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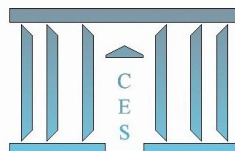
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Multidimensional Pigou-Dalton Transfers And Social Evaluation Functions

Marcello Basili

DEPS and SEM, University of Siena

Paulo Casaca*

UFMG, Universidade Federal de Minas Gerais

Alain Chateauneuf†

IPAG Business School and PSE-CES, University of Paris-I

Maurizio Franzini

DIPECODIR, University of Rome La Sapienza

Abstract

We axiomatize, in the multidimensional case, a social evaluation function that can accommodate a natural Pigou-Dalton principle and correlation increasing majorization. This is performed by building upon a simple class of inframodular functions proposed by Müller and Scarsini under risk.

Keywords: multidimensional inequality, Pigou-Dalton transfer, increasing majorization, inframodular functions, Human Development Index.

*Corresponding Author. Université Paris 1 Panthéon-Sorbonne. 106 - 112 Boulevard de l'Hôpital, 75013 Paris, France; e-mail: paulorscasaca@yahoo.com.br

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1 Introduction

There has been a resurgence of interest in multidimensional social evaluation functions mainly due to new techniques that extend in the multidimensional setting the pioneering works by Atkinson (1970; 1987), Kolm (1976a;b; 1977) and Sen (1976). In particular, Tsui (1995; 1999) and Gajdos and Weymark (2005) have offered axiomatic approaches to designing income inequality measures in a multiattribute context.

Although Tsui mainly used the additive approach, Gajdos and Weymark built upon the generalized Gini social function. These two different approaches are not at all innocuous. The former aggregates the attributes of each individual and then additively aggregates the resulting values; the latter evaluates the different attributes through a specific aggregation and then simply aggregates the values.

Note also that although Tsui's (1999) approach is a 'traditional' additive evaluation, Gajdos and Weymark (2005) adopted a non-additive approach, which was introduced by Weymark in his seminal paper in 1981.

In this paper, we follow the traditional additive approach, but instead of imposing the majorization theory of the m -dimensional case as in Tsui (1995), we confine ourselves to accommodating a particular Pigou-Dalton transfer, which we believe is relevant. Furthermore our approach is consistent with the meaningful property of correlation increasing majorization (e.g. Tsui, 1999).

The paper is organized as follows. *Section 2* analyzes the related literature. *Section 3* presents our motivations. *Section 4* states the framework and the axioms. *Section 5* offers our main Theorem, specifically the *Theorem 2* which characterizes our social evaluation function. *Section 6* aims at reducing the number of parameters, namely, at specifying some fundamental ψ function, for our purpose. Thus we deliver a tractable relative inequality index in the *Corollary* of *Theorem 3*. *Section 7* shows that our evaluation function can accommodate correlation increasing majorization. *Section 8* evaluates the appliance of our inframodular social evaluation, by comparing it with the famous Human Development Index (HDI) by using effective data. *Section 9* concludes and proofs can be found in the Annex.

2 Related literature

In the last two decades, consensus has been emerging among many scientists, particularly economists, about the multidimensional aspect of individual well-being that cannot be reduced to a unique monetary dimension. The origin of the multidimensional approach to poverty and/or inequality can be traced to the works of Rawls and Sen on ethical principles. In this perspective, as an example, health and education are important dimensions of individual well-being to be considered along with income. Crucially, many additional aspects of well-being may not be compared on a true cardinal basis; nonetheless, the literature assumes that all attributes have a cardinal meaning (Allison and Foster, 2004).

Unidimensional methods of evaluating poverty and inequality are applied when a single well-defined dimensional variable, i.e., income, has been selected. The selected variable can be cardinal, more often, or ordinal, and in this unidimensional environment, rank is defined regarding the minimum level, named the poverty line, below which a person is considered poor. If the assumed notion of poverty (inequality) is considered a useful proxy (unidimensional lens) that emerges after taking into account (and merging) different dimensions, then a unidimensional methodology is coherent and consistent.

On the contrary, if an aggregate variable cannot be constructed because there is a set of non-cardinal and non-summable attributes (health, education, talent, capabilities, etc.) in addition to income or consumption, then identification and measurement problems emerge. Identification means, for example, that is necessary to define the conditions under which a person is considered poor, and a measure must evaluate how much poverty there is overall (Sen, 1976). Poverty and inequality, like development, are multidimensional. In the economic literature, for ranking individuals who differ in attributes, there are two main approaches: The first uses a social welfare function (i.e. Atkinson and Bourguignon, 1982); the second extends the Foster et al. (1984) class of indices (decomposable poverty measures based on powers of normalized shortfalls) by using inequality statistics and measuring individual attributes with a utility function, so that the resulting univariate distribution vector of utilities is valued through an inequality index. Atkinson and Bourguignon (1982) attach to the attributes a symmetrical role; in fact, they introduce the crucial idea of complementarity or substitutability between attributes, which may be expressed as a social taste either for increasing correlation or for decreasing correlation. They study the correspondence between stochastic dominance conditions and the welfare interpretation of the value judgments. Interesting enough, comparing distributions according either to an additive separable social welfare function with concave utilities or to the sequence of Pigou–Dalton progressive transfers leads to the same conclusion as resorting to Lorenz curves or computing the amount of aggregate poverty gaps. In a seminal paper, Maasoumi (1986) introduces a two-stage approach for defining the class of generalized entropy measures.

More interesting for our research, in 1995 Tsui introduces a multidimensional generalization of the Atkinson-Kolm-Sen approach to measure inequality. Tsui defines a set of axioms “which are generalizations of their counterparts in the unidimensional context and are often considered to have a higher degree of acceptability.” (1995; p. 254). In this way, he distinguishes absolute and relative multidimensional equality indices based on a social evaluation function. Gajdos and Weymark (2005) follow Tsui’s general approach and obtain an univariate distribution vector of utilities that is valued using an inequality index. In particular, Gajdos and Weymark extend generalized Gini social evaluation functions to the multivariate case with a comonotonic independence axiom, such that “two allocations are said to be comonotonic if all individuals are ranked identically in all attributes (i.e., the richest is also the most educated etc.), and the ranking between two comonotonic allocations is not reversed by the addition of a comonotonic allocation.” (Galichon and Henry, 2012; p. 1513). In 2006,

Duclos et al. showed that it is possible to make sensible comparisons of poverty when accounts for multiple dimensions of well-being and that multidimensional comparisons can also differ from univariate comparisons in each individual dimension. More recently, Müller and Scarsini (2012) introduced inframodular utility functions that represent the attitude of an agent who dislikes transfers that move mass from inside a multidimensional interval to the sets above and below it, mimicking the mean preserving spread. Since a function is concave if and only if it has non-increasing differences, in the same way for multivariate functions, inframodularity (generalized concavity) implies that there are non-increasing differences. In this way, Müller and Scarsini use inframodular functions to model risk aversion that involves substitutable, but not complementary, commodities.

3 Main motivation

In the unidimensional case, when considering n individuals $1, \dots, j, \dots, n$ with incomes $x_1 \leq x_2 \leq \dots \leq x_j \leq \dots \leq x_n$, it is usually assumed that if $x_j < x_{j+1}$ then a transfer $\varepsilon > 0$ from individual $j + 1$, to individual j such that $x_j + \varepsilon \leq x_{j+1} - \varepsilon$ reduces inequality. This transfer is a Pigou-Dalton transfer. Note that this is equivalent to assuming that if $x_j \leq x_{j+1}$, then modifying x_j into $x_j - \varepsilon$ and x_{j+1} into $x_{j+1} + \varepsilon$ increases inequality.

Thus turning to the m -dimensional case $m \geq 1$, which is the topic of this paper, where each individual j has a column-vector $A_j \in \mathbb{R}^m$ of m attributes, we get that the $m \times n$ matrix $A = (A_1, \dots, A_j, \dots, A_n)$ summarizes the data. Imagine that for two individuals their respective column attributes are X and Y , with $X \leq Y$ (i.e. $x_i \leq y_i, \forall i = 1, \dots, m$) and let $\varepsilon \in \mathbb{R}_+^m$ with $\varepsilon \neq 0$. In such a situation, transfer ε from X to Y will be a regressive Pigou-Dalton transfer because it would increase inequalities.¹

Our goal is to axiomatize additive social evaluation functions, i.e., social evaluation functions $I : A \rightarrow \mathbb{R}$ such that $I(A) = \sum_{j=1}^n u(A_j)$ where $u : \mathbb{R}^m \rightarrow \mathbb{R}$ agrees with a diminishing social evaluation in the case of such a Pigou-Dalton regressive transfer as above. It is immediate that $\forall (X, Y) \in \mathbb{R}^m \times \mathbb{R}^m$ such that $X \leq Y$ and $\forall \varepsilon \geq 0$, u should satisfy:

$$u(X) - u(X - \varepsilon) \geq u(Y + \varepsilon) - u(Y) \quad (1)$$

This is the usual property of concavity in the *one-dimensional* case, at least when u is continuous. Actually, we show (see the *Annex*) that *inframodular functions*, extensively studied by Marinacci and Montrucchio (2005) and proposed by Müller and Scarsini (2001) as a meaningful representation of risk

¹Since this paper has been performed, we have been aware of a similar Pigou-Dalton principle introduced by Bosmans et al. (2009). Nevertheless main differences persist between the two papers: our definition is model-free and our main motivation is to link this principle to inframodularity.

aversion in the multidimensional case, satisfy this desired property (1), and are consistent with the relevant property of correlation increasing majorization. Inframodular functions may not be concave (e.g. Marinacci and Montruccio, 2005); therefore, as observed by Müller and Scarsini (2001), inframodular functions do not match the property of risk aversion, which states that adding a random vector E with mean 0 to a constant multivariate vector is always unfavorable. Note also that multidimensional concave functions may not be inframodular. A typical example is the three-dimensional HDI function $v : (x_1, x_2, x_3) \in [0, 1]^3 \rightarrow v(x_1, x_2, x_3) = x_1^{\frac{1}{3}} x_2^{\frac{1}{3}} x_3^{\frac{1}{3}}$. Therefore, an additive social evaluation function based on HDI might not respect the natural multidimensional Pigou-Dalton principle evocated above. As for an example let us consider two individuals with row attributes respectively $X = (0.8, 0.64, 0.729)$ and $Y = (0.9, 0.81, 0.729)$ and let us transfer $\varepsilon = (0, 0, 0.1)$ from X to Y . A simple computation delivers $v(X - \varepsilon) + v(Y + \varepsilon) > v(X) + v(Y)$.

Indeed, the updated HDI² denoted v above would not value the two previous situations as in the former computation, but the ranking would be again disputable since:

$$HDI(X, Y) = v\left(\frac{X + Y}{2}\right) = v\left(\frac{X - \varepsilon + Y + \varepsilon}{2}\right) = HDI(X - \varepsilon, Y + \varepsilon)$$

As a consequence this index would not take into account what appears as a clear deterioration of the social function with respect to inequality, when modifying (X, Y) into $(X - \varepsilon, Y + \varepsilon)$.

This contrasts with the additive social evaluation function that we propose in this paper (e.g. *Theorem 3* and *Section 8*), namely $I(X, Y) = u(X) + u(Y)$ where $u(x_1, x_2, x_3) = \ln \frac{x_1 + x_2 + x_3}{3}$ which delivers,

$$I(X, Y) = \ln(0.587) > I(X - \varepsilon, Y + \varepsilon) = \ln(0.583)$$

hence a ranking in accordance with the intuition.

As pointed out in the introduction, we intend to test the pertinence of inframodular social evaluations when compared to the famous HDI. This explains why we will focus on particularly tractable and meaningful inframodular functions u as proposed by Müller and Scarsini (2001; 2012) through:

$$u(x_1, \dots, x_i, \dots, x_m) = \psi\left(\sum_{i=1}^m \alpha_i x_i\right) \quad (2)$$

where ψ is concave and $\alpha_1, \dots, \alpha_m \geq 0$.

Note that such an u is a valuable function for our purpose, since, first, it makes sense to weight the different attributes in accordance with their importance $\alpha_i \geq 0$, and second, with such a ψ concave, the resulting u is *inframodular*, so it agrees with our definition of increasing inequality.

²In 2010 the HDI functional has changed its additive form to a multiplicative form as mentioned above. Section 8 includes a discussion about this ‘new’ HDI. Details are in Zambrano (2014).

Accordingly, in this paper, we mainly axiomatize the social evaluation function of the type given by (2), present the natural usual axioms aiming to specify ψ , and propose a simple relative inequality index. Moreover, as pointed out in the Introduction, our social evaluation functions is proved to satisfy the condition of correlation increasing majorization.

4 Framework and axioms

We consider n individuals $1, \dots, j, \dots, n$ and $A_j \in \mathbb{R}^m$ is the column-vector of the m attributes of this individual $A_j = \begin{pmatrix} a_{1j} \\ \vdots \\ a_{ij} \\ \vdots \\ a_{mj} \end{pmatrix}$; indeed, the same m

attributes are considered for each individual.³

Henceforth, for n given individuals, $A = (A_1, \dots, A_j, \dots, A_n)$ is the $m \times n$ matrix summarizing the considered population.

\mathcal{A} denotes the set of real matrices. If $a_{ij} \geq 0 \forall i$ and $\forall j$, we use the notation \mathcal{A}_+ , and if $a_{ij} > 0 \forall i$ and $\forall j$, we use the notation \mathcal{A}_{++} .

Thus \succsim is a preference relation on \mathcal{A} (if \mathcal{A}_+ or \mathcal{A}_{++} is used, this is specified in the *Theorems*). Here, we present the version of the axioms for \mathcal{A} . Indeed, for $A, B \in \mathcal{A}$, $A \succsim B$ means A is weakly preferred to B , etc.

Then \succsim is supposed to express the preferences of the policy-maker or the modeler for global welfare, taking into account that inequalities have a bad impact on welfare, but also that all attributes are ‘positive’; that is, any increase in some attribute has a positive effect on welfare.

The first three axioms are standard; therefore, they do not require a particular explanation:

A.1 \succsim is a weak order; i.e., \succsim is a transitive, complete hence reflexive binary relation on \mathcal{A} .

A.2 Continuity: Let $B \in \mathcal{A}$ be given, then $\{A \in \mathcal{A} | A \succsim B\}$ and $\{A \in \mathcal{A} | A \preceq B\}$ is closed in the usual topology of $\mathbb{R}^{n \times m}$.

A.3 Monotonicity: $\forall A, B \in \mathcal{A}$, $a_{ij} \geq b_{ij} \forall i, j$, implies $A \succsim B$; if furthermore $A \neq B$, then $A \succ B$.

For $A \in \mathcal{A}$ and A'_j a column of \mathbb{R}^m , (A'_j, A_{-j}) denotes the matrix A where column A_j has been replaced by column A'_j . Thus, the classical independence axiom that states that the impact for the ranking of replacing a given individual by another one is the same if all the other individuals remain unchanged.

A.4 Independence: $\forall j$ and $\forall A, B (A_j, A_{-j}) \succsim (A'_j, A_{-j}) \iff (A_j, B_{-j}) \succsim (A'_j, B_{-j})$.

Below is the classical anonymity index that states that the value of a distribution does not depend on the identity; only the value of the attributes matters.

³Note that throughout the paper we assume $n \geq 3$.

A.5 Anonymity: for any permutation matrix Π and for all $A \in \mathcal{A}$, one has $A \sim \Pi A$; i.e., $(A_1, \dots, A_j, \dots, A_n) \sim (A_{\sigma(1)}, \dots, A_{\sigma(j)}, \dots, A_{\sigma(n)})$ where $\sigma : [1, n] \rightarrow [1, n]$ is a bijection.

The last two axioms, to the best of our knowledge, are new and are crucial for our purpose.

Roughly speaking, the additivity axiom states that if an individual with a given vector of attributes is indifferent between two different lists of attributes for the others, he remains indifferent if he is endowed with a new vector of attributes.

A.6 Additivity: $\forall A, A_j, B_j, C_j, (A_j, A_{-j}) \sim (B_j, A_{-j}) \implies (A_j + C_j, A_{-j}) \sim (B_j + C_j, A_{-j})$.

The following Pigou-Dalton principle is the direct translation of the fact that if for an individual (j_1) all the attributes are smaller than for another one (j_2), then transferring for any i a value $\varepsilon_i \geq 0$ of attribute i from j_1 to j_2 clearly should increase the inequality (strictly increase if some $\varepsilon_i > 0$), thus leading to a worse situation.

A.7 Pigou-Dalton principle: Let $A = (A_1, \dots, A_j, \dots, A_n)$ such that for some j_1, j_2 one has $A_{j_1} \leq A_{j_2}$ and let $\varepsilon \in \mathbb{R}_+^m$ then:

$A_\varepsilon = (A_1, \dots, A_{j_1-1}, A_{j_1} - \varepsilon, A_{j_1+1}, \dots, A_{j_2-1}, A_{j_2} + \varepsilon, A_{j_2+1}, \dots, A_n) \precsim A$; furthermore, $A_\varepsilon \prec A$ if $\varepsilon \geq 0$ but $\varepsilon \neq 0$.

5 Multidimensional social evaluation functions

In this Section, multidimensional social evaluation functions are defined and characterized. *Theorem 1* offers axiomatization of the additive social evaluation function.

Theorem 1 *A preference relation \succsim on \mathcal{A} satisfies A.1, A.2, A.3, A.4, A.5 if and only if there exists: $u : \mathbb{R}^m \rightarrow \mathbb{R}$ increasing and continuous satisfying (1), such that:*

$$\forall A, B \in \mathcal{A}, A \succsim B \iff \sum_{j=1}^n u(A_j) \geq \sum_{j=1}^n u(B_j)$$

where u is defined up to a positive affine transformation (the proof is in the Appendix).

We come now to the main result of this paper in which we characterize social evaluation functions built upon the special type of inframodular functions satisfying (2) as proposed in a different framework by Müller and Scarsini. Such a social function agrees with our Pigou-Dalton principle A.7 and with the property of correlation increasing majorization as shown in Section 7.

Theorem 2 *A preference relation on \mathcal{A} satisfies A.1, A.2, A.3, A.4, A.5, A.6 and A.7 if and only if there exist $\alpha_i > 0$, $i = 1, \dots, m$, such that $\sum_{i=1}^m \alpha_i = 1$ and*

there exists $\psi : \mathbb{R} \rightarrow \mathbb{R}$ strictly increasing, strictly concave and continuous such that:

$$A \succsim B \iff \sum_{j=1}^n \psi\left(\sum_{i=1}^m \alpha_i \cdot a_{ij}\right) \geq \sum_{j=1}^n \psi\left(\sum_{i=1}^m \alpha_i \cdot b_{ij}\right).$$

Furthermore, such α'_i 's are unique and ψ is defined up to an increasing affine transformation.

Proof. The necessary part of the proof is straightforward since inframodular functions satisfy A.7 (see Lemma 1 in the Annex); thus we confine ourselves to proving the sufficiency part of the proof. From Theorem 1, we already know that the preference relations \succeq_j on \mathbb{R}^m of every individual j are identical.

Let us denote $\overset{*}{\succsim}$ this common preference relation and let us show that there exists $\alpha_i > 0, i = 1, \dots, m$, which $\sum_{i=1}^m \alpha_i = 1$ such that $\forall (X, Y) \in \mathbb{R}^m \times \mathbb{R}^m$,

$$X \overset{*}{\succsim} Y \iff \sum_{i=1}^m \alpha_i x_i \geq \sum_{i=1}^m \alpha_i y_i.$$

Moreover, A.1, A.2, A.3, A.4 imply that $\overset{*}{\succsim}$ satisfies:

A*.1: $\overset{*}{\succsim}$ is a weak order;

A*.2: continuity: $X^{(p)}, X, Y \in \mathbb{R}^m$ then:

A*.2.1: $X^{(p)} \overset{*}{\succsim} Y, \forall p, X^{(p)} \downarrow X \implies X \overset{*}{\succsim} Y$

A*.2.2: $X^{(p)} \overset{*}{\precsim} Y, \forall p, X^{(p)} \uparrow X \implies X \overset{*}{\precsim} Y$

A*.3: monotonicity: $X, Y \in \mathbb{R}^m, X \geq Y \implies X \overset{*}{\succsim} Y$, furthermore

if $X \neq Y \implies X \overset{*}{\succ} Y$

A*.4: additivity: $\forall X, Y, Z \in \mathbb{R}^m, X \overset{*}{\sim} Y \implies X + Z \overset{*}{\sim} Y + Z$

For $X \in \mathbb{R}^m$ denote $I(X) : \equiv \text{Inf} \left\{ x \mathbb{R} \mid x \cdot 1 \overset{*}{\succsim} X \right\}$, it is easy to see that

$I(X)$ exists in \mathbb{R} , that $X \overset{*}{\succsim} Y$ if and only if $I(X) \geq I(Y)$ and that $I(x \cdot 1) = x$,

$$\forall x \in \mathbb{R}, \text{ where indeed } 1 = \begin{pmatrix} 1 \\ \cdot \\ \cdot \\ \cdot \\ 1 \end{pmatrix} \in \mathbb{R}^m$$

Let e_i be the i_{th} vector of the canonical basis of \mathbb{R}^m , i.e., $e_i = \begin{pmatrix} 0 \\ \cdot \\ 1 \\ \cdot \\ 0 \end{pmatrix}$ the

i_{th} row, and let $\alpha_i := I(e_i)$, from A*.3, since $I(0) = 0$, one gets $\alpha_i > 0$.

We now intend to show that

$$(3) \quad I(X) = \sum_{i=1}^m \alpha_i x_i$$

Note that since $I(1) = 1$ this will entail $\sum_{i=1}^m \alpha_i = 1$.

In order to prove (3), let us show first that: $\forall Y, Z \in \mathbb{R}^m$ one has

$$(4) \quad I(Y + Z) = I(Y) + I(Z)$$

Since $Y \sim^* I(Y) \cdot 1$ and $Z \sim^* I(Z) \cdot 1$ A*.4 implies (4), since $Y + Z \sim^* I(Y) \cdot 1 + Z$ and $I(Y) \cdot 1 + Z \sim^* (I(Y) + I(Z)) \cdot 1$, gives $Y + Z \sim^* (I(Y) + I(Z)) \cdot 1$

It turns out that $I(X) = \sum_{i=1}^m I(x_i \cdot e_i)$. It remains to show that $I(x_i \cdot e_i) = x_i I(e_i)$.

It is enough to prove that.

$$(5) \quad I(x \cdot X) = x \cdot I(X) \quad \forall x \in \mathbb{R} \text{ and } \forall X \in \mathbb{R}^m.$$

This has been already proved for $x = 0$. So let us assume $x \in \mathbb{R}^*$.

Assume first $x \in \mathbb{Q}^*$ i.e. $x = \frac{p}{q}$, $p \in \mathbb{N}^*$, $q \in \mathbb{Z}^*$.

From (4) $I(\frac{p}{q} \cdot X) = I(p \cdot \frac{X}{q}) = p \cdot I(\frac{X}{q})$ but $I(X) = I(q \cdot \frac{X}{q}) = q \cdot I(\frac{X}{q})$

Therefore, $I(\frac{p}{q} \cdot X) = \frac{p}{q} \cdot I(X)$.

From A*.2 it is simple to see that $X_n \downarrow X \implies I(X_n) \downarrow I(X)$ and

$X_n \uparrow X \implies I(X_n) \uparrow I(X)$, and that I is monotone i.e. $X \geq Y \implies I(X) \geq I(Y)$.

So let us consider now $x \in \mathbb{R}$ and $x_n \in \mathbb{Q}$, $x_n \downarrow x$, $y_n \in \mathbb{Q}$, $y_n \uparrow x$, from A*.3 $x_n \cdot X \geq x \cdot X \geq y_n \cdot X$ implies:

$$x_n \cdot X \stackrel{*}{\succsim} x \cdot X \stackrel{*}{\succsim} y_n \cdot X \quad \text{so } I(x_n \cdot X) \geq I(x \cdot X) \geq I(y_n \cdot X)$$

Therefore, $x_n \cdot I(X) \geq I(x \cdot X) \geq y_n \cdot I(X) \quad \forall n$, so letting $n \longrightarrow +\infty$,

we get $I(x \cdot X) = x \cdot I(X)$, which completes the fact that:

$$X \stackrel{*}{\succsim} Y \text{ if and only if } \sum_{i=1}^m \alpha_i x_i \geq \sum_{i=1}^m \alpha_i y_i.$$

We end the proof by showing that up to an increasing affine transformation there exists a unique $\psi : \mathbb{R} \rightarrow \mathbb{R}$ strictly increasing, strictly concave, and continuous such that $\forall A, B \in \mathcal{A}$:

$$A \stackrel{*}{\succsim} B \iff \sum_{j=1}^n \psi\left(\sum_{i=1}^m \alpha_i \cdot a_{ij}\right) \geq \sum_{j=1}^n \psi\left(\sum_{i=1}^m \alpha_i \cdot b_{ij}\right).$$

From Theorem 1 there exists - up to a positive affine transformation - a unique u increasing and continuous satisfying (1) such that:

$$X, Y \in \mathbb{R}^m, X \stackrel{*}{\succsim} Y \iff u(X) \geq u(Y).$$

Since $I : \mathbb{R}^m \rightarrow \mathbb{R}$ is also a strictly increasing and continuous representation of $\stackrel{*}{\succsim}$, there exists up to a positive affine transformation a strictly increasing continuous function $\psi : \mathbb{R} \rightarrow \mathbb{R}$ such that $u = \psi \circ I$.

It remains to be proved that ψ is strictly concave.

It is enough to show that $\forall (a, b) \in \mathbb{R}^2$, $a \leq b$, $\forall \varepsilon > 0$ one has $\psi(a) - \psi(a - \varepsilon) > \psi(b + \varepsilon) - \psi(b)$.

It is immediate to find $X, Y \in \mathbb{R}^m$ such that $X \leq Y$ and

$\sum_{i=1}^m \alpha_i x_i = a$, $b = \sum_{i=1}^m \alpha_i y_i$, so from A.7, we get

$$u(X) - u(X - \varepsilon \mathbf{1}) > u(Y + \varepsilon \mathbf{1}) - u(Y)$$

Therefore, $\psi(a) - \psi(a - \varepsilon) > \psi(b + \varepsilon) - \psi(b)$, which completes the proof of *Theorem 2* ■

6 Specification of ψ and a relative inequality index

To specify ψ , we introduce two axioms that have a long tradition; see, for instance, Kolm (1976a;b) and, more recently, Gajdos and Weymark (2005).

A.8 $\forall A, B \in \mathcal{A}$ and $\forall \lambda \in \mathbb{R}$, $A \sim B \iff A + \lambda \mathbf{1} \sim B + \lambda \mathbf{1}$ where $\mathbf{1}$ is the matrix $m \times n$ with 1 everywhere.

This axiom quoted as a ‘leftist’ point of view by Kolm regarding inequalities expresses that the inequalities remain unchanged if the same amount is added to all attributes and to all individuals.

The following axiom quoted as a ‘centrist’ point of view by Kolm regarding inequalities, and that applies only if all the attributes are strictly positive, i.e., $A \in \mathcal{A}_{++}$, expresses that inequalities remain unchanged if all attributes are multiplied by the same positive number $\lambda > 0$ for all individuals.

A.9 $\forall A, B \in \mathcal{A}_{++}$ and $\forall \lambda > 0$, $A \sim B \iff \lambda A \sim \lambda B$

Theorem 3 Assume that the preference relation \succsim on \mathcal{A} satisfies A.1 to A.7, then:

- up to an increasing affine transformation $\psi(t) = -e^{-at}$ with $a > 0$ if and only if A.8 is satisfied;
- up to an increasing affine transformation either $\psi(t) = \ln(t)$, $\forall t > 0$ or $\psi(t) = t^a$, $\forall t > 0$ where $a \neq 0$, $a < 1$ if and only if A.9 is satisfied.

Proof. We prove only the *if part*; the *only if part* is straightforward.

Assume A.8 is satisfied. It is easy to see that if $\sum_{j=1}^n \psi(x_j) = \sum_{j=1}^n \psi(y_j)$ where $x_j, y_j \in \mathbb{R}$, we must have $\sum_{j=1}^n \psi(x_j + k) = \sum_{j=1}^n \psi(y_j + k) \quad \forall k \in \mathbb{R}$, and then we can apply the results of the classical one-dimensional social welfare theory (see e.g. Kolm, 1976a) to get the desired result.

Assume A.9 is satisfied. It is easy to see that $\sum_{j=1}^n \psi(x_j) = \sum_{j=1}^n \psi(y_j)$ where $x_j > 0, y_j > 0, \forall j$ implies $\sum_{j=1}^n \psi(\lambda x_j) = \sum_{j=1}^n \psi(\lambda y_j) \quad \forall \lambda > 0$, and then we can apply the results of the classical one-dimensional social welfare theory⁴ ■

⁴See e.g. Kolm (1976a;b) or Atkinson (1970) to get the result

Remark Note that in cases in which all attributes are strictly positive, and if we adopt axiom *A.9* then $\forall A \in \mathcal{A}_{++}$ one could adopt the social evaluation function $J(A) = \prod_{j=1}^n \left(\sum_{i=1}^m \alpha_i \cdot a_{ij} \right)$.

Indeed, in such a case: $I(A) = \sum_{j=1}^n \ln \left(\sum_{i=1}^m \alpha_i \cdot a_{ij} \right) = \ln \left(\prod_{j=1}^n \left(\sum_{i=1}^m \alpha_i \cdot a_{ij} \right) \right)$.

We focus on the relative inequality index that is linked with the choice of $\psi(\cdot) = \ln(\cdot)$. This index appears to be one of the most tractable and relevant in our framework.

6.1 Corollary of *Theorem 3*

The corresponding inequality index related to the social evaluation function defined on \mathcal{A}_{++} , the set of $m \times n$ matrices with positive elements, satisfying *A.1* to *A.7* and *A.9* with $\psi(t) = \ln(t)$, with $t > 0$ is relative and has the form

$$1 - \left(\frac{\prod_{j=1}^n \sum_{i=1}^m \alpha_i \cdot a_{ij}}{\sum_{i=1}^m \alpha_i \cdot \mu_i} \right)^{\frac{1}{n}}$$

Where μ_i , $i = 1, \dots, m$, is the mean of i^{th} attribute.

Proof. Following Tsui (1995) and Kolm (1977), let us define the multidimensional inequality index $I_R(A)$ for $A \in \mathcal{A}_{++}$ as $I_R(A) = 1 - \delta(A)$ where $\delta(A) \in [0, 1]$ is defined by $I(A) = I(\delta(A) \cdot A_\mu)$

where A_μ is the $m \times n$ matrix where each column writes $\begin{pmatrix} \mu_1 \\ \cdot \\ \mu_i \\ \cdot \\ \mu_m \end{pmatrix}$. From

$$I(A) = \sum_{j=1}^n \ln \left(\sum_{i=1}^m \alpha_i \cdot a_{ij} \right) = \ln \prod_{j=1}^n \left(\sum_{i=1}^m \alpha_i \cdot a_{ij} \right) \text{ and } I(\delta(A)) = \ln \left(\delta(A)^n \cdot \left(\sum_{i=1}^m \alpha_i \cdot \mu_i \right)^n \right)$$

one gets the desired result, namely, $I_R(A) = 1 - \left(\frac{\prod_{j=1}^n \sum_{i=1}^m \alpha_i \cdot a_{ij}}{\sum_{i=1}^m \alpha_i \cdot \mu_i} \right)^{\frac{1}{n}}$ ■

7 Agreeing with correlation increasing majorization.

Correlation increasing majorization (*CIM*) is a concept due to Boland and Proschan (1988) and introduced into the inequality literature by Tsui (1999). As pointed out by Tsui (1999), this type of majorization is known as an ordering of dependence in statistics (e.g. Shaked, 1982) and in economics of risks as ‘pairwise more risk’ (Richard, 1975). Note that *CIM* or the *majorization axiom* corresponds to Atkinson-Bourguignon ordering (Atkinson and Bourguignon, 1982), but Gajdos and Weymark (2005) observed that Bourguignon and Chakravarty (2003) raised reservations about this axiom, because *CIM* does not take into

account individual preferences. Since the point of view of our social evaluation is to consider a policy-maker or else a modeler who aims to consider each individual in the same way, we do not concur with the previous reservation and agree with the motivating examples given by Tsui in 1999.

Let A, B, C be the following three matrices summarizing the distributions of attributes,⁵

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 3 & 1 \\ 3 & 2 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 3 & 2 \\ 1 & 2 & 3 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{pmatrix}$$

Similar to Tsui (1999), we feel intuitively compelled to agree that the distribution of attributes summarized by C is most unequal followed by B and then by A .

Note that column B_1 and B_3 are nothing other than $B_1 = A_1 \wedge A_3$ and $B_3 = A_1 \vee A_3$, where \wedge and \vee are the classical operators min and max.

Furthermore, $C_2 = B_2 \wedge B_3$ and $C_3 = B_2 \vee B_3$.

It is time to define formally *correlation increasing majorization*.

First, let us introduce some definitions (see Boland and Proschan, 1988).

Concept of correlation increasing transfer (CIT)

Let $A, B \in \mathcal{A}$, then B is obtained from A by a *CIT* if there exists j_1, j_2 , $j_1 \neq j_2$ such that $B_{j_1} = A_{j_1} \wedge A_{j_2}$ and $B_{j_2} = A_{j_1} \vee A_{j_2}$

A *CIT* is strict whenever neither $A_{j_1} \leq A_{j_2}$ nor $A_{j_2} \leq A_{j_1}$.

Concept of correlation increasing majorization (CIM)

Let $A, B \in \mathcal{A}$, then $A >_c B$; i.e., A is strictly less unequal for the *CIM* if B may be derived from A by a permutation of columns and a finite sequence of the correlation increasing transfers at least one of which is strict.

We can now state and prove that our social evaluation functional of *Theorem 2* as well as any strict inframodular social functional of *Theorem 1* respect *CIM*.

We say that an inframodular function u is strict if:

$\forall (X, Y) \in \mathbb{R}^m \times \mathbb{R}^m$, $X < Y$, i.e., $X \leq Y$, $X \neq Y$ and $\varepsilon \geq 0$, $\varepsilon \in \mathbb{R}^m$, $\varepsilon \neq 0$,

one has $u(X + \varepsilon) - u(X) > u(Y + \varepsilon) - u(Y)$

Note this is the case for the inframodular function in *Theorem 2*.

Theorem 4 *Any strict inframodular social evaluation functional respects CIM.*

Proof. It is enough to prove that if A_1 and A_2 are two columns in \mathbb{R}^m , and neither $A_1 \leq A_2$ nor $A_2 \leq A_1$, then the inframodular function u satisfies $u(A_1) + u(A_2) > u(A_1 \wedge A_2) + u(A_1 \vee A_2)$.

The proof is in the Annex ■

Remark *CIM and multidimensional majorization*

⁵Each column represents an individual.

It is worth noticing that the type of reduction of inequality or in other words of majorization envisioned in this paper is different from the usual multidimensional majorization. Actually, this feature is particularly clear when considering *CIM* (as observed by Zoli, 2009). Consider distributions A and B such that

$$A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}, \text{ clearly } A >_c B,$$

but one cannot find any bistochastic matrix $\Pi = \begin{pmatrix} \lambda & 1-\lambda \\ 1-\lambda & \lambda \end{pmatrix}$ such that $\Pi B = A$.

The reader will find in Müller and Scarsini (2012) an extensive study of the sequence of inframodular transfers, which correspond to our type of majorization. This is performed in the case of general multidimensional probability measures. It would be interesting to derive the corresponding transfers in our simpler framework. This will be the subject of a future paper.

8 Empirical Analysis

Based on *Theorem 3*, we specialize $\psi(\cdot)$ as $\psi(\cdot) = \ln(\cdot)$, thus, considering cases in which all of the attributes are strictly positive we adopt $\forall A \in \mathcal{A}_{++}$ as a ‘mean’ social evaluation function,

$$J(A) = \frac{1}{n} \sum_{j=1}^n \ln \sum_{i=1}^m \alpha_i a_{ij}$$

We aim at evaluating the pertinence of this inframodular function using data. In order to test it, we decided to make a comparison with another function that is not inframodular, namely the famous Human Development Index (HDI). This index was launched by United Nations in 1990.⁶ The classical version of this index basically works with three variables (life expectancy (h), education (e) and income (w)) and provides a value, which allows to obtain ranking among countries. Recently, in 2010, this index has been updated⁷ and now its form is,

$$H(h, e, w) = h^{\frac{1}{3}} e^{\frac{1}{3}} w^{\frac{1}{3}}$$

where each variable is an index between 0 and 1 and consists, basically, in the population mean, for example, $h = \frac{1}{n} \sum_{j=1}^n h_j$.

Besides the famous role of HDI, we can simply consider this index as a way to aggregate different attributes, as well. As $H(h, e, w)$ provides an outcome between 0 and 1, we decided to extract the ‘certainty equivalent’ of $J(A)$, i.e., $I(A) = \exp^{J(A)}$. Then, now we have both indexes providing results in the range $[0, 1]$.

⁶For more details, see UNDP (1990).

⁷Zambrano (2014) has a discussion about this new index, its computation and axiomatization. See also Herrero et al. (2010).

The database chosen is the Brazilian national exam for high school students.⁸ The final notes in this exam are split up in five categories: natural sciences, human sciences, languages, mathematics and essay writing. In order to apply $H(\cdot)$ in these data, we focused our analysis in only three attributes, namely, natural sciences (a^s), languages (a^l) and mathematics (a^m).

The population (n) is the number of students in each town. Following HDI rules, here we also give the same weight to the attributes. The function $J(A)$ in this case, writes as,

$$J(A) = \frac{1}{n} \sum_{j=1}^n \ln \left(\frac{a_j^s + a_j^l + a_j^m}{3} \right)$$

It is widely known that classical HDI formula does not consider in its calculation the level of inequality within a country. For this, a specific index is available called IHDI.⁹ However, we are interested in contrast with the classical HDI to detect in which extent $I(A)$ is influenced by the ‘intra’ inequalities. In other words, we want to see whether this function delivers a worst result for towns which have more inequality among their students. In this case, as the classical HDI neglects inequality characteristics, this comparison could be a good option to test the effectiveness of this function with respect to inequality.

Once Brazil has 5570 towns, then we confined our analysis to Minas Gerais state. Below the descriptive statistics are presented.

Table 1 - Descriptive Statistics

Variable	Mean	Standard Deviation	Minimum	Maximum
Science	0.49611	0.07456	0.34200	0.87640
Language	0.51850	0.06725	0.30620	0.79440
Math	0.49559	0.11077	0.31850	0.97360
Number of Students	677,127			
Results by Town				
Function	Mean	Standard Deviation	Minimum	Maximum
$I(A)$	0.48412	0.01851	0.43577	0.53025
$H(A)$	0.48765	0.01913	0.43748	0.53450
Number of Towns	853			

Firstly, the difference between the $I(A)$ and $H(A)$ outcomes is relatively small. Their correlation coefficient is 0,999. The similarity of the outcomes is

⁸This exam is called ENEM (Exame Nacional do Ensino Médio – National high school exam). This exam is non-mandatory and has been used both as an admission test for enrollment in federal universities and educational institutes, as well as for certification for a high school degree.

⁹Kovacevic (2010) offers a good review and discussion about the importance of the inequality to evaluate the human development.

suitable, because it shows that this function provides the outcomes in a similar sense as HDI usually does. Nevertheless, we may see through the descriptive statistics table that there are some differences between both functions' results and we are interested in them.

For example, despite the strong closeness among the outcomes, we found that $H(A)$ is always bigger than $I(A)$ for every town, and this difference varies. Thus, since HDI does not consider inequality in its computation, we would like to know if the size of the difference between the functions is related with inequality level of the towns. In other words, we want to see whether inequality is positively correlated with $H(A) - I(A)$. To measure the inequality in this case, we summed the values of the attributes for each student and extracted the standard deviation of this transformed variable. We want to analyze the relation between these two variables in order to assert whether $I(A)$ takes inequality into account or not.

To answer this question, we need to evaluate these variables jointly. Below in *Figure 1*, one will find the dispersion graph of these two variables.

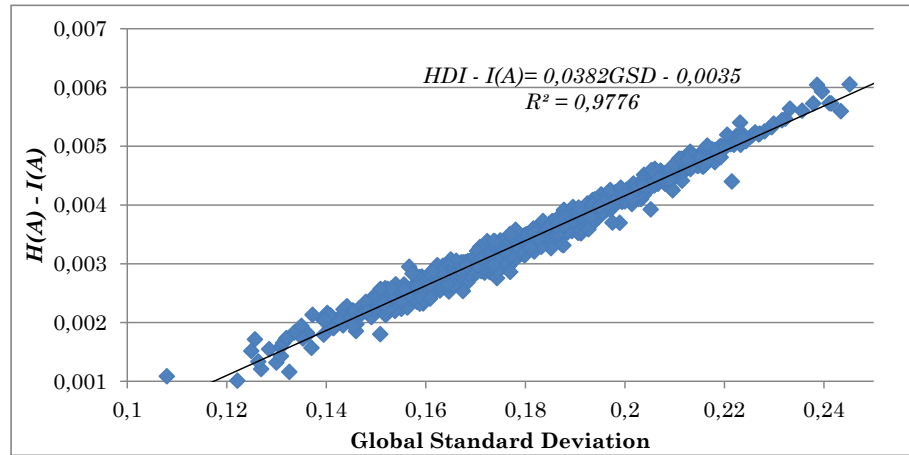


Figure 1: **Dispersion graph between $H(A) - I(A)$ and the global standard deviation of the attributes**

We, also, computed a linear regression as a hypothetical exercise. The equation is written in the graph and depicted by the black line. The value of R^2 attests that standard deviation explains almost 98% of the $H(A) - I(A)$ behavior. The positive relation between the variables is quite substantial.

Therefore, based on these results we suggest that this function could be a good alternative to IHDI. In short, we provide an inframodular function which can be used to aggregate several attributes (with different weights, if necessary), and takes into account the inequality inside the analyzed 'population'.

9 Concluding remarks

This paper aimed at characterizing a simple ‘additive’ social evaluation function based on a particular type of inframodular function proposed by Müller and Scarsini. In the multidimensional case it allows to respect what can be considered a natural Pigou-Dalton principle. Furthermore, if the policy-maker aims at treating every individual equally, which might be fair, our social evaluation functions agree with the property of correlation increasing majorization, already suggested by Tsui.

Building upon a long tradition, we specify our functions in order to obtain a simple tractable relative inequality index. Finally we propose an empirical analysis aiming at evaluating the pertinence of a specific inframodular evaluation function *à la* Müller and Scarsini, when compared to the famous HDI functional.

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10 Annex

Marinacci and Montrucchio (2005) provided a through analysis of ‘ultramodular functions’, thus (by reversing the inequality in the definition) of what Müller and Scarsini (2012) called ‘inframodular functions’.

Definition A function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is said to be *inframodular* if its increments are decreasing, namely:

$$(6) \quad f(x+h) - f(x) \geq f(y+h) - f(y) \\ \text{for all } x, y \in \mathbb{R}^n \text{ with } x \leq y \text{ and } h \in \mathbb{R}^n, h \geq 0$$

We intend now to prove that inframodular functions agree with our Pigou-Dalton regressive transfers (see Introduction).

Lemma 1 if $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is inframodular then f satisfies *Property 7* below:

$$(7) \quad f(x) - f(x-h) \geq f(y+h) - f(y) \\ \text{for all } x, y \in \mathbb{R}^n \text{ with } x \leq y \text{ and } h \in \mathbb{R}^n, h \geq 0$$

Proof. Let $x, y \in \mathbb{R}^n$ $x \leq y$ and $h \geq 0$. Set $x' = x - h$ and $y' = y$, $h' = h$ so $x' \leq y'$ and $h' \geq 0$, therefore

$$(6) \implies f(x' + h) - f(x') \geq f(y' + h) - f(y') \\ \text{i.e. } f(x) - f(x-h) \geq f(y+h) - f(y) \quad \blacksquare$$

Lemma 2 (Proof of *Theorem 4*)

First, it is known that if f is *inframodular*, then f is *submodular*, i.e., $\forall a, b \in \mathbb{R}^m \quad f(a) + f(b) \geq f(a \wedge b) + f(a \vee b)$ (see e.g. Marinacci and Montrucchio, 2005).

Let us show it again for sake of completeness.

Let $x = a \wedge b$, so $a = a \wedge b + h$ with $h \geq 0$.

Let $y = b$ one has $x \leq y$ and $h \geq 0$ so f inframodular implies

$$f(x+h) - f(x) \geq f(y+h) - f(y)$$

$$f(a) - f(a \wedge b) \geq f(b+a-a \wedge b) - f(b)$$

but $b+a-a \wedge b = a \vee b$; hence, the result:

$$f(a) + f(b) \geq f(a \wedge b) + f(a \vee b).$$

Thus one has $u(A_1) + u(A_2) \geq u(A_1 \vee A_2) + u(A_1 \wedge A_2)$

Since by hypothesis neither $A_1 \leq A_2$ nor $A_2 \leq A_1$, u strict inframodular implies $u(A_1) + u(A_2) > u(A_1 \vee A_2) + u(A_1 \wedge A_2)$

Actually since not $A_2 \leq A_1$, we get $A_1 \wedge A_2 < A_2$ so letting $x = A_1 \wedge A_2$, $y = A_2$, $\xi = A_1 - A_1 \wedge A_2$, we get:

$$u(x+\xi)+u(y) > u(y+\xi)+u(x) \text{ i.e. } u(A_1)+u(A_2) > u(A_1 \vee A_2)+u(A_1 \wedge A_2).$$

■

11 Appendix

Theorem 1

Proof. We discuss only the sufficiency part, since the necessary proof is immediate.

From A.1, A.2, A.3, A.4 (*weak order, continuity, monotonicity and independence*) and $n \geq 3$, Theorem 3 in Debreu (1960) implies that there exist n increasing and continuous functions $u_j : \mathbb{R}^m \rightarrow \mathbb{R}$ such that $A \succsim B \iff \sum_{j=1}^n u_j(A_j) \geq \sum_{j=1}^n u_j(B_j)$ where the u_j are unique up to affine transformation $\alpha u_j + \beta_j$ with $\alpha > 0$ and $\beta_j \in \mathbb{R}$. Thus, we can assume that $\forall j \ u_j(0) = 0$. From A.5 (*anonymity*), let us see that we can assume that there exists $u : \mathbb{R}^m \rightarrow \mathbb{R}$, increasing and continuous such that: $A \succsim B \iff \sum_{j=1}^n u(A_j) \geq \sum_{j=1}^n u(B_j)$. So, fix the u_j such that $u_j(0) = 0 \ \forall j$. By symmetry, we just need to prove that $u_1 = u_2$. Take any $A_1 \in \mathbb{R}^m$ and consider $(A_1, 0, A_3, \dots, A_n)$ and $(0, A_1, A_3, \dots, A_n)$. Through A.5: $u_1(A_1) + u_2(0) + \sum_{j=3}^n u_j(A_j) = u_1(0) + u_2(A_1) + \sum_{j=3}^n u_j(A_j)$, this entails straightforwardly $u_1(A_1) = u_2(A_1)$; thus, $u_1 = u_2 = \dots = u_n = u$. So there exists $u : \mathbb{R}^m \rightarrow \mathbb{R}$ increasing continuous (satisfying $u_j(0) = 0$) such that $A \succsim B \iff \sum_{j=1}^n u(A_j) \geq \sum_{j=1}^n u(B_j)$. Clearly, u is defined up to a positive affine transformation ■