OECD” equivalence scale⁴. The latter assigns a value of 1 to the first adult in the household, 0.5 to each remaining adult, and 0.3 to each person younger than 14.

Disposable income includes net income from work, other private income not related to work, pensions and other social transfers. Net money income includes all income sources received by the household and by each of its current members in the year preceding the survey. Social insurance contributions, pay-as-you-earn taxes, and non-money income are not included in this definition of income.

The decomposition of the ER polarization index by income sources is based on three income sources:

1. Benefits (benefits) that include: old-age and survivor’ benefits, unemployment benefits, sickness benefits, disability benefits, education-related allowances, family/children related allowances, social exclusion not classified elsewhere, housing allowances
2. Income from rental of a property or land, interest, dividends, profit from capital investments in unincorporated business (property and interest)
3. Income available before including sources 1 and 2 (income before)

³ Expression (13) reminds us of the decomposition of the Gini index by income sources (see, [Lerman and Yitzhaki 1985](#)) where the contribution of an income source to the overall Gini index is a function of the share of this source in total income, of the Gini index of this source and of the Gini-correlation between this source and total income. In (13) the contribution of an income source to the overall ER index is a function of the two components of the ER index for this source, and of two correlation measures. However the share of the source does not appear. In Appendix A, we provide a more detailed decomposition where the parallel with the traditional decomposition of the Gini index by income sources becomes evident.

⁴ For a survey of equivalence scales and related income distribution issues, and some comparisons of scale relativities, see [Coulter et al. \(1992\)](#).

Table A1 in the Appendix A gives, for each of these countries, the average value of these income sources, the average total income and the population size.

Table 1 refers to data in Euros. We give there the value of the *ER* index when the parameter β is equal to 2.5 and when it is equal to 1 (Gini related measure⁵). We also computed, as suggested by Esteban and Ray (1994), the *ER* index with these two values of the parameter β , for the case where the logarithm of income rather than income was the variable under study. Table 1 gives also, when income and not the logarithm of income is used, the relative contributions of the different income sources, to the *ER* index. It appears that the most important (relative) contribution to the value of the *ER* index is that of income before transfer (62.4%) while this source has a share in total income of 70.7%. On the contrary, benefits and ‘property income and interest’ have a higher relative contribution to the *ER* index (respectively 25.4% and 12.2%) than their share in total income (23.2% and 6.1%). We may also observe that the contributions of these sources to the Gini-related index (parameter β equal to 1) is quite similar to their contributions to the *ER* index (65.6, 24.9, and 9.5%). They actually lie between their contributions to the average total income and to the *ER* index.

When introducing the logarithm of income into the formulation of the *ER* index with $\beta = 2.5$, we observe that this index is quite small (0.045) when compared to its value (0.577) when $\beta = 1$.

Table 1. Contributions of the income sources to the *ER* index (based on income data in Euros).

Measure Computed	Value for Total Income	Relative Contribution of Income Before	Relative Contribution of Benefits	Relative Contribution of Property Income and Interest
Average income with absolute contribution of income sources	15,634	11,060	3626	948
Average income with relative contribution of income sources	100%	70.70%	23.20%	6.10%
<i>ER</i> with parameter β equal to 2.5 computed on basis of relative incomes (relative contributions of income sources)	0.038	62.4%	25.4%	12.2%
<i>ER</i> with parameter β equal to 1 computed on basis of relative incomes (relative contributions of income sources)	0.645	65.6%	24.9%	9.5%
<i>ER</i> with parameter β equal to 2.5 and logarithms of incomes	0.045			
<i>ER</i> with parameter β equal to 1 (like Gini) and logarithms of incomes	0.577			

Table 2 is similar to Table 1 but here all the computations are derived from PPP income data. While the relative contributions of the three income sources to the average EU PPP income (on the basis of the countries for which data were available) are quite similar to those presented in Table 1, the computation of the *ER* index and of the contributions of the income sources to this index show a somehow different picture. When the parameter β is equal to 2.5, it appears that the *ER* index is lower than in Table 1, whether this index is derived from income data or from the logarithm of incomes. What is more interesting is that there is an important decrease in the relative contribution

⁵ When, in expression (1), we divide the income data by the average income and assume that $\beta = 1$, *ER* will equal to twice the traditional Gini index. What is called the absolute Gini index, is actually the product of the Gini index by the mean, so that when $\beta = 1$ and we use absolute incomes and not relative incomes in (1) *ER* will be equal to twice the absolute Gini index.

of income before transfer (from 62.4% to 54.4%) when $\beta = 2.5$ and from 65.6% to 59.6% when $\beta = 1$. On the contrary, there is an increase in the relative contribution of benefits: from 25.4% to 28.6% when $\beta = 2.5$ and from 24.9% to 27.7% when $\beta = 1$. A similar increase is observed for property income and interest since the relative contribution rises from 12.2% to 17.0% when $\beta = 2.5$ and from 9.5% to 12.8% when $\beta = 1$. In short, when using PPP rather than current data, polarization and inequality turn out to be smaller, but the relative contribution of benefits and property income and interest to polarization and inequality rises.

Table 2. Contributions of the income sources to the ER index (based on PPP income data).

Measure Computed	Value for Total Income	Relative Contribution of Income Before	Relative Contribution of Benefits	Relative Contribution of Property Income and Interest
Average income with absolute contribution of income sources	17,048	12,233	3892	924
Average income with relative contribution of income sources	100%	71.7%	22.8%	5.4%
ER with parameter β equal to 2.5 computed on basis of relative incomes (relative contributions of income sources)	0.024	54.4%	28.6%	17.0%
ER with parameter β equal to 1 computed on basis of relative incomes (relative contributions of income sources)	0.413	59.6%	27.7%	12.8%
ER with parameter β equal to 2.5 and logarithms of incomes	0.026			
ER with parameter β equal to 1 (like Gini) and logarithms of incomes	0.478			

5. Concluding Comments

This paper has shown how it is possible to express the Esteban and Ray (1994) ER index in matrix form. Such a formulation greatly simplifies the decomposition of this index by income sources. We gave a simple empirical illustration showing that this breakdown gives useful information as to the impact of the different income sources on the polarization of incomes. This illustration was based first on income data in Euros and then on PPP income data. We could also apply the proposed breakdown to an analysis of the polarization of the distribution of wages or earnings. If we estimate a traditional earnings function, we could then easily derive the contribution to the polarization of wages of the explanatory variables of such a function. Indeed, we intend to explore these issues in future empirical work.

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Appendix A. The Similarity between the Decomposition by Income Sources of the Gini Index and of the ER Index

Remember that expression (4) is written as

$$ER = \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) [\tau' G \theta] + \left(\sum_{i=1}^n \mu_i v_i^\beta\right) [v' G \eta] \tag{A1}$$

where τ' is a (1 by n) row vector, written as $\tau' = \left[\left(\frac{v_1^\beta}{\sum_{i=1}^n v_i^\beta} \right) \dots \left(\frac{v_i^\beta}{\sum_{i=1}^n v_i^\beta} \right) \dots \left(\frac{v_n^\beta}{\sum_{i=1}^n v_i^\beta} \right) \right]$, θ a (n by 1) column vector of the income shares $\left(\frac{\mu_i v_i}{\sum_{i=1}^n \mu_i v_i} \right)$, η' a (n by 1) row vector whose typical element η_i is written as $\eta_i = \left(\frac{\mu_i v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right)$ and v' a row vector of the population shares.

We can rewrite (A1) as

$$ER = \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) \left[\tau' G \left(\sum_{j=1}^J \theta_j \right) \right] + \left(\sum_{i=1}^n \mu_i v_i^\beta\right) \left[v' G \left(\sum_{j=1}^J \eta_j \right) \right] \tag{A2}$$

where

$$\theta_j = \left(\frac{\mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) = \left(\frac{\mu_{ij} v_i}{\sum_{i=1}^n \mu_{ij} v_i} \right) \left(\frac{\sum_{i=1}^n \mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) \tag{A3}$$

$$\eta_j = \left(\frac{\mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right) = \left(\frac{\mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) \left(\frac{\sum_{i=1}^n \mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right) \tag{A4}$$

Given that the G -matrix is a linear operator we then derive that

$$\begin{aligned} ER &= \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) \left[\sum_{j=1}^J \tau' G \theta_j \right] + \left(\sum_{i=1}^n \mu_i v_i^\beta\right) \left[\sum_{j=1}^J v' G \eta_j \right] \\ \Leftrightarrow ER &= \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) \left[\sum_{j=1}^J \tau' G \left\{ \left(\frac{\mu_{ij} v_i}{\sum_{i=1}^n \mu_{ij} v_i} \right) \left(\frac{\sum_{i=1}^n \mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) \right\} \right] + \\ &\quad \left(\sum_{i=1}^n \mu_i v_i^\beta\right) \left[\sum_{j=1}^J v' G \left\{ \left(\frac{\mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) \left(\frac{\sum_{i=1}^n \mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right) \right\} \right] \\ \Leftrightarrow ER &= \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) \left\{ \left[\sum_{j=1}^J \left(\frac{\sum_{i=1}^n \mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) \tau' G \left(\frac{\mu_{ij} v_i}{\sum_{i=1}^n \mu_{ij} v_i} \right) \right] \right\} + \\ &\quad \left(\sum_{i=1}^n \mu_i v_i^\beta\right) \left\{ \left[\sum_{j=1}^J \left(\frac{\sum_{i=1}^n \mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) v' G \left(\frac{\mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) \right] \right\} \end{aligned}$$

If instead of ranking the incomes μ_{ij} by decreasing values of the incomes μ_i , we rank them by decreasing values of the incomes μ_{ij} , and call $\tilde{\mu}_{ij}$ this re-ordered vector, we end up with

$$\begin{aligned} ER &= \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) \left\{ \left[\sum_{j=1}^J \left(\frac{\sum_{i=1}^n \mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) \left[\tau' G \left(\frac{\tilde{\mu}_{ij} v_i}{\sum_{i=1}^n \mu_{ij} v_i} \right) \right] \right] \left(\left(\frac{\mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) / \left(\frac{\tilde{\mu}_{ij} v_i}{\sum_{i=1}^n \mu_{ij} v_i} \right) \right) \right\} \\ &\quad + \left(\sum_{i=1}^n \mu_i v_i^\beta\right) \left\{ \left[\sum_{j=1}^J \left(\frac{\sum_{i=1}^n \mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right) \left[v' G \left(\frac{\tilde{\mu}_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) \right] \right] \left(\left(\frac{\mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right) / \left(\frac{\tilde{\mu}_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) \right) \right\} \\ \Leftrightarrow ER &= \left(\sum_{i=1}^n v_i^\beta\right) \left(\sum_{i=1}^n \mu_i v_i\right) \sum_{j=1}^J \alpha_j \beta_j \gamma_j + \left(\sum_{i=1}^n \mu_i v_i^\beta\right) \sum_{j=1}^J \lambda_j \nu_j \rho_j \end{aligned}$$

with

$$\begin{aligned} \alpha_j &= \left(\frac{\sum_{i=1}^n \mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) \\ \beta_j &= \tau' G \left(\frac{\tilde{\mu}_{ij} v_i}{\sum_{i=1}^n \mu_{ij} v_i} \right) \\ \gamma_j &= \left(\left(\frac{\mu_{ij} v_i}{\sum_{i=1}^n \mu_i v_i} \right) / \left(\frac{\tilde{\mu}_{ij} v_i}{\sum_{i=1}^n \mu_{ij} v_i} \right) \right) \end{aligned}$$

$$\lambda_j = \left(\frac{\sum_{i=1}^n \mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_i v_i^\beta} \right)$$

$$v_j = v' G \left(\frac{\tilde{\mu}_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right)$$

$$\rho_j = \left(\left(\frac{\mu_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) / \left(\frac{\tilde{\mu}_{ij} v_i^\beta}{\sum_{i=1}^n \mu_{ij} v_i^\beta} \right) \right)$$

where α_j and λ_j are similar to income shares, β_j and v_j are components of the *ER* index for income source j and γ_j and ρ_j are 'correlation measures'.

In other words, we have here quite a similar decomposition to that proposed by Lerman and Yitzhaki (1985) for the Gini index.

Table A1. Database.

Country	Mean Total Income	Income Before %	Benefits %	Property and Interest %	Total Population
AT	26,662.48	70.0%	27.3%	2.7%	7,963,391
BE	24,520.2	74.0%	24.6%	1.3%	9,319,177
BG	4164.49	76.4%	21.9%	1.7%	6,235,715
EE	11,043.97	82.3%	16.6%	1.1%	1,113,681
EL	9161.76	75.7%	19.8%	4.5%	8,092,137
ES	16,370.34	71.9%	24.4%	3.6%	42,446,793
FR	25,730.84	65.1%	24.4%	10.5%	55,793,599
HR	6663.01	80.3%	18.3%	1.4%	3,225,726
HU	5474.74	75.6%	23.2%	1.2%	8,332,493
LT	7742.34	81.6%	16.6%	1.8%	2,417,930
LV	8135.06	80.3%	18.6%	1.2%	1,708,676
PL	6912.37	80.9%	18.2%	0.9%	32,623,207
PT	10,892.61	79.1%	17.7%	3.2%	8,183,986
RO	2850.82	82.4%	17.5%	0.1%	15,991,057
RS	3214.21	74.2%	25.0%	0.8%	5,432,579
SE	29,761.2	77.6%	17.6%	4.8%	7,647,944
SI	13,678.07	73.8%	23.5%	2.7%	1,794,388

Country codes: AT (Austria), BE (Belgium), BG (Bulgaria), EE (Estonia), EL (Greece), ES (Spain), FR (France), HR (Croatia), HU (Hungary), LT (Lithuania), LV (Latvia), PL (Poland), PT (Portugal), RO (Romania), RS (Serbia), SE (Sweden), SI (Slovenia).

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