Inequality measures and the issue of negative income (Misure di disuguaglianza: il problema dei redditi negativi)

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Abstract Income distribution studies have a long history in economic and statistical literature. Many results in such research area are provided by the standard inequality measure Gini coefficient, traditionally defined for non-negative income. In this paper the issue of negative income is faced and a specific reformulation of the Gini coefficient is introduced. More precisely, a new Gini coefficient normalization, held by the Pigou-Dalton transfers principle fulfillment, is presented.

Key words: Gini coefficient, negative income, Pigou-Dalton transfers principle, normalization term.

Abstract (in Italiano) Gli studi legati alla distribuzione dei redditi rappresentano un filone di ricerca molto ampio e contraddistinto da una lunga tradizione sia in ambito economico, sia in ambito statistico. In tale contesto, considerevoli risultati sono stati raggiunti con riferimento alla misura standard di disuguaglianza, ossia il coefficiente di Gini, originariamente definito per redditi non-negativi. Il contributo proposto in questo articolo si focalizza sull'estensione del coefficiente di Gini in presenza di redditi negativi, attraverso la definizione di un nuovo termine di normalizzazione basato sul principio dei trasferimenti di Pigou-Dalton.

1 Introduction

The measurement and assessment of income distribution represents an active research area which achieves a great interest in a wide set of scientific disciplines, such as economics and statistics. In such a context, the Gini coefficient appears as the common and most popular measure of inequality of income or wealth, as supported by its several developments and applications in literature. However, such applications are mainly restricted to the case of non-negative income, in order to fulfill the classical Gini coefficient formalization. For this reason, as stated by [4], removing the negative values from the analysis is basic as otherwise the need of resorting

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to a more complex methodology arises. Omission of negative income typically represents an usual procedure which finds a wide validation in many research papers (see e.g. [5, 7]). The main troubles associated to the treatment of negative income regard the violation of the normalization principle. In fact, the inclusion of incomes taking negative values implies that the standard Gini coefficient formula can achieve values greater than +1. To overcome this disadvantage, the Gini coefficient has to be adjusted in order to assure that its range is bounded between 0 and +1. [1] attempted to reformulate and normalize the Gini coefficient to make comparability between the distributions without negative income and the distributions with some negative income values be attained. The work presented in [1] was subsequently finished off by [2], who provided a correct expression for the Gini coefficient normalization term. In this paper the issue of negative income is further on stressed and a new reformulation of the Gini coefficient suitable for such purpose proposed. We believe that even if negative income can appear as an unfamiliar concept, it is worth noting the ways in which it can arise. Typically, in real surveys beside many positive income values one can observe also negative ones. It happens when assessing families financial assets such as, for instance, capital gains.

The paper is structured as follows: Section 2 is addressed to an overview of the existing contributions dealing with the extension of inequality measures when negative income appears and Section 3 focuses on a reformulation of the Gini coefficient shedding the light on a new Gini coefficient normalization proposal.

2 Background

A first attempt in providing an appropriate normalization term for the Gini coefficient when negative income is involved was given by [1]. Let N be the number of considered income units and Y_i the income of the i-th unit ordered in a non-decreasing sense. Let us denote with y_i the income share of the i-th unit, that is $y_i = \frac{Y_i}{N\mu_Y}$, with μ_Y corresponding to the average of Y. By resorting to the absolute mean difference-based formula, the normalized Gini coefficient G_{CTR}^{-1} can be expressed as

$$G_{CTR} = \frac{1 + \left(\frac{2}{N}\right) \sum_{i=1}^{k} i y_i - \left(\frac{1}{N}\right) \sum_{i=k+1}^{N} y_i (1 + 2(N - i))}{1 + \left(\frac{2}{N}\right) \sum_{i=1}^{k} i y_i},\tag{1}$$

with k defined in such a way: $\sum_{i=1}^k y_i = 0$. It is worth noting that the previous condition is rather uncommon since it results very unlike that in a sequence of incomes the first k values provide a null sum. Actually, [1] consider also the more general scenario when $\sum_{i=1}^k y_i < 0$ and $\sum_{i=k+1}^N y_i > 0$ even if a correct formulation of the corresponding Gini coefficient was finally accomplished by [2]. According to [2] the normalized Gini coefficient G_{BS}^2 in (1) for the more general case can be written

¹ CTR is the acronym of Chen, Tsaur and Rhai.

² BS is the acronym of Berebbi and Silber.

as

$$G_{BS} = \frac{\frac{2}{N} \sum_{i=1}^{N} i y_i - \sum_{i=1}^{N} y_i \frac{N+1}{N}}{1 + \left(\frac{2}{N}\right) \sum_{i=1}^{k} i y_i + \frac{1}{N} \sum_{i=1}^{k} y_i \left[\frac{\sum_{i=1}^{k} y_i}{y_{k+1}} - (1+2k)\right]},$$
 (2)

where y_{k+1} represents the first income share such that the $\sum_{i=1}^{k+1} y_i > 0$. Even if [1], and subsequently [2], provided the Gini coefficient normalization resorting to the absolute mean difference-based formula, at the same time they built the normalization term by taking into account the geometrical construction made by linking the Gini coefficient with the concentration area.

For a discussion about the properties of the Gini coefficient adjusted for negative income see [6].

3 Our contribution: extension and normalization of the Gini coefficient

The purpose of this section is threefold. First, in Subsection 3.1 we aim at extending the Gini coefficient computation when beside negative income also weights are taken into account. Furthermore, in Subsection 3.2 we illustrate some empirical examples shedding the light on some abnormal behaviors of the Gini normalization proposed by Berebbi and Silber. Finally, Subsection 3.3 deals with our proposal based on providing a new Gini coefficient normalization adjusted for the presence of negative values.

3.1 The Gini coefficient extension for weighted data

The classical Gini coefficient of an attribute *Y* with non-negative values can be translated into the below absolute mean difference-based expression:

$$G = \frac{1}{2\mu_Y N^2} \sum_{i=1}^{H} \sum_{j=1}^{H} |Y_i - Y_j| p_i p_j,$$
 (3)

where H is the total number of considered income units, p_i and p_j are weights associated to Y_i and Y_j such that $\sum_{i=1}^{H} p_i = N$ and μ_Y is the Y average value³. Typically, weights are introduced (and provided, for instance, by the Bank of Italy) to bring to the entire world or to use the equivalence scales (obtaining equivalent incomes) with the aim of making incomes own by income units with different size comparable. By extending (3) to the case of also negative income values, the Berebbi and Silber's Gini coefficient (hereafter denoted by G_{RS}^*) becomes

³ Note that weights p_i and N are non-integer.

$$G_{BS}^* = \frac{1}{2\mu_Y^* N^2} \sum_{i=1}^H \sum_{j=1}^H |Y_i - Y_j| p_i p_j, \tag{4}$$

where the normalization term μ_V^* results as

$$\mu_Y^* = \mu_Y + \frac{1}{2N^2} \sum_{i=1}^k \sum_{j=1}^k |Y_i - Y_j| p_i p_j + \frac{1}{N^2} \sum_{i=1}^k |Y_i - Y_{k+1}| p_i p_{k+1}^*$$
 (5)

with $p_{k+1}^* = p_{k+1} \frac{|\Sigma_{j-1}^k Y_j p_j|}{Y_{k+1} p_{k+1}}$. Expression in (5) is an extension of the Berebbi and Silber's normalization when also non-integer weights are taken into account.

Moreover, (4) can be developed as in (6). Indeed, by dividing (4) by μ_Y and denoting by $y_i = \frac{Y_i p_i}{\sum_{i=1}^{H} Y_i p_i}$ and $N_k = \sum_{i=1}^{k} p_i$, (4) can be expressed as

$$G_{BS}^{*} = \frac{\frac{2}{N} \sum_{i=1}^{H} \sum_{j=1}^{i} y_{i} p_{j} - \sum_{i=1}^{H} y_{i} \frac{N + p_{i}}{N}}{1 + \frac{2}{N} \sum_{i=1}^{k} \sum_{j=1}^{i} y_{i} p_{j} + \frac{1}{N} \sum_{i=1}^{k} y_{i} \left[\frac{p_{k+1}}{y_{k+1}} \sum_{i=1}^{k} y_{i} - 2N_{k} \right] - \frac{1}{N} \sum_{i=1}^{k} y_{i} p_{i}}, \quad (6)$$

which coincides with (2) in case $p_i = p_j = 1, \forall i, j = 1, ..., H$. For the sake of brevity, the proof of such equivalence is not reported here but it is available on request.

3.2 Some abnormal behaviors of G_{BS}^*

The Gini coefficient normalization firstly introduced by [1] and subsequently completed by [2] presents some abnormal behaviors in detecting the existing inequality between income distributions. Let us consider the three different income scenarios, regarding ten income units, reported in Table 1. For the sake of simplicity, let us suppose that $p_i = p_j = 1$, $\forall i = j$.

Scenarios	Scenarios Income vector Y									
Scenario (a):	-5	-5	-5	-5	-5	-5	-5	-5	-5	45.01
Scenario (b):	-45	0	0	0	0	0	0	0	0	45.01
Scenario (c):	-15	-10	-8	-7	-5	0	0	0	0	45.01

Table 1 Income distribution scenarios

Case (a) describes an income distribution characterized by almost all negative income values except for the last one which is positive, case (b) is representative of a distribution where only two income units own an income which takes on one hand a negative value and on the other hand a positive value and finally in case (c) some income units have negative income, some others null income and only one has a positive income. It would be then rational expecting that the Gini coefficient computed in the three scenarios varies in order to take into account income inequalities.

This does not occur, in fact as highlighted in Table 2, the normalized Gini coefficient G_{BS}^* is similar in all the considered situations and close to one, missing in measuring the existing income inequalities.

	Scenario (a)	Scenario (b)	Scenario (c)
G_{BS}^*	0.999999995	0.999999996	0.999999996

Table 2 G_{BS}^* results in scenarios (a), (b) and (c)

Such a result validates our proposal in reconsidering a new normalization for the Gini coefficient when negative income is involved. The main features of our proposed Gini coefficient normalization are discussed in the following subsection.

3.3 The reconsidered Gini coefficient normalization for negative income

The purpose of this subsection is introducing a new proposal for the Gini coefficient normalization when income distribution involves also negative values. More precisely, due to the drawbacks related to the contribution presented by [2], a more proper normalization is provided.

Let us take into account the specific scenario where, given H income units, the total positive (T^+) and negative $(-T^-)$ income are assigned only to two single units and all the others have no income, i.e. $Y = \{-T^-, 0, 0, 0, \dots, 0, 0, 0, T^+\}$. Such a context corresponds to scenario illustrated in (b). Furthermore, let $p_1 = p_H = 1$, while in all the other H - 2 cases p_i are non-integer and such that $\sum_{i=2}^{H-1} p_i = N - 2$. The income inequality maximization, here denoted by Δ_{max} , is then computed according to the absolute mean difference-based formula as follows:

$$\Delta_{max} = \frac{1}{N^2} \sum_{i=1}^{H} \sum_{i=1}^{H} |Y_i - Y_j| p_i p_j = 2 \frac{(N-1)(T^+ + T^-)}{N^2} = 2\mu_Y^{RSV}, \tag{7}$$

where T^- points out the absolute value of negative income and μ_Y^{RSV4} corresponds to $\frac{(N-1)(T^++T^-)}{N^2}$.

The same result can be reached by linking the Gini coefficient with the concentration area A obtained as

$$A = \frac{1}{2} - \frac{1}{2N} \frac{T^{+} - T^{-}}{T^{+}} + \frac{1}{2N(T^{+} - T^{-})} T^{-} + \frac{N - 2}{N(T^{+} - T^{-})} T^{-} + \frac{1}{2N(T^{+} - T^{-})} \frac{T^{-}}{T^{+}} T^{-}.$$
(8)

⁴ RSV is the acronym of Raffinetti, Siletti and Vernizzi.

One can prove that (7) and (8) coincide when (8) is multiplied by $(4/N)(T^+ - T^-)$. A comparison between the concentration area computed by [1] and [2] and that provided in (8) shows that the Gini coefficient normalization given by [2] does not take into account the term $-\frac{1}{2N}\frac{T^+ - T^-}{T^+}$ which for $N \to \infty$, becomes close to zero. However, this does not happen for a finite number N of observations.

It is worth noting, by fulfilling the following three conditions,

- 1. the income average value $\mu_Y = (T^+ T^-)/N$ has to be preserved in any income redistribution process;
- 2. T^+ is the maximum positive value that can not be exceeded in any income redistribution process;
- 3. T^- is the minimum negative value that can not be exceeded in any income redistribution process,

that any other income redistribution, according to the "Pigou-Dalton transfers principle" (see [3]) and based on income transfers among N>2 units, provides a mean difference smaller than $2[(N-1)/N^2](T^++T^-)$. Therefore, the Gini coefficient can be normalized by the term $2\mu_Y^{RSV}=2\frac{(N-1)(T^++T^-)}{N^2}$, or for $N\to\infty$, by $2\frac{(T^++T^-)}{N}$ 6. Alternatively one can resort also to the normalization term provided in (8), which however asymptotically holds⁷.

To validate the new introduced Gini coefficient normalization, let us reconsider the three different income distribution examples presented in Subsection 3.2. Results are displayed in Table 3.

	Scenario (a)	Scenario (b)	Scenario (c)
$\overline{G_{RSV}}$	0.555604932	1	0.834586280

Table 3 G_{RSV} results in scenarios (a), (b) and (c)

Findings in Table 3 show the attitude of the proposed normalized Gini coefficient $G_{RSV}^{\ 8}$ in detecting the real existing inequalities among the different scenarios. We remark that the Berebbi and Silber normalization in (2) holds for strictly positive income average, whereas our contributed normalization holds also in case of negative or null income average.

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⁵ The "Pigou-Dalton transfers principle" requires inequality measures decreasing as a consequence of any progressive transfer from a richer to a poorer person, preserving rank-order of incomes.

⁶ We remark that when $N \to \infty$, the simplifying assumption $p_1 = p_H = 1$ can be released.

⁷ If on one hand, as stated by [1], the normalized Gini coefficient in (4) reaches value +1 if $N \to \infty$, on the other hand our normalized Gini coefficient achieves value +1 also when N is very small, as shown in case (b) of Table 3.

⁸ $G_{RSV} = \frac{1}{2\mu_{RSV}N^2} \sum_{i=1}^{H} \sum_{j=1}^{H} |Y_i - Y_j| p_i p_j.$

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