

Subgroup Decomposition of the Gini Coefficient: A New Solution to an Old Problem*

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Abstract

We study inequality decomposition by population subgroups. We define properties of a satisfactory decomposition and ask what these properties imply for the decomposition of familiar inequality indices. We find that the Gini coefficient, the generalized entropy indices, and the Foster-Shneyerov indices all admit satisfactory decomposition formulas derived from a common set of axioms. While our axiomatic approach recovers the known decomposition formulas for the generalized entropy and the Foster-Shneyerov indices, it leads us to a novel decomposition formula for the Gini coefficient. The decomposition of the Gini coefficient can be easily computed, has both a geometric and an arithmetic intuition, and behaves better compared to existing decomposition formulas for the Gini coefficient.

I Introduction

Consider a population that is partitioned into subgroups. Each individual has an income and belongs to one of the subgroups. The task of inequality decomposition is to quantify the extent to which aggregate inequality reflects inequality within subgroups versus differences between subgroups.

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The Gini coefficient is the most commonly used measure of inequality¹, yet lacks a universally accepted decomposition formula. Existing approaches to decompose the Gini coefficient produce within-group or between-group inequality terms that behave counter-intuitively, and often rely on introducing a third term that describes neither within-group inequality nor between-group inequality. As a consequence, some researchers have advocated using other inequality measures with satisfactory decomposition formulas like the generalized entropy indices (Bourguignon, 1979; Shorrocks, 1980; Cowell, 1980; Cowell and Kuga, 1981; Shorrocks, 1984) or the Foster-Shneyerov (Foster and Shneyerov, 2000) indices in the context of decomposition analyses.

In this paper, we derive a novel decomposition of the Gini coefficient. The formula is stated in proposition 8. This decomposition for the Gini coefficient follows from a set of axioms that also implies the decomposition formulas for the generalized entropy and the Foster-Shneyerov indices. Moreover, the implied decomposition formulas are unique given our axioms. Within-group inequality in the Gini decomposition is proportional to a weighted power mean of subgroup Gini coefficients and between-group inequality can be computed as a residual. The decomposition has both a geometric and an arithmetic interpretation and behaves better compared to existing decomposition formulas for the Gini coefficient.

The decomposition of the Gini coefficient has additional desirable properties that the decompositions of other inequality indices lack. For example, reducing the distance between all incomes and the population average by a fixed percentage lowers the within-group and the between-group terms by the same percentage. Moreover, lump-sum transfers between subgroups do not affect the level of within-group inequality. Our results thus suggest that the Gini coefficient is an attractive choice of inequality measure, not only due to its intuitive arithmetic definition and geometric relation to the Lorenz curve, but also because it admits a satisfactory decomposition into within-group and between-group inequality.

An important result in this paper concerns the possible functional form that within-group inequality can take. We show that imposing intuitive restrictions on the behavior of the within-group inequality term implies that it must take the form of a quasi-arithmetic mean of subgroup inequalities with a Cobb-Douglas weight function in the subgroups' population and income shares. Within-group inequality in the decompositions of the generalized entropy or Foster-Shneyerov indices is a weighted arithmetic

¹The Gini coefficient is also one of the earliest measures of inequality. It was introduced by Corrado Gini in 1914 and constitutes an algebraic equivalent of the geometric measure of inequality devised by Max Otto Lorenz in 1905.

average of subgroup inequalities, whereas in the decomposition of the Gini coefficient it is a weighted power mean of subgroup Gini coefficients.

Between-group inequality summarizes how differences between subgroup income distributions contribute to aggregate inequality. For given subgroup inequality levels and population shares, generalized entropy indices depend on differences in average incomes between subgroups, while Foster-Shneyerov indices depend on differences in power-means. Between-group inequality reflects these different aggregativity properties so that for the generalized entropy indices it is a function of subgroup means while for the Foster-Shneyerov indices it instead depends on differences in power means. We show that for given subgroup Gini coefficients and population shares, the aggregate Gini coefficient generally depends on the entire distribution of income in each subgroup. Consequently, between-group inequality in the Gini decomposition generally depends on differences between subgroups in all moments of the income distribution.

We use the Gini decomposition to analyze the importance of demographic characteristics for household-level income inequality, and to study how differences in male versus female income distributions contribute to individual-level income inequality in the United States. For household-level income inequality, we find that the observed increase in the Gini coefficient over the past several decades is entirely driven by increased inequality within demographic subgroups. We also find that the convergence of male and female income distributions has significantly reduced individual-level income inequality. The decrease in inequality is mostly driven by the convergence in average income between men and women while differences in the shape of the income distributions continue to contribute to overall inequality.

Throughout this paper, we make use of the fact that the Lorenz curve of an aggregate population can be related to the Lorenz curves of the subgroups via the geometric operation of Minkowski addition. This geometric representation of the aggregation problem unlocks a powerful theorem from convex geometry, the Brunn-Minkowski theorem. We use these insights to discuss the extent of the subgroup inconsistency problem of the Gini coefficient². The Brunn-Minkowski theorem is also at the core of the arithmetic and geometric interpretation of the decomposition of the Gini coefficient.

The rest of the paper is structured as follows. In section II, we discuss the related literature. In section III, we introduce the notation and define the inequality indices discussed in this paper. In section IV, we introduce the axiomatic framework for subgroup

²The subgroup inconsistency problem has been discussed by Mookherjee and Shorrocks (1982), Cowell (1988), and Sen and Foster (1997).

decomposition of inequality indices. In section V, we derive the decompositions of the generalized entropy and the Foster-Shneyerov indices from our axioms and show that there does not exist a decomposition for Atkinson indices that satisfies the same list of axioms. In section VI, we discuss the aggregativity properties of the Gini coefficient, derive the unique decomposition of the Gini coefficient, and discuss its interpretation. We apply this decomposition to the US income distribution in section VII. Section VIII concludes. All proofs are presented in appendix A.

II Related Literature

In a series of seminal papers, Bourguignon (1979), Shorrocks (1980, 1984), Cowell (1980), and Cowell and Kuga (1981) propose a set of strict decomposability criteria and characterize the implied class of decomposable inequality indices. In particular, they require that a decomposable inequality index can be written as a function of subgroup inequality indices and aggregate characteristics³. Further, they require that between-group inequality depends only on aggregate characteristics of the subgroups and that within-group inequality is a weighted arithmetic average of subgroup inequality indices, where the weights only depend on aggregate characteristics but do not need to sum to one. Finally, they require that aggregate inequality is the sum of within-group inequality and between-group inequality. The main result from this literature is that the only continuous inequality indices that satisfy these decomposability criteria as well as the standard axioms for inequality indices⁴ are the generalized entropy indices⁵. These indices include the Theil index, the mean log deviation, and half the squared coefficient of variation.

Foster and Shneyerov (2000) start from a different definition of decomposability. They require that aggregate inequality is the sum of the inequality levels in two counterfactual income distributions that are defined for a given representative income function. Within-group inequality is the level of inequality when all subgroup distributions are re-scaled to the overall representative income level. Between-group inequality is the level of inequality when each individual's income is replaced with the subgroup's representative income. Since re-scaling income distributions to obtain within-group inequality and eliminating within-group inequality to obtain between-group inequality leads to

³By aggregate characteristics of the subgroups, we mean their total population and total income.

⁴The standard axioms for inequality indices are symmetry, population independence, scale independence, and the Pigou-Dalton principle.

⁵Foster and Shneyerov (1999) later show that continuity is not required for the result.

the same decomposition, Foster and Shneyerov (2000) call these measures *path independent*. Among the generalized entropy indices, which use the arithmetic average as representative income function, only the second Theil index has this path independence property⁶. To investigate the existence of other path-independent inequality indices, Foster and Shneyerov (2000) allow for more general representative income functions and find that for each power mean of order q , there is an associated path independent inequality index. Moreover, they show that this one parameter family of inequality indices, which includes the mean log deviation and the variance of logs⁷, is uniquely determined (up to a scalar multiple) by the path independence property.

The Gini coefficient is the most popular inequality index, yet it is not decomposable by either definition. Nevertheless, many suggestions have been made for how to decompose the aggregate Gini coefficient by population subgroups, including Soltow (1960), Bhattacharya and Mahalanobis (1967), Rao (1969), Mehran (1975), Piesch (1975), Mangahas (1975), Paglin (1975), Pyatt (1976), Murray (1979), Dagum (1980), Nygård and Sandström (1981), Kakwani (1984), Silber (1989), Yitzhaki and Lerman (1991), Lambert and Aronson (1993), Yitzhaki (1994), Dagum (1997), Aaberge (2000), Frosini (1989, 1990, 2003), Aaberge et al. (2005), Lambert and Decoster (2005), Deutsch and Silber (2007), Monti (2007), Costa (2008), Okamoto (2009), Shorrocks (2013), Mussard and Mornet (2019), and Attili (2021)⁸. Arguably the most influential decomposition of the Gini coefficient is due to Bhattacharya and Mahalanobis (1967) and mimics the decomposition of the generalized entropy indices. Within-group inequality is a weighted arithmetic average of subgroup Gini coefficients while between-group inequality depends only on the aggregate characteristics of the subgroups. This decomposition, however, introduces a residual term that is neither within-group nor between-group inequality and relates to the overlap of subgroup distributions⁹. It also suffers from counter-intuitive behavior, such as the fact that the level within-group inequality decreases in the number of subgroups even if all subgroup distributions are identical.

In this paper, we contribute both to the literature on inequality decomposition and to the literature on the Gini coefficient. We propose a novel definition of decomposability which admits the generalized entropy indices, Foster-Shneyerov indices, and the Gini coefficient as decomposable inequality indices. Unlike earlier definitions, our definition

⁶This has also been noted by Shorrocks (1980, 1988) and Anand (1983).

⁷Note that the variance of logs is not, strictly speaking, an inequality measure as it violates the Pigou-Dalton principle, see Foster and Ok (1999).

⁸See Giorgi (2011) for a history of different suggestions to decompose the Gini coefficient.

⁹Lambert and Aronson (1993) provide a geometric interpretation for this three-part decomposition.

of decomposability does not require the inequality index to satisfy any aggregativity properties. Instead, we call an inequality index decomposable if it can be written as a sum of a within-group inequality term and a between-group inequality term where each of these terms satisfies a given list of axioms. We show that our proposed list of axioms uniquely determines the decomposition formulas for the generalized entropy and the Foster-Shneyerov families, and also implies a novel and unique decomposition for the Gini coefficient.

An important result of this paper is that our axioms imply that within-group inequality must have the form of a weighted quasi-arithmetic mean with weights being a Cobb-Douglas function of subgroup population and income shares¹⁰ This result follows from similar axioms as those used to characterize means or aggregation functions¹¹. Our result can therefore be viewed more generally as a characterization of quasi-arithmetic aggregation functions when each aggregated value is accompanied by two independent measures of size or importance and when only weak reflexivity is imposed.

III Notation and Definitions

We use Y to denote the income distribution of the population and Y_i to denote the income distribution of subgroup i . Subgroup i 's population size and aggregate income are denoted by n_i and \bar{y}_i , respectively, and population and income shares are denoted by π_i and θ_i , respectively.

Throughout this paper, we consider continuous income distributions. Let \mathcal{D} denote the space of probability distributions supported on the positive real line. An inequality index $I: \mathcal{D} \rightarrow \mathbb{R}_+$ is a function that maps \mathcal{D} to the non-negative real numbers and satisfies the five standard axioms of anonymity, scale independence, population independence, normalization, and the Pigou-Dalton principle of transfers.

In this paper, we discuss four different classes of inequality indices. Generalized

¹⁰Note, however, that within-group inequality is not generally a mean as the weights do not always sum to one and thus the value of the function does not always lie between the minimum and maximum of the arguments.

¹¹For summaries of these literatures, see the introductions in Matkowski and Páles (2015), Burai et al. (2021), and Grabisch et al. (2011).

entropy indices are a one-parameter family of inequality indices defined as

$$GE_\alpha(Y) = \begin{cases} \frac{1}{\alpha(\alpha-1)} \int f(y) \left(\left(\frac{y}{\mu} \right)^\alpha - 1 \right) dy & \text{for } \alpha \neq 0, 1 \\ \int f(y) \ln \frac{y}{\mu} dy & \text{for } \alpha = 0 \\ \int f(y) \frac{y}{\mu} \ln \frac{y}{\mu} dy & \text{for } \alpha = 1 \end{cases}$$

where Y is the aggregate income distribution, f its probability density function, and μ is the aggregate mean income. Another well known family of decomposable inequality indices is the Foster-Shneyerov family¹² introduced by Foster and Shneyerov (2000) and defined as

$$FS_q(Y) = \begin{cases} \frac{1}{q} \ln \frac{\mu_q(Y)}{g(Y)} & \text{for } q \neq 0 \\ \frac{1}{2} \text{Var}(\ln Y) & \text{for } q = 0 \end{cases}$$

where $\mu_q(Y) = \left(\int y^q f(y) dy \right)^{\frac{1}{q}}$ is the power mean of order q of Y and $g(Y) = e^{\int \ln y f(y) dy}$ is the geometric mean of Y . The Gini coefficient is defined as

$$G(Y) = 2 \int_0^1 p - L(p) dp,$$

where L is the Lorenz curve defined as

$$L(p) = \frac{\int_0^p t f(t) dt}{\int_0^\infty t f(t) dt}.$$

Finally, the family of Atkinson indices (Atkinson, 1970) defined as

$$A_\varepsilon(Y) = 1 - \frac{\mu_{1-\varepsilon}(Y)}{\mu(Y)} \quad \text{for } \varepsilon < 1$$

where $\mu_{1-\varepsilon}(Y)$ is the power mean of order $1-\varepsilon$ of Y and $\mu_0(Y) := g(Y)$ is the geometric mean of Y .

¹²Note that only Foster-Shneyerov indices with parameter values $q \geq 1$ are strict inequality indices. For other parameter values, the Pigou-Dalton transfer principle is violated.

IV Axiomatic Framework

We define a subgroup decomposition of an inequality index I as the sum of a within-group inequality term W and a between-group inequality term B that add up to aggregate inequality. We propose a list of axioms that must be satisfied by any within-group or between-group inequality term. We call an inequality measure decomposable if it admits a decomposition that satisfies these axioms.

IV.1 Within-group Inequality

Let I denote an inequality index. We require that within-group inequality depends only on subgroup inequality levels and aggregate characteristics, that is, total population and total income. Let $W_k: \mathbb{R}_+^{3k} \rightarrow \mathbb{R}_+$ denote within-group inequality for a population consisting of k subgroups. Finally, let I_i, n_i, \bar{y}_i denote subgroup i 's level of inequality, total population, and total income. We posit the following axioms:

1. **Regularity.** W_k is continuous, and strictly increasing in each I_i if $n_i, \bar{y}_i > 0$.
2. **Symmetry.** $W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = W_k(P((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)))$ for any permutation P and for all $(I_i, n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_+^{3k}$.
3. **Scale and population independence.**

$$W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = W_k((I_1, an_1, b\bar{y}_1), \dots, (I_k, an_k, b\bar{y}_k))$$

for all $a, b > 0$ and for all $(I_i, n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_+^{3k}$.

4. **Normalization.** $W_k((0, n_1, \bar{y}_1), \dots, (0, n_k, \bar{y}_k)) = 0$ for all $(n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_+^{2k}$.
5. **Weak reflexivity.** $W_k((I, a_1 n, a_1 \bar{y}), \dots, (I, a_k n, a_k \bar{y})) = I$ for all $(I, n, \bar{y})_{i=1}^k \in \mathbb{R}_+^{3k}$ and for all (a_i) such that $\sum_{i=1}^k a_i = 1$.
6. **Replacement.**

$$\begin{aligned} & W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) \\ &= W_{k-m+1}\left(\left(\tilde{I}, \sum_{i=1}^m n_i, \sum_{i=1}^m \bar{y}_i\right), (I_{m+1}, n_{m+1}, \bar{y}_{m+1}), \dots, (I_k, n_k, \bar{y}_k)\right) \end{aligned}$$

for all $(I_i, n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_+^{3k}$ and for all $m \leq k$, where $\tilde{I} = W_m((I_1, n_1, \bar{y}_1), \dots, (I_m, n_m, \bar{y}_m))$.

Symmetry ensures that within-group inequality is independent of the labels given to each subgroup and is sometimes also called anonymity. Scale and population independence ensures that the decomposition is independent of the size of the population and the unit of measurement for income. Normalization ensures that within-group inequality is zero when there is no inequality within any subgroup. Weak reflexivity ensures that within-group inequality equals aggregate inequality when the population consists of only one group, and that within-group inequality does not change if that group is subdivided into subgroups with identical income distributions. Finally, replacement ensures that the level of within-group inequality does not change if we merge any number of subgroups into one composite subgroup and redistribute incomes in that group so that the level of inequality equals the level of within-group inequality among its constituent subgroups.

A first important result in this paper is that the six axioms stated above imply strong restrictions on the functional form that within-group inequality can have. This result is stated in theorem 1.

Theorem 1. *Let $(W_k)_{k=1}^\infty$ be a sequence of functions $W_k: \mathbb{R}_+^{3k} \rightarrow \mathbb{R}_+$. Then, (W_k) satisfies axioms 1-6 if and only if*

$$W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = f^{-1}\left(\sum_{i=1}^k \pi_i^{1-\alpha} \theta_i^\alpha f(I_i)\right), \quad (1)$$

for all $k \geq 1$ and for all $(I_i, n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_+^{3k}$, for some continuous strictly increasing function f such that $f(0) = 0$ and for some $\alpha \in \mathbb{R}$, where $\pi_i = \frac{n_i}{\sum_{i=1}^k n_i}$ and $\theta_i = \frac{\bar{y}_i}{\sum_{i=1}^k \bar{y}_i}$.

Theorem 1 holds that within-group inequality must take the form of a quasi-arithmetic mean of subgroup inequalities with weights that do not necessarily sum up to one. Moreover, the weight function must be a Cobb-Douglas function in the subgroups' population and income shares. The proof proceeds by showing that the axioms imply that for given aggregate characteristics W_k satisfies bisymmetry. Bisymmetry has been used by Aczél (1948) and Münnich et al. (2000) to characterize quasi-arithmetic means. The rest of the proof shows that the generating function f and the weight function do not depend on either the number of subgroups or their aggregate characteristics. The full proof can be found in appendix A.

The functional form in theorem 1 nests the expressions for within-group inequality in the standard decomposition formulas for the generalized entropy (Bourguignon, 1979; Shorrocks, 1980; Cowell and Kuga, 1981) and the Foster-Shneyerov indices (Foster

and Shneyerov, 2000), but is more general than the functional form assumed in these papers. Interestingly, however, the literature has also suggested expressions for within-group inequality that are ruled out by this theorem. For example, Shorrocks (2013) proposes an algorithm that results in a within-group inequality terms that generally does not have the form of a quasi-arithmetic mean. Moreover, the most common decomposition formula for the Gini coefficient due to Bhattacharya and Mahalanobis (1967) also violates theorem 1. In particular, it violates weak reflexivity as within-group inequality does not equal aggregate inequality even if all subgroups are identical.

IV.2 Between-group Inequality

Between-group inequality results from differences in income distributions across subgroups. What differences in subgroup income distributions give rise to between-group inequality depends on the choice of inequality index and its aggregativity properties¹³. Axioms for between-group inequality of a given inequality index are therefore stated in terms of its aggregativity properties. For this purpose, we define the concept of Ω -aggregativity.

Definition 1. *An inequality measure $I: \mathcal{D} \rightarrow \mathbb{R}_+$ is Ω -aggregative if there exists a function $F: \mathbb{R}_+^k \times \mathbb{R}_+^{mk} \times \mathbb{R}_+^k \rightarrow \mathbb{R}_+$ such that*

$$I(Y) = F(I(Y_1), \dots, I(Y_k), \Omega_1, \dots, \Omega_k; n_1, \dots, n_k),$$

where Y_i is the income distribution of subgroup i and Ω_i is the minimal finite set of moments of subgroup i 's income distribution such that the above formula holds, and m is the cardinality of Ω_i for all i . If there does not exist a finite Ω such that above formula holds, then I is Ω -nonaggregative.

Generalized entropy indices and Atkinson indices are inequality measures that can be written as functions of subgroup inequality levels, average incomes, and population sizes. Hence, these measures are means-aggregative with $\Omega_i = \{\mathbb{E}[Y_i]\}$. The Foster-Shneyerov index of order $q \neq 0$ is aggregative with q th moments, that is $\Omega_i = \{\mathbb{E}[Y_i^q]\}$.

¹³Suppose all subgroups have the same mean income and the same level of inequality. Weak reflexivity then implies that within-group inequality must equal the level of inequality within the subgroups. Depending on the inequality index, however, aggregate inequality can strictly exceed the level of within-group inequality. For example, Foster-Shneyerov indices for the aggregate population will generally be larger than the level of inequality in the subgroups. This excess inequality at the aggregate level is attributed to between-group inequality. On the other hand, aggregate generalized entropy indices do not produce between-group inequality in this case.

In proposition 7 in section VI, we show that the aggregate Gini coefficient can generally not be computed from subgroup Gini coefficients, population sizes, and any finite number of subgroup moments. For a given inequality index, it is useful to also introduce Ω -equivalence, which defines an equivalence relation between distributions.

Definition 2. *For a given inequality index, we call two distributions Ω -equivalent if they are identical in all moments contained in Ω . If Ω does not exist, then we call two distributions Ω -equivalent if they are the same distribution.*

Let I be a given inequality index. If I is Ω -aggregative, then we require that between-group inequality is a function of the subgroup moments contained in Ω and population sizes, that is, $B_k: \mathbb{R}_+^{mk} \times \mathbb{R}_+^k \rightarrow \mathbb{R}_+$, where k is the number of subgroups and m is the cardinality of Ω . If I is Ω -nonaggregative, then we allow between-group inequality to be a function of subgroup income distributions and population sizes, that is, $B_k: V^k \rightarrow \mathbb{R}_+$, where $V = \mathcal{D} \times \mathbb{R}_+$. We posit the following axioms:

7. **Zero between-group inequality.** If all subgroup income distributions are Ω -equivalent, then B_k is equal to zero.
8. **Conditional distribution independence.** For given aggregate characteristics $\{n_1, \dots, n_k, \bar{y}_1, \dots, \bar{y}_k\}$, if there exists a function $F: \mathbb{R}_+^k \rightarrow \mathbb{R}_+$ such that

$$I(Y) = F(I(Y_1), \dots, I(Y_k)),$$

then B_k does not depend on the distribution of income within subgroups.

Zero between-group inequality ensures that there is no between-group inequality when subgroup income distributions are identical in all moments that can affect aggregate inequality conditional on subgroup inequalities. Conditional distribution independence ensures that whenever aggregate inequality can be computed from subgroup inequality levels and aggregate characteristics, between-group inequality only depends on aggregate characteristics of the subgroups. Since axioms 7 and 8 depend on the aggregativity properties of a given inequality index, they do not imply a general functional form for between-group inequality. However, as we show below, axioms 1 through 8 are restrictive enough to uniquely determine the decomposition formulas of the generalized entropy indices, the Foster-Shneyerov indices, and the Gini coefficient.

V Decomposition of the Generalized Entropy, Foster-Shenyerov, and Atkinson Indices

The axiomatic framework introduced in the previous section leads to the standard decomposition formulas for the generalized entropy and the Foster-Shenyerov indices. Moreover, these are the only decomposition formulas for the indices that satisfy all our axioms. We also show that there does not exist a decomposition for Atkinson indices that satisfies all axioms.

Proposition 1. *For all $\alpha \in \mathbb{R}$, the decomposition of generalized entropy index*

$$GE_\alpha(Y) = GE_\alpha^W(Y) + GE_\alpha^B(Y) = \sum_{i=1}^k \pi_i^{1-\alpha} \theta_i^\alpha GE_\alpha(Y_i) + GE_\alpha(\bar{Y}), \quad (2)$$

where \bar{Y} is derived from Y by replacing each individuals income by the relevant subgroup mean, is the unique decomposition that satisfies axioms 1-8.

Proposition 2. *For all $q \in \mathbb{R}$, the decomposition of Foster-Shenyerov index*

$$FS_q(Y) = FS_q^W(Y) + FS_q^B(Y) = \sum_{i=1}^k \pi_i FS_q(Y_i) + FS_q(\tilde{Y}), \quad (3)$$

where \tilde{Y} is derived from Y by replacing each individuals income by the relevant subgroup power mean of order q , is the unique decomposition that satisfies axioms 1-8.

The proofs of propositions 1 and 2 are in appendix A.

We next turn to the Atkinson indices. These indices are means-aggregate and therefore ordinally equivalent to generalized entropy indices (Shorrocks, 1984). However, there is no universally accepted way to decompose Atkinson indices into within-group and between-group inequality that sum to aggregate inequality¹⁴. We show that there cannot exist an additive decomposition of Atkinson indices that satisfies all of our axioms.

Proposition 3. *For any $\varepsilon < 1$, the Atkinson index A_ε cannot be decomposed into a within-group inequality term and a between-group inequality term that satisfy axioms 1-8 and sum to the aggregate Atkinson index.*

¹⁴The most common decomposition of the Atkinson index was derived by Blackorby et al. (1981) and consists of a within-group inequality term W and a between-group inequality term B that combine into the aggregate Atkinson index as $A = W + B - WB$.

The proof is in appendix A.

VI Decomposition of the Gini Coefficient

In this section, we derive a novel decomposition for the Gini coefficient from the axiomatic framework introduced in section IV and show that it is unique. We also show that the decomposition for the Gini coefficient satisfies additional useful properties that are not imposed as axioms in our framework. Furthermore, we discuss how the decomposition of the Gini coefficient can be interpreted both arithmetically and geometrically. To understand the implication of axioms 7 and 8 for the Gini coefficient, we first discuss the aggregativity properties of the Gini coefficient and show that it is Ω -nonaggregative.

VI.1 Aggregativity Properties of the Gini Coefficient

It is well known that the Gini coefficient cannot be computed from the subgroup Gini coefficient, means, and population size (Bourguignon, 1979). With the notion of Ω -aggregativity, one may hope that there exists a finite set of higher-order moments that make the Gini Ω -aggregative. This turns out not to be the case.

We first show that the Lorenz curve of an aggregate population can be related to the Lorenz curves of the subgroups via the geometric operation of Minkowski addition. This geometric representation of the aggregation problem unlocks a powerful theorem from convex geometry, the Brunn-Minkowski theorem, which we use to show that the Gini coefficient is Ω -nonaggregative. The Brunn-Minkowski theorem is also at the core of the arithmetic and geometric interpretation of the decomposition of the Gini coefficient.

The Gini coefficient is traditionally defined as twice the area between the Lorenz curve and the line of perfect equality. For our purposes, it is useful to consider the region within the unit square that is bounded by the Lorenz curve and its centrally reflected counterpart. Let us call this region the *Lorenz region*. Clearly, the Gini coefficient is equal to the area of the Lorenz region.

For a given partition of the population into subgroups, we can also define subgroup Lorenz regions by scaling each subgroup's Lorenz region with a vector (π_i, θ_i) where π_i and θ_i are subgroup i 's population and income shares, respectively.

Definition 3. *The Lorenz region Λ_i of subgroup i with population share π_i and income share θ_i is defined as the region bounded by the scaled Lorenz curve and its centrally reflected counterpart in a rectangle with side lengths π_i and θ_i .*

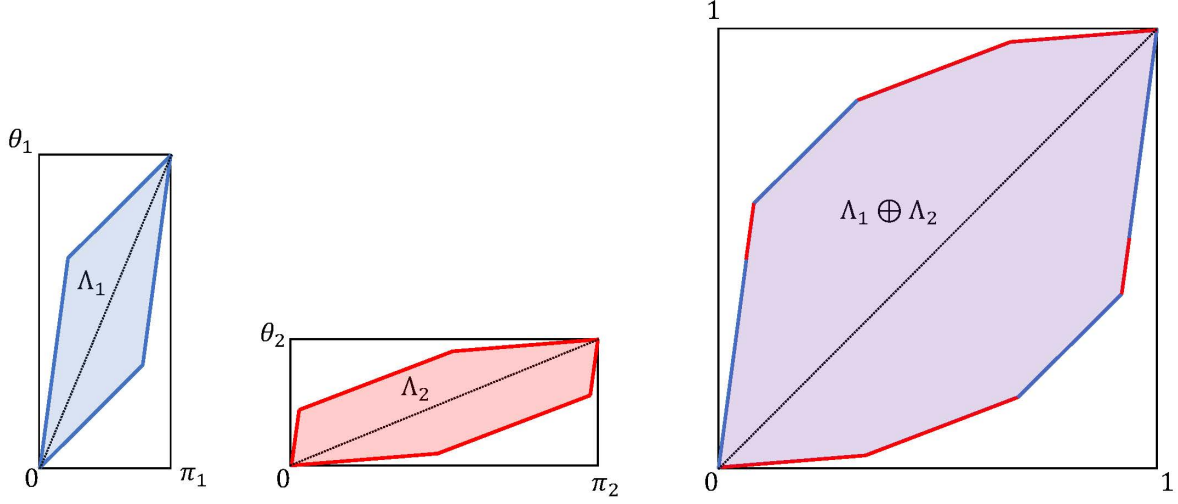


Figure 1: Aggregation of subgroup Lorenz regions. In this example, there are two different levels of income in subgroup 1 and three different levels of income in subgroup 2. In the aggregate Lorenz curve, individuals from both groups are ordered by their income. Geometrically, this corresponds to arranging the linear segments and the subgroup Lorenz curves in ascending order by their slopes. The resulting aggregate Lorenz region coincides with the Minkowski sum of the subgroup Lorenz regions.

Proposition 4 states that the Lorenz region of the aggregate population is the Minkowski sum of the subgroup Lorenz regions. The Minkowski addition of two subgroup Lorenz regions in the case of discrete income distributions is illustrated in figure 1.

Proposition 4. *A population consisting of $k \geq 2$ subgroups with Lorenz regions $\Lambda_1, \Lambda_2, \dots, \Lambda_k$ has an aggregate Lorenz region*

$$\Lambda = \Lambda_1 \oplus \Lambda_2 \oplus \dots \oplus \Lambda_k,$$

where \oplus denotes the Minkowski sum of sets.

The proposition is proven as part of theorem 1 in Zagier (1983).

This geometric representation of the aggregation problem provides helpful insight into a notorious behavior of the Gini coefficient that is sometimes called *subgroup inconsistency*: an increase in subgroup inequality, while keeping subgroup means and population sizes constant, can lead to a decrease in aggregate inequality. Different examples of specific distributions and transfers that produce this behavior are given in the literature (see, for example, Mookherjee and Shorrocks (1982) or Cowell (1988)). However, the conditions for the emergence of this behavior have not been stated ex-

plicitly.

Proposition 5 states that the Gini coefficient satisfies a weak subgroup consistency property. That is, any transfers within subgroups that increase subgroup inequalities according to the Lorenz criterion must increase the aggregate Gini coefficient. Equivalently, if transfers within subgroups reduce the aggregate Gini coefficient, it must be the case that there exists at least one subgroup whose new Lorenz curve intersects with the old Lorenz curve. The proof of proposition 5 makes use of the fact that outward shifts of the subgroup Lorenz curves must result in an outward shift of the population Lorenz curve.

Proposition 5 (Weak subgroup consistency of the Gini coefficient). *Implementing transfers within one or more subgroups that increase the level of subgroup inequality according to the Lorenz criterion must also increase the aggregate Gini coefficient.*

Another useful implication of representing the aggregate Gini coefficient as the area of the Minkowski sum of subgroup Lorenz regions is that we can make use of an important inequality relating the areas of compact sets: the Brunn-Minkowski theorem. Applied to the case of subgroup Lorenz regions, the Brunn-Minkowski theorem provides a lower bound for the aggregate Gini coefficient in terms of subgroup Gini coefficients and aggregate characteristics.

Proposition 6 (Brunn-Minkowski theorem). *For a population with income distribution Y and Lorenz object Λ consisting of k subgroups with income distributions Y_1, Y_2, \dots, Y_k , Lorenz regions $\Lambda_1, \Lambda_2, \dots, \Lambda_k$, population shares $\pi_1, \pi_2, \dots, \pi_k$, and income shares $\theta_1, \theta_2, \dots, \theta_k$, we have*

$$G(Y) = \text{Vol}(\Lambda) \geq \left(\sum_{i=1}^k \sqrt{\text{Vol}(\Lambda_i)} \right)^2 = \left(\sum_{i=1}^k \sqrt{\pi_i \theta_i G(Y_i)} \right)^2$$

with equality holding if and only if $\Lambda_1, \Lambda_2, \dots, \Lambda_k$ are homothetic.

The subgroup Lorenz regions are homothetic if and only if the distribution of income is identical across subgroups. We use this property to show that the Gini coefficient is Ω -nonaggregative.

Proposition 7. *The Gini coefficient is Ω -nonaggregative. That is, there does not exist a finite Ω , such that $G(Y_1, Y_2, \dots, Y_k; n_1, \dots, n_k) = F(G(Y_1), \dots, G(Y_k), \Omega_1, \dots, \Omega_k; n_1, \dots, n_k)$.*

VI.2 Subgroup Decomposition of the Gini Coefficient

We show that there exists a unique decomposition of the Gini coefficient into a within-group and a between-group term that satisfies all the axioms introduced in section IV. In this decomposition, within-group inequality is given by the Brunn-Minkowski lower bound for given areas of the subgroup Lorenz regions.

Proposition 8. *A decomposition for the Gini coefficient satisfies axioms 1-8 if and only if the within-group inequality term is*

$$G^W(Y) = \left(\sum_{i=1}^n \sqrt{\pi_i \theta_i} G(Y_i) \right)^2, \quad (4)$$

where π_i and θ_i are the population and income share of subgroup i , respectively. Between-group inequality is the difference between the aggregate Gini coefficient and within-group inequality.

Within-group inequality in (4) is a weighted power-mean of subgroup inequalities where each subgroup is weighted by the geometric mean of their income and population share. This decomposition of the Gini coefficient is the only decomposition that satisfies all our axioms on the behavior of within-group and between-group inequality.

The decomposition of the Gini coefficient has a number of additional desirable properties that are not imposed as axioms. First, within-group inequality is homogeneous¹⁵. Any redistribution of incomes within subgroups that reduces subgroup Gini coefficients by some common factor will also reduce within-group inequality by the same factor. Note, however, that as a power mean the within-group inequality term is not linear in the subgroup Gini coefficients.

The between-group inequality term for the Gini coefficient should summarize the contribution to aggregate inequality stemming from all differences across subgroup income distributions. It is therefore desirable that for any given distribution of income, between group inequality decreases if subgroup distributions become more similar. This is indeed the case. Randomly permuting the group affiliation of a sample of individuals results in subgroup distributions that are more similar to each other¹⁶. Proposition 9 states that this operation reduces between group inequality.

¹⁵Homogeneity is in fact the only additional property that power means have over quasi-arithmetic means implied by theorem 1.

¹⁶Equivalently, you may consider replacing the incomes of a fixed fraction of randomly sampled individuals in each group with a random draw from the aggregate income distribution. Equivalency between these two operations is a direct consequence of the anonymity property of inequality indices.

Proposition 9. *In an infinite population, randomly permuting the group affiliations for a random subset of individuals weakly reduces between-group inequality (while keeping the aggregate Gini coefficient constant). Between-group inequality remains constant if and only if all groups have the same distribution of income.*

We can also consider operations that should reduce both within-group inequality and between-group inequality. For example, redistributing incomes so that the difference between any given income and the average income is reduced by a fixed percentage clearly reduces aggregate inequality. Moreover, as such a redistribution at the same time compresses subgroup income distributions and brings them closer to each other, one may expect it to also decrease within-group and between-group inequality by similar proportions. Proposition 10 shows that this is indeed the case¹⁷. This property implies that a universal basic income financed by a flat tax which ignores group affiliations would not affect the share of aggregate inequality that is attributed to group differences.

Proposition 10. *Let $y(i)$ denote the income of quantile i , and let μ denote the average income in the population. For some $\alpha \in [0, 100]$, replacing every income $y(i)$ by $\tilde{y}(i) = y(i) - \frac{\alpha}{100}(y(i) - \mu)$ reduces within-group inequality and between-group inequality for the Gini coefficient by α percent.*

It is often of interest to know by how much inequality would be reduced if all subgroups had the same average level of income. For many inequality indices, such as the Theil index, this question is not easily answered. As was pointed out already by Shorrocks (1980), there is no obvious operation that would eliminate differences in average incomes between subgroups while keeping within-group inequality for any given generalized entropy index fixed¹⁸. Proposition 11 shows that the decomposition of the Gini coefficient admits an operation that eliminates differences between subgroups in average incomes while keeping within-group inequality fixed. This property can be used to further decompose between-group inequality into a first part that reflects differences in means and a second part that reflects differences in the shape of the distribution. We implement such an exercise in the final section of this paper.

Proposition 11. *Lump-sum transfers between subgroups can only affect between-group inequality, but not within-group inequality for the Gini coefficient.*

¹⁷Note that generalized entropy indices lack this property.

¹⁸The only exception is the mean log deviation, which belongs to both the generalized entropy and the Foster-Shneyerov families, and has the property that scaling the incomes in each subgroup by a common factor so that subgroup means are equalized leaves within-group inequalities fixed. In fact, all Foster-Shneyerov indices have this property by design.

Finally, we note that scaling or translating all incomes does not affect the share of aggregate inequality that is attributed to within-group or between-group inequality in the Gini decomposition. While the scale independence of the decomposition follows directly from the scale independence of the Gini coefficient and therefore applies to the decomposition of all scale-independent inequality indices, translation independence is a feature of only the Gini decomposition. This property is convenient, for example, if one is interested in decomposing inequality of income above the subsistence level, but lacks a good estimate of the level of subsistence expenses.

Proposition 12. *The Gini decomposition is both scale and translation invariant. Specifically, changing incomes at each quantile i from $y(i)$ to $\tilde{y}(i) = ay(i) + b$ does not affect the relative magnitude of within-group and between-group inequality for any $a > 0$ and $b \in \mathbb{R}$ such that all subgroups have positive average income.*

VI.3 Interpretation of the Gini Decomposition

Just like the Gini coefficient can be defined and interpreted both geometrically and arithmetically, our proposed decomposition of the Gini coefficient has both a geometric and an arithmetic interpretation. The within-group inequality term in (4) is the Brunn-Minkowski lower bound for the area of the Lorenz region given the areas of the subgroup Lorenz regions. Since the area of the Minkowski sum of two sets increases in both the sizes and the dissimilarity of the added sets, the area of the Minkowski sum that exceeds the lower bound can be interpreted as a measure of dissimilarity of the sets¹⁹. Thus, the between-group inequality term for the Gini coefficient gives the amount of inequality that is due to subgroup income distributions not being identical.

For the arithmetic interpretation of the Gini decomposition, we consider a discrete population. The Gini coefficient can then be defined as the sum of absolute income differences between all possible pairs of individuals in a population P , normalized by twice the product of the average income and the total number of such pairs:

$$G = \frac{\sum_{x,y \in P} |x - y|}{2\mu N^2}. \quad (5)$$

Decomposing the Gini coefficient amounts to decomposing the total sum of absolute differences in the numerator of expression (5). Differences between incomes of individ-

¹⁹In convex geometry, the square root of the excess area of the Minkowski sum over the Brunn-Minkowski lower bound is sometimes called the Brunn-Minkowski deficit, which has been shown to relate to the relative asymmetry of the added sets, see for example Figalli et al. (2009).

uals in the same subgroup should clearly contribute to within-group inequality²⁰. Note, however, that the overall sum of differences in the population always strictly exceeds the sum of differences within subgroups except when all incomes are identical and the sum of differences is zero. Hence, within-group inequality must be larger than the sum of absolute differences within subgroups. Proposition 13 shows that within-group inequality in the Gini decomposition can be expressed as the minimal sum of absolute differences in the population for given total absolute differences in each subgroup, normalized by twice the product of the average income and the total number of possible pairs in the population. The proof of the proposition re-states this complex arithmetic minimization problem in geometric form and relates it to the Brunn-Minkowski inequality.

Proposition 13. *Let S_m be a collection of real numbers for each $m \in \{1, \dots, M\}$. Then, the following inequality holds:*

$$\sum_{x,y \in \bigcup_{m=1}^M S_m} |x - y| \geq \left(\sum_{m=1}^M \sqrt{\sum_{x,y \in S_m} |x - y|} \right)^2. \quad (6)$$

Equality holds if and only if the distribution of numbers in each collection S_m is identical. Moreover, we have

$$G^W = \frac{1}{2\mu N^2} \left(\sum_{m=1}^M \sqrt{\sum_{x,y \in S_m} |x - y|} \right)^2,$$

where μ and N are the mean and the cardinality of the union of collections S_m .

VII Empirical Application

We use the decomposition for the Gini coefficient to answer two questions. To what extent does household income inequality in the US population reflect inequality within “similar” households versus differences between demographic subgroups? And how do differences in male versus female income distributions contribute to overall income inequality in the United States?

It is well known that income inequality in the United States has increased substantially since the early 1980s. Some of this increase may be driven by globalization or

²⁰The sum of absolute differences within subgroups, divided by $2\mu N^2$, constitutes in fact the within-group inequality in the influential decomposition of the Gini coefficient due to Bhattacharya and Mahalanobis (1967).

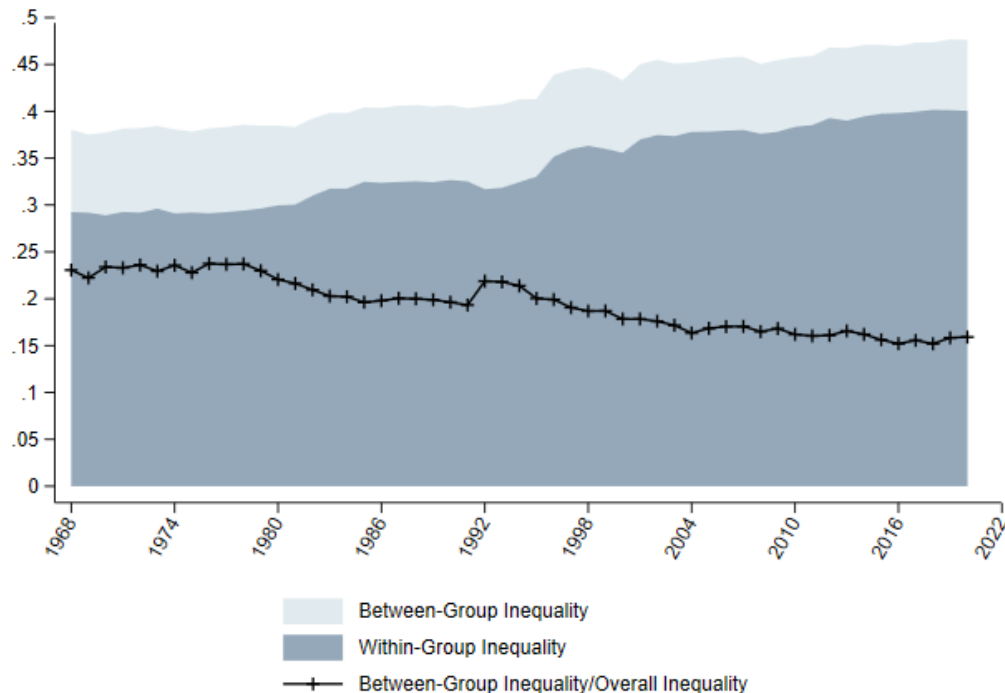


Figure 2: Household-level income inequality within and between demographic subgroups. Households are grouped by the age (≤ 35 , 36–45, 46–55, 56–65, > 65), education (no college, some college, Bachelor’s degree, more than Bachelor’s degree), sex (male, female), and race (Black, White, other) of the household head.

skill-biased technical change that affects the relative demand for high-skilled and low-skilled labor. Interestingly, however, it has been shown that inequality is also increasing within narrowly defined demographic groups.²¹ It is then natural to ask what share of the aggregate inequality can be accounted for by demographic characteristics and whether that has changed over time. The results from section VI allow us to study this question using the Gini coefficient as our measure of inequality²².

We use data on household-level income in the United States from the Current Population Surveys (CPS). To define demographic subgroups, we group households by the age, education, sex, and race of the household head. Figure 2 shows the evolution of within-group and between-group inequality over the time period 1968–2020. Over this time period, the aggregate Gini coefficient increases from 0.38 to 0.48. The rise in aggregate inequality is entirely driven by a rise in within-group inequality, which increases

²¹In an influential paper, Juhn et al. (1993) document rising wage inequality among men who are otherwise similar in terms of education and labor market experience. Since then, a large literature in labor economics devoted to explaining the rise in “residual inequality” has emerged.

²²Cowell and Jenkins (1995) study this question using different Atkinson indices.

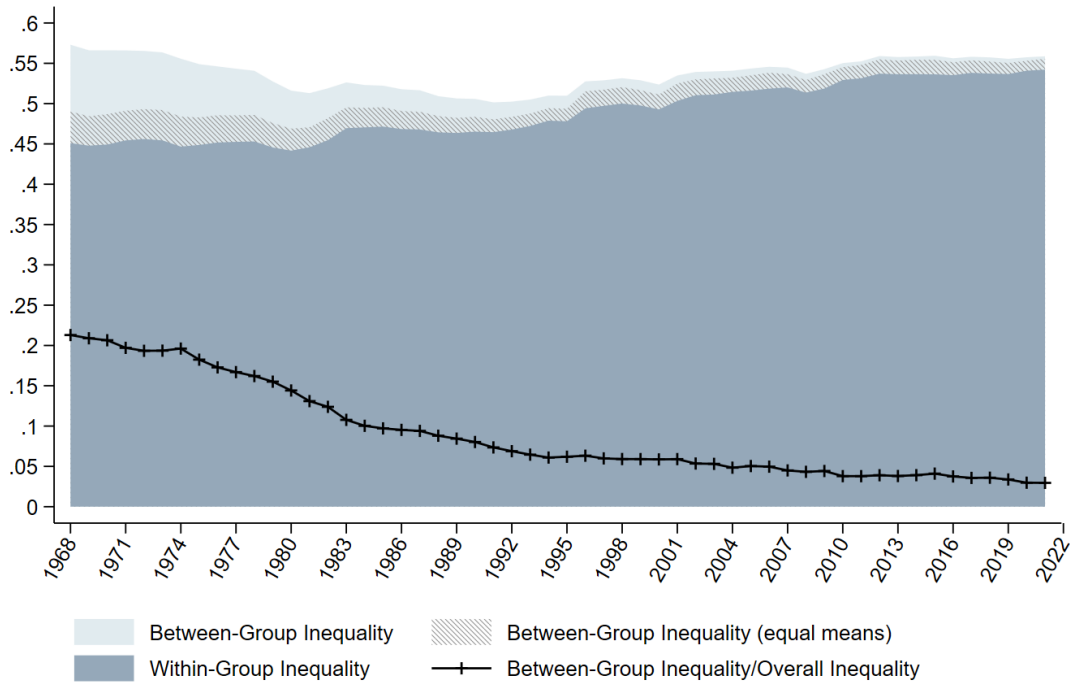


Figure 3: Individual-level income inequality within and between sexes. The shaded region shows between-group inequality when sex differences in average income are removed by implementing suitable lump-sum transfers.

from 0.29 to 0.40. Between-group inequality, on the other hand, decreases slightly from 0.09 to 0.08. As a consequence, the share of aggregate inequality that can be attributed to demographic characteristics decreases from 23 to 16 percent.

Among the most fundamental transformations of the distribution of income are the rise of female labor force participation and the convergence of male and female income distributions over time. Using the CPS data on individual-level incomes and grouping individuals by sex, figure 3 show the evolution of within-group and between-group inequality since 1968. We document a strong decline in the share of the aggregate Gini coefficient that can be attributed to differences between the income distributions of men and women (from 21% in 1968 to 3% in 2020). Making use of the fact that lump-sum transfers between subgroups do not affect within-group inequality, we can isolate the part of between-group inequality that remains after differences in average income between men and women are removed. While differences in average incomes contributed significantly to the level of aggregate inequality early in the sample period, we find that removing differences in average incomes via lump-sum transfers hardly

reduces the Gini coefficient in recent years²³. Nevertheless, differences between men and women in other moments of the income distribution still explain about 2.5% of overall inequality as measured by the Gini coefficient.

VIII Conclusion

In this paper, we suggest an axiomatic framework to derive subgroup decomposition formulas for different measure of inequality. We find that the Gini coefficient, the generalized entropy indices, and the Foster-Shneyerov indices all admit satisfactory decomposition formulas derived from a common set of axioms. The implied decomposition of the Gini coefficient is novel, easy to compute, and has both a geometric and an arithmetic interpretation.

²³This finding does not mean that the average incomes of men and women have almost converged. In fact, average income for women in 2019 remains at only 63% of that of men.

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Appendix

A Proofs

A.1 Proof of Theorem 1

It is easy to verify that (1) satisfies axioms 1-6. The remainder of the proof shows that if (W_k) satisfies axioms 1-6, then (1) holds.

We first show that for given $k \geq 2$ and aggregate characteristics $(n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_{++}^{2k}$, W_k must have a weaker functional form.

Lemma 2. *Suppose (W_k) satisfies axioms 1-6. Then, for given $k \geq 2$ and $(n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_{++}^{2k}$, there exist a continuous increasing function $f: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ such that $f(0) = 0$ and $a_i > 0$ for all $i = 1, \dots, k$, such that*

$$W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = f^{-1} \left(\sum_{i=1}^k a_i f(I_i) \right) \quad (7)$$

for all $i = 1, \dots, k$ and for all $(I_i)_{i=1}^k \in \mathbb{R}_+^k$.

Proof. Suppose $k \geq 2$ and $(n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_{++}^{2k}$ are given and define $B: \mathbb{R}_+^k \rightarrow \mathbb{R}_+$ as $B(x_1, \dots, x_k) = W_k((x_1, n_1, \bar{y}_1), \dots, (x_k, n_k, \bar{y}_k))$. Now, B satisfies the bisymmetry equation

$$B(B(x_{11}, \dots, x_{1k}), \dots, B(x_{k1}, \dots, x_{kk})) = B(B(x_{11}, \dots, x_{k1}), \dots, B(x_{1k}, \dots, x_{kk}))$$

for all $(x_{ij})_{i,j=1}^k \in \mathbb{R}_+^{k \cdot k}$. To show this, we use replacement, symmetry, and scale and population independence:

$$\begin{aligned} & B(B(x_{11}, \dots, x_{1k}), \dots, B(x_{k1}, \dots, x_{kk})) \\ &= W_k((W_k((x_{11}, n_1, \bar{y}_1), \dots, (x_{1k}, n_k, \bar{y}_k)), n_1, \bar{y}_1), \dots, (W_k((x_{k1}, n_1, \bar{y}_1), \dots, (x_{kk}, n_k, \bar{y}_k)), n_k, \bar{y}_k)) \\ &= W_k((W_k((x_{11}, \frac{n_1}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), \dots, (x_{1k}, \frac{n_1}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k)), n_1, \bar{y}_1), \\ & \quad \dots, (W_k((x_{k1}, \frac{n_k}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), \dots, (x_{kk}, \frac{n_k}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k)), n_k, \bar{y}_k)) \\ &= W_k((x_{11}, \frac{n_1}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), \dots, (x_{1k}, \frac{n_1}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k), \\ & \quad \dots, (x_{k1}, \frac{n_k}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), \dots, (x_{kk}, \frac{n_k}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k)) \\ &= W_k((x_{11}, \frac{n_1}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), \dots, (x_{k1}, \frac{n_k}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), \end{aligned}$$

$$\begin{aligned}
& \dots, (x_{1k}, \frac{n_1}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k), \dots, (x_{kk}, \frac{n_k}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k) \\
&= W_k(W_k((x_{11}, \frac{n_1}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), \dots, (x_{k1}, \frac{n_k}{\sum_{i=1}^k n_i} n_1, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_1), n_1, \bar{y}_1), \\
& \quad \dots, W_k((x_{1k}, \frac{n_1}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_1}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k), \dots, (x_{kk}, \frac{n_k}{\sum_{i=1}^k n_i} n_k, \frac{\bar{y}_k}{\sum_{i=1}^k \bar{y}_i} \bar{y}_k), n_k, \bar{y}_k) \\
&= W_k((W_k((x_{11}, n_1, \bar{y}_1), \dots, (x_{k1}, n_k, \bar{y}_k), n_1, \bar{y}_1), \\
& \quad \dots, w_k((x_{1k}, n_1, \bar{y}_1), \dots, (x_{kk}, n_k, \bar{y}_k), n_k, \bar{y}_k) \\
&= B(B(x_{11}, \dots, x_{k1}), \dots, B(x_{1k}, \dots, x_{kk})).
\end{aligned}$$

Thus, B is a continuous function that is strictly increasing in each of its arguments, symmetric, and satisfies bisymmetry. Following Aczél (1948), we show that we can use B to construct a function $M: \mathbb{R}_+^k \rightarrow \mathbb{R}_+$ that has the same aforementioned properties as B but is also reflexive. Define $F(z) := B(z, \dots, z)$ and $F^2(z) := B(B(z, \dots, z), \dots, B(z, \dots, z))$ for all $z \in \mathbb{R}_+$. Now, $F: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ and $F^2: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ are continuous and strictly increasing functions. Thus, $F^{-1}: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ exists and is also continuous and strictly increasing. Define $M(x_1, \dots, x_k) = F^{-1}(B(x_1, \dots, x_k))$. It follows that M is continuous and strictly monotonic in each x_i . Since

$$M(x, \dots, x) = F^{-1}(B(x, \dots, x)) = F^{-1}(F(x)) = x,$$

for any $x \in \mathbb{R}_+$, we see that M is reflexive. To show bisymmetry, let $(x_{ij})_{i,j=1}^k \in \mathbb{R}_+^{k \cdot k}$ and define

$$\begin{aligned}
z_1 &:= M(x_{11}, \dots, x_{1k}), \text{ i.e., } B(x_{11}, \dots, x_{1k}) = B(z_1, \dots, z_1) \\
\bar{z}_1 &:= M(x_{11}, \dots, x_{k1}), \text{ i.e., } B(x_{11}, \dots, x_{k1}) = B(\bar{z}_1, \dots, \bar{z}_1) \\
&\vdots \\
z_k &:= M(x_{k1}, \dots, x_{kk}), \text{ i.e., } B(x_{k1}, \dots, x_{kk}) = B(z_k, \dots, z_k) \\
\bar{z}_1 &:= M(x_{1k}, \dots, x_{kk}), \text{ i.e., } B(x_{1k}, \dots, x_{kk}) = B(\bar{z}_k, \dots, \bar{z}_k)
\end{aligned}$$

and

$$\begin{aligned}
z &:= M(z_1, \dots, z_k), \text{ i.e., } B(z_1, \dots, z_k) = B(z, \dots, z) \\
\bar{z} &:= M(\bar{z}_1, \dots, \bar{z}_k), \text{ i.e., } B(\bar{z}_1, \dots, \bar{z}_k) = B(\bar{z}, \dots, \bar{z}).
\end{aligned}$$

By the bisymmetry of B , we get

$$\begin{aligned}
F^2(z) &= B(B(z, \dots, z), \dots, B(z, \dots, z)) \\
&\quad B(B(z_1, \dots, z_k), \dots, B(z_1, \dots, z_k)) \\
&\quad B(B(z_1, \dots, z_1), \dots, B(z_k, \dots, z_k)) \\
&\quad B(B(x_{11}, \dots, x_{1k}), \dots, B(x_{k1}, \dots, x_{kk})) \\
&\quad B(B(x_{11}, \dots, x_{k1}), \dots, B(x_{1k}, \dots, x_{kk})) \\
&\quad B(B(\bar{z}_1, \dots, \bar{z}_1), \dots, B(\bar{z}_k, \dots, \bar{z}_k)) \\
&\quad B(B(\bar{z}_1, \dots, \bar{z}_k), \dots, B(\bar{z}_1, \dots, \bar{z}_k)) \\
&\quad B(B(\bar{z}, \dots, \bar{z}), \dots, B(\bar{z}, \dots, \bar{z})) = F^2(\bar{z}),
\end{aligned}$$

and hence, we get $z = \bar{z}$ which is equivalent to

$$M(M(x_{11}, \dots, x_{1k}), \dots, M(x_{k1}, \dots, x_{kk})) = M(M(x_{11}, \dots, x_{k1}), \dots, M(x_{1k}, \dots, x_{kk})).$$

Thus, M satisfies bisymmetry. We have shown that M satisfies the conditions of Theorem 2 in Münnich et al. (2000), which gives us

$$M(x_1, \dots, x_k) = f^{-1} \left(\sum_{i=1}^k b_i f(I_i) \right),$$

for some continuous increasing function $f: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ such that $f(0) = 0$ and for some $b_i > 0$ such that $\sum_{i=1}^k b_i = 1$. Thus, $B(x_1, \dots, x_k) = F(f^{-1}(\sum_{i=1}^k b_i f(I_i))) = \psi(\sum_{i=1}^k b_i f(I_i))$, where $\psi := F \circ f^{-1}$ is a strictly monotonic function. Using bisymmetry of B , we have

$$B(B(x_1, \dots, x_k), \dots, B(x_1, \dots, x_k)) = B(B(x_1, \dots, x_1), \dots, (x_k, \dots, x_k))$$

for all $(x_i)_{i=1}^k \in \mathbb{R}_+^k$, and thus,

$$\psi \left(f \circ \psi \left(\sum_{i=1}^k b_i f(x_i) \right) \right) = \psi \left(\sum_{i=1}^k b_i f \circ \psi(f(x_i)) \right)$$

for all $(x_i)_{i=1}^k \in \mathbb{R}_+^k$. Let $\phi := f \circ \psi$, and $z_i = f(x_i)$ for all i . Then, we have

$$\phi\left(\sum_{i=1}^k b_i z_i\right) = \sum_{i=1}^k b_i \phi(z_i)$$

for all $(z_i)_{i=1}^k \in \mathbb{R}_+^k$. This is Jensen's equation which has the unique solution $\phi(x) = ax + b$ for some $a \neq 0$ and $b \in \mathbb{R}$. Thus,

$$f(B(x_1, \dots, x_k)) = \phi\left(\sum_{i=1}^k b_i f(x_i)\right) = \sum_{i=1}^k a_i f(x_i) + b,$$

where $a_i = ab_i$, and hence

$$B(x_1, \dots, x_k) = f^{-1}\left(\sum_{i=1}^k a_i f(x_i) + b\right).$$

Now, since $f(0) = 0$, we have by the zero inequality axiom,

$$B(0, \dots, 0) = f^{-1}(b) = 0,$$

which implies $b = 0$. Thus, we get

$$W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = f^{-1}\left(\sum_{i=1}^k a_i f(I_i)\right)$$

for given $(n_i, \bar{y}_i) \in \mathbb{R}_{++}^{2k}$ and for all $(I_i)_{i=1}^k \in \mathbb{R}_+^k$. □

In lemma 2, the constants a_i and the generating function f can depend on k and $(n_i, \bar{y}_i)_{i=1}^k$. It follows from the symmetry of W_k that (a_i) and f are invariant under any permutation of (I_i, n_i, \bar{y}_i) .

Note that f and cf generate the same W_k for any constant $c > 0$. Thus, if $g(x) = cf(x)$ for all x , we call f and g the same generating function. We next show that the generating function is independent of k and $(n_i, \bar{y}_i)_{i=1}^k$. First, we show that the generating function does not change if we add an arbitrary subgroup to the existing population.

Lemma 3. *Suppose $(W_k)_{k=2}^\infty$ satisfies axioms 1-6. Let $k \geq 2$. Then, $W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k))$ and $W_{k+1}((I_1, n_1, \bar{y}_1), \dots, (I_{k+1}, n_{k+1}, \bar{y}_{k+1}))$ have the same generating function for all $(I_i, n_i, \bar{y}_i)_{i=1}^{k+1} \in \mathbb{R}_+^{3(k+1)}$.*

Proof. By replacement, we have

$$\begin{aligned} & W_{k+1}((I_1, n_1, \bar{y}_1), \dots, (I_{k+1}, n_{k+1}, \bar{y}_{k+1})) \\ &= W_2\left(\left(W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)), \sum_{i=1}^k n_i, \sum_{i=1}^k \bar{y}_i\right), (I_{k+1}, n_{k+1}, \bar{y}_{k+1})\right) \end{aligned}$$

for all $(I_i, n_i, \bar{y}_i)_{i=1}^{k+1} \in \mathbb{R}_+^{3k}$. Suppose $(n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_{++}^{2(k+1)}$ are given. Using lemma 2, we get

$$f^{-1}\left(\sum_{i=1}^{k+1} a_i f(I_i)\right) = g^{-1}\left(b_1 g\left(h^{-1}\left(\sum_{i=1}^k c_i h(I_i)\right)\right) + b_2 g(I_{k+1})\right)$$

or

$$\sum_{i=1}^{k+1} a_i f(I_i) = f \circ g^{-1}\left(b_1 g \circ h^{-1}\left(\sum_{i=1}^k c_i h(I_i)\right) + b_2 g(I_{k+1})\right)$$

for all $(I_i, n_i, \bar{y}_i)_{i=1}^{k+1} \in \mathbb{R}_+^{3(k+1)}$, where f, g, h are continuous and strictly increasing functions and a_i, b_i, c_i are some constants for all $i = 1, \dots, k+1$. Set $x_i = f(I_i)$ for all $i = 1, \dots, k+1$. Then we have

$$\sum_{i=1}^{k+1} a_i x_i = f \circ g^{-1}\left(b_1 g \circ h^{-1}\left(\sum_{i=1}^k c_i h \circ f^{-1}(x_i)\right) + b_2 g \circ f^{-1}(x_{k+1})\right). \quad (8)$$

Since $f \circ g^{-1}$ and $g \circ f^{-1}$ are differentiable almost everywhere, we get

$$(f \circ g^{-1})'\left(b_1 g \circ h^{-1}\left(\sum_{i=1}^k c_i h \circ f^{-1}(x_i)\right) + b_2 g \circ f^{-1}(x_{k+1})\right) b_2 (g \circ f^{-1})'(x_{k+1}) = a_{k+1}$$

for almost every $(x_i)_{i=1}^k \in \mathbb{R}_+^k$. Now, either $(f \circ g^{-1})'$ and $(g \circ f^{-1})'$ are constants almost everywhere, or $b_1 g \circ h^{-1}\left(\sum_{i=1}^k c_i h \circ f^{-1}(x_i)\right) + b_2 g \circ f^{-1}(x_{k+1}) = \phi(x_{k+1})$ for some function ϕ . Because

$$b_1 g \circ h^{-1}\left(\sum_{i=1}^k c_i h \circ f^{-1}(x_i)\right) + b_2 g \circ f^{-1}(x_{k+1}) = b_1 g(W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k))),$$

where the right-hand side is strictly increasing in each I_i and thus the left-hand side is strictly increasing in each x_i , the latter cannot be true. Thus, we get

$$(f \circ g^{-1})'(x) = c \text{ a.e.}$$

$$\begin{aligned}
&\Longleftrightarrow f \circ g^{-1}(x) = \int_0^x c dt = cx \quad a.e. \\
&\Longleftrightarrow g^{-1}(x) = f^{-1}(cx) \quad a.e. \\
&\Longleftrightarrow f(x) = cg(x) \quad a.e.
\end{aligned}$$

for some constant $c > 0$. Since f and g are continuous functions, we get $f(x) = cg(x)$ for all $x \in \mathbb{R}_+$. Plugging this result into (8) and differentiating with respect to x_i for some $i = 2, \dots, k$, we get

$$b_1(f \circ h^{-1})' \left(\sum_{i=1}^k c_i h \circ f^{-1}(x_i) \right) c_i (h \circ f^{-1})'(x_i) = a_i$$

almost everywhere. Again, we get that either $(f \circ h^{-1})'$ and $(h \circ f^{-1})'$ are constants almost everywhere, or $b_1(f \circ h^{-1})' \left(\sum_{i=1}^k c_i h \circ f^{-1}(x_i) \right) = \phi(x_i)$ for some function ϕ . Since

$$b_1(f \circ h^{-1})' \left(\sum_{i=1}^k c_i h \circ f^{-1}(x_i) \right) = b_1 f \left(W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) \right),$$

where the right-hand side is strictly increasing in each I_i and thus the left-hand side is strictly increasing in each x_i , the latter cannot be true. Hence, we get that $(f \circ h^{-1})'$ is constant almost everywhere which implies $f(x) = ch(x)$ for all $x \in \mathbb{R}_+$. Therefore, $W_n((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k))$ and $W_{k+1}((I_1, n_1, \bar{y}_1), \dots, (I_{k+1}, n_{k+1}, \bar{y}_{k+1}))$ have the same generating function for all $k \geq 2$ and for all $(I_i, n_i, \bar{y}_i)_{i=1}^{k+1}$. \square

Now, it follows that the generating function is independent of k and $(n_i, \bar{y}_i)_{i=1}^k$. First, for any $I_1, n_1, \bar{y}_1, I_2, n_2, \bar{y}_2$ and $I'_1, n'_1, \bar{y}'_1, I'_2, n'_2, \bar{y}'_2$, $W_2((I_1, n_1, \bar{y}_1), (I_2, n_2, \bar{y}_2))$ and $W_2((I'_1, n'_1, \bar{y}'_1), (I'_2, n'_2, \bar{y}'_2))$ have the same generating function. To see why this is true, note that $W_2((I_1, n_1, \bar{y}_1), (I_2, n_2, \bar{y}_2))$ and $W_4((I_1, n_1, \bar{y}_1), (I_2, n_2, \bar{y}_2), (I'_1, n'_1, \bar{y}'_1), (I'_2, n'_2, \bar{y}'_2))$ have the same generating function by lemma 3, and $W_2((I'_1, n'_1, \bar{y}'_1), (I'_2, n'_2, \bar{y}'_2))$ and $W_4((I_1, n_1, \bar{y}_1), (I_2, n_2, \bar{y}_2), (I'_1, n'_1, \bar{y}'_1), (I'_2, n'_2, \bar{y}'_2))$ have the same generating function again by lemma 3 and the symmetry of W_4 . Finally, W_k has the same generating function for all $k \geq 2$ and for all $(n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_{++}^{2k}$ since W_2 has the same generating function for all $(n_i, \bar{y}_i)_{i=1}^k$, and we can construct W_k from W_2 without changing the generating function by using lemma 3 iteratively.

Finally, we show that the constants a_i in

$$W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = f^{-1}\left(\sum_{i=1}^k a_i f(I_i)\right)$$

always have the form $a_i = \pi_i^{1-\alpha} \theta_i^\alpha$.

Lemma 4. *In equation (7), $a_i = \pi_i^{1-\alpha} \theta_i^\alpha$ for all $(I_i, n_i, \bar{y}_i)_{i=1}^{k+1} \in \mathbb{R}_+^{3k}$.*

Proof. Using replication and lemmas 2 and 3, we get

$$W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = f^{-1}\left(\sum_{i=1}^k a_i f(I_i)\right) = f^{-1}\left(b_1 \sum_{i=1}^{k-1} c_i f(I_i) + b_2 f(I_k)\right) \quad (9)$$

for all $k \geq 2$ and $(I_i, n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_+^{3k}$ where $a_i = a(n_i, \bar{y}_i, \dots, n_k, \bar{y}_k)$, $c_i = c(n_i, \bar{y}_i, \dots, n_{k-1}, \bar{y}_{k-1})$, $b_1 = (\sum_{i=1}^{k-1} n_i, \sum_{i=1}^{k-1} \bar{y}_i, n_k, \bar{y}_k)$, and $b_2 = b(n_1, \bar{y}_1, \sum_{i=1}^{k-1} n_i, \sum_{i=1}^{k-1} \bar{y}_i)$ for some functions a , b , and c . By setting $I_i = 0$ for $i = 1, \dots, k-1$, we get

$$a(n_k, \bar{y}_k, \dots) = b\left(n_1, \bar{y}_1, \sum_{i=1}^{k-1} n_i, \sum_{i=1}^{k-1} \bar{y}_i\right)$$

for all k and $(n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_+^{2k}$. By scale and population independence, we get

$$a(n_k, \bar{y}_k, \dots) = b(\pi_k, \theta_k, 1 - \pi_k, 1 - \theta_k) = w(\pi_k, \theta_k)$$

for some function w . Plugging this result back into equation (9), we get

$$w\left(\frac{n_i}{\sum_{i=1}^k n_i}, \frac{\bar{y}_i}{\sum_{i=1}^k \bar{y}_i}\right) = w\left(\frac{\sum_{i=1}^{k-1} n_i}{\sum_{i=1}^k n_i}, \frac{\sum_{i=1}^{k-1} \bar{y}_i}{\sum_{i=1}^k \bar{y}_i}\right) w\left(\frac{n_i}{\sum_{i=1}^{k-1} n_i}, \frac{\bar{y}_i}{\sum_{i=1}^{k-1} \bar{y}_i}\right),$$

which generalizes to

$$w(ab, cd) = w(a, c)w(b, d)$$

for all $a, b, d, c \in (0, 1)$. Let $a = e^{x_1}$, $b = e^{x_2}$, $c = e^{x_3}$, $d = e^{x_4}$. Then, we have

$$\varphi(x_1 + x_2, x_3 + x_4) = \varphi(x_1, x_3) + \varphi(x_2, x_4)$$

for all $x_1, x_2, x_3, x_4 \in (-\infty, 0)$ where $\varphi(x, y) := \ln w(e^x, e^y)$. By Aczél and Dhombres (1989) Corollary 2 on page 35, we get $\varphi(x, y) = c_1 x + c_2 y$ which implies $w(\pi, \theta) = \pi^{c_1} \theta^{c_2}$ for all $\pi, \theta \in (0, 1)$.

Now, using the weak reflexivity, we get

$$w(t\pi, t\theta) + w((1-t)\pi, (1-t)\theta) = w(\pi, \theta)$$

or

$$(t\pi)^{c_1}(t\theta)^{c_2} + ((1-t)\pi)^{c_1}((1-t)\theta)^{c_2} = \pi^{c_1}\theta^{c_2}$$

which is equivalent to

$$t^{c_1+c_2} + (1-t)^{c_1+c_2} = 1.$$

Since the left-hand side is strictly increasing in $c_1 + c_2$, then $c_1 + c_2 = 1$ is the unique solution. Thus, we have

$$W_k((I_1, n_1, \bar{y}_1), \dots, (I_k, n_k, \bar{y}_k)) = f^{-1}\left(\sum_{i=1}^k \pi_i^{1-\alpha} \theta_i^\alpha f(I_i)\right)$$

for all $k \geq 2$ and $(I_i, n_i, \bar{y}_i)_{i=1}^k \in \mathbb{R}_{++}^{3k}$. □

Due to continuity of W_k , these results extend to cases where $n_i = 0$ or $\bar{y}_i = 0$ for some i .

A.2 Proof of Proposition 1

The within-group inequality term in proposition 1 is an arithmetic average of subgroup inequalities with a Cobb-Douglas weight structure in the subgroups' income and population shares. It therefore satisfies axioms 1 through 6. Because generalized entropy indices are means-aggregative, axioms 7 requires that between-group inequality is zero when all subgroups have equal means. The between-group inequality term in proposition 1 is equal to aggregate inequality if all income are replaced by the subgroup means and is therefore indeed zero if all subgroups have equal means.

Next, we show that (2) is the unique decomposition for generalized entropy indices that satisfies axioms 1-8. Let $\alpha \in \mathbb{R}$ and suppose that $GE_\alpha(Y) = W(Y) + B(Y)$ is a decomposition of a generalized entropy index of parameter α that satisfies axioms 1-8. Then, by axiom 8, B must be invariant under transfers within subgroups. Hence, we can set $GE_\alpha(Y_i) = 0$ for all $i = 1, \dots, k$ without changing the value of B . By the zero within-group inequality axiom, $W = 0$ and thus we get $B(Y) = GE_\alpha(\tilde{Y})$. Finally, we get $W = \sum_{i=1}^k \pi_i^{1-\alpha} \theta_i^\alpha GE_\alpha(Y_i)$ by subtracting B from $GE_\alpha(Y)$.

A.3 Proof of Proposition 2

Let $q \in \mathbb{R}$. By equation (3), the Foster-Shneyerov index with parameter q is Ω -aggregative with Ω containing the q th order moment.

Theorem 1 is satisfied with $f(x) = x$ and $\alpha = 0$. Thus, (3) satisfies axioms 1-6. The inequality of power means²⁴ states that power means with different exponents are equal if and only if all arguments are equal. Thus, between-group inequality is zero if and only if all subgroups have equal moments of order q , that is, if subgroup distributions are Ω -equivalent. Thus, axiom 7 holds. Finally, the condition in axiom 8 never holds for any Foster-Shneyerov index with $q \neq 1$, and in the case of $q = 1$, the index and the decomposition coincides with the generalized entropy index of parameter $\alpha = 0$, which is shown to satisfy all axioms in proposition 1.

For uniqueness, let $FS_q(Y) = W(Y) + B(Y)$ be a decomposition that satisfies axioms 1-8. Now, consider a case where all subgroups are Ω -equivalent, that is, have equal q -means. Then, equation (3) reduces to

$$FS_q(Y) = \sum_i \pi_i FS_q(Y_i)$$

for all $\pi_i \in [0, 1]$ and $FS_q(Y_i) \in \mathbb{R}_+$. By axiom 7, between-group inequality is equal to zero and we get $W = \sum_i \pi_i FS_q(Y_i)$. Using theorem 1, we get

$$f^{-1} \left(\sum_{i=1}^k \pi_i^{1-\alpha} \theta_i^\alpha f(FS_q(Y_i)) \right) = \sum_i \pi_i FS_q(Y_i)$$

or

$$\sum_i \pi_i^{1-\alpha} \theta_i^\alpha f(FS_q(Y_i)) = f \left(\sum_i \pi_i FS_q(Y_i) \right) \quad (10)$$

for all $\pi_i \in [0, 1]$, $\theta_i \in [0, 1]$, and $FS_q(Y_i) \in \mathbb{R}_+$, where f is a strictly monotonic function with $f(0) = 0$ and $\alpha \in \mathbb{R}$. Since cf gives the same W as f for any $c \neq 0$, we can assume $f(1) = 1$ without loss of generality. By setting $FS_q(Y_i) = 1$ for some i and $FS_q(Y_j) = 0$ for all $j \neq i$, we get $\pi_i^{1-\alpha} \theta_i^\alpha = f(\pi_i)$ for all $\pi_i, \theta_i \in [0, 1]$. Thus, we must have $\alpha = 0$ and hence $f(x) = x$ for all $x \in [0, 1]$. Now, equation (10) becomes

$$\sum_i \pi_i f(FS_q(Y_i)) = f \left(\sum_i \pi_i FS_q(Y_i) \right). \quad (11)$$

²⁴see e.g. Theorem 1 on p. 203 in Bullen (2013)

for all $\pi_i \in [0, 1]$ and $FS_q(Y_i) \in \mathbb{R}_+$. The only strictly monotonic solution to equation (11) is $f(x) = ax$ for some $a \neq 0$ ²⁵. With normalization $f(1) = 1$ we get $a = 1$, and thus $W = \sum_i \pi_i FS_q(Y_i)$ and $B = FS_q(\mu_{q,i} u_{n_i})$.

A.4 Proof of Proposition 3

Suppose that $A_\varepsilon(Y) = W(Y) + B(Y)$ is a decomposition of Atkinson index of parameter ε that satisfies axioms 1-8. Since Atkinson index for any $\varepsilon < 1$ is a monotonic transformation of a corresponding generalized entropy indices, it is means-aggregative. Thus, by axiom 8, we get that B is always independent of subgroup income distributions. We may thus redistribute incomes such that $A_\varepsilon(Y_i) = 0$ for all $i = 1, \dots, k$ without changing the value of B . Hence, we get $B(Y) = A_\varepsilon(\bar{Y})$, where \bar{Y} is derived from Y by replacing each individual's income by the respective subgroup mean. By subtracting B from A_ε , we get

$$W(Y) = \left(\sum_{i=1}^n \pi_i^\varepsilon \theta_i^{1-\varepsilon} \right)^{\frac{1}{1-\varepsilon}} - \left(\sum_{i=1}^n \pi_i^\varepsilon \theta_i^{1-\varepsilon} (1 - A_{\varepsilon,i})^{1-\varepsilon} \right)^{\frac{1}{1-\varepsilon}}.$$

Now, W does not satisfy axiom 6 (replacement), which is a contradiction. Thus, there is no decomposition for Atkinson indices that satisfies all the proposed axioms.

A.5 Proof of Proposition 5

Transfers that increase subgroup inequality by the Lorenz criterion shift the subgroup Lorenz curves outwards so that the pre-transfers Lorenz region are subsets of the post-transfers Lorenz regions. Since the Minkowski addition is monotonic, that is $\text{Vol}(A \oplus B) \leq \text{Vol}(\tilde{A} \oplus \tilde{B})$ if $A \subseteq \tilde{A}$ and $B \subseteq \tilde{B}$, such transfers also shift the population Lorenz curve outwards. Hence, the aggregate Gini coefficient must increase.

A.6 Proof of Proposition 6

The inequality follows directly from the Brunn-Minkowski theorem.

A.7 Proof of Proposition 7

In the proof, we use a following lemma.

²⁵see Aczél (1966).

Lemma 5. *There exist two distinct distributions that share the same moments and Gini coefficients, which are all finite.*

Proof. Let X_c be a distribution with density function

$$f_c(x) = (1 + c \sin(2\pi \ln(x))) \frac{1}{\sqrt{2\pi}} \chi_{[0,\infty)}(x) \frac{1}{x} e^{-\frac{(\ln(x))^2}{2}}$$

that depends on parameter c . Note that X_0 is the lognormal distribution corresponding to parameters $\mu = 0$, $\sigma = 1$. It can be shown that²⁶ for $c \in [-1, 1]$, all moments of X_c are finite and do not depend on c . We need to show that the Gini coefficients of X_c are equal for two different values of c . Since the Gini coefficient can be written as

$$G = 1 - 2 \int_0^1 L(F) dF,$$

it suffices to show that $\int_0^1 L_c(F) dF$ is equal for two different values of c , where L_c is the Lorenz curve of X_c .

Since

$$L_c(F(x)) = \frac{1}{\mu} \int_0^x t f_c(t) dt$$

where μ does not depend on c , we need to show that there exist two distinct values of c for which

$$A_c := \int_0^1 \int_0^{F^{-1}(x)} t f_c(t) dt dx$$

are equal. Now,

$$\begin{aligned} A_c &= \int_0^1 \int_0^{F^{-1}(x)} t f_c(t) dt dx = \int_0^1 \int_0^{F^{-1}(x)} (1 + c \sin(2\pi \ln(t))) \frac{1}{\sqrt{2\pi}} e^{-\frac{(\ln(t))^2}{2}} dt dx \\ &= \frac{1}{\sqrt{2\pi}} \int_0^\infty \int_{F(t)}^1 (1 + c \sin(2\pi \ln(t))) e^{-\frac{(\ln(t))^2}{2}} dx dt \\ &= \frac{1}{\sqrt{2\pi}} \int_0^\infty (1 + c \sin(2\pi \ln(t))) e^{-\frac{(\ln(t))^2}{2}} (1 - F(t)) dt \\ &= \frac{1}{\sqrt{2\pi}} \int_0^\infty (1 + c \sin(2\pi \ln(t))) e^{-\frac{(\ln(t))^2}{2}} \left(1 - \int_0^t (1 + c \sin(2\pi \ln(x))) \frac{1}{\sqrt{2\pi}} \frac{1}{x} e^{-\frac{(\ln(x))^2}{2}} dx \right) dt \\ &= \frac{1}{\sqrt{2\pi}} \int_0^\infty (1 + c \sin(2\pi \ln(t))) e^{-\frac{(\ln(t))^2}{2}} \left(1 - \int_{-\infty}^{\ln(t)} (1 + c \sin(2\pi s)) \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds \right) dt \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^\infty (1 + c \sin(2\pi z)) e^{-\frac{z^2}{2}} e^z \left(1 - \int_{-\infty}^z (1 + c \sin(2\pi s)) \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds \right) dz \end{aligned}$$

²⁶See, for example, Schmüdgen (2017).

$$= \frac{1}{\sqrt{2\pi}}(a + bc + dc^2),$$

where

$$\begin{aligned} a &= \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} e^z \left(1 - \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds \right) dz \\ b &= \int_{-\infty}^{\infty} \sin(2\pi z) e^{-\frac{z^2}{2}} e^z dz - \int_{-\infty}^{\infty} \sin(2\pi z) e^{-\frac{z^2}{2}} e^z \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds dz \\ &\quad - \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} e^z \int_{-\infty}^z \sin(2\pi s) \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds dz \\ c &= \int_{-\infty}^{\infty} \sin(2\pi z) e^{-\frac{z^2}{2}} e^z \int_{-\infty}^z \sin(2\pi s) \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds dz. \end{aligned}$$

Now, if $b = 0$, then we have the result since either $d = 0$ and A_c does not depend on c or $d \neq 0$ and $A_c = A_{-c}$ for any $c \in [-1, 1]$.

First, note that

$$\int_{-\infty}^{\infty} \sin(2\pi z) e^{-\frac{z^2}{2}} e^z dz = \int_{-\infty}^{\infty} \sin(2\pi z) e^{-\frac{1}{2}(z-1)^2} e^{\frac{1}{2}} dz = e^{\frac{1}{2}} \int_{-\infty}^{\infty} \sin(2\pi x) e^{-\frac{1}{2}(x)^2} dx = 0$$

as an integral of an odd function. Thus,

$$\begin{aligned} b &= \int_{-\infty}^{\infty} \sin(2\pi z) e^{-\frac{z^2}{2}} e^z \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds dz - \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} e^z \int_{-\infty}^z \sin(2\pi s) \frac{1}{\sqrt{2\pi}} e^{-\frac{s^2}{2}} ds dz \\ &= -e^{\frac{1}{2}} \int_{-\infty}^{\infty} \int_{-\infty}^z e^{-\frac{1}{2}(z-1)^2} e^{-\frac{s^2}{2}} (\sin(2\pi z) + \sin(2\pi s)) ds dz \\ &= -e^{\frac{1}{2}} \int_{-\infty}^{\infty} \int_{-\infty}^{x+1} e^{-\frac{x^2}{2}} e^{-\frac{s^2}{2}} (\sin(2\pi x) + \sin(2\pi s)) ds dx \\ &= -e^{\frac{1}{2}} \int_{-\infty}^1 \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} e^{-\frac{(x+t)^2}{2}} (\sin(2\pi x) + \sin(2\pi(x+t))) dx dt \\ &= -e^{\frac{1}{2}} \int_{-\infty}^1 e^{-\frac{1}{4}t^2} \int_{-\infty}^{\infty} e^{-(x+\frac{1}{2}t)^2} (\sin(2\pi x) + \sin(2\pi(x+t))) dx dt \\ &= -e^{\frac{1}{2}} \int_{-\infty}^1 e^{-\frac{1}{4}t^2} \int_{-\infty}^{\infty} e^{-(x+\frac{1}{2}t)^2} (2\sin(\pi x + \pi(x+t)) \cos(\pi x - (\pi(x+t)))) dx dt \\ &= -2e^{\frac{1}{2}} \int_{-\infty}^1 e^{-\frac{1}{4}t^2} \int_{-\infty}^{\infty} e^{-(x+\frac{1}{2}t)^2} (\sin(2\pi(x+\frac{1}{2}t)) \cos(-\pi t)) dx dt \\ &= -2e^{\frac{1}{2}} \int_{-\infty}^1 e^{-\frac{1}{4}t^2} \cos(-\pi t) \int_{-\infty}^{\infty} e^{-s^2} \sin(2\pi s) ds dt = 0, \end{aligned}$$

since $\int_{-\infty}^{\infty} e^{-s^2} \sin(2\pi s) ds = 0$ as an integral of an odd function. □

Let $k \in \mathbb{N}$ and suppose there exists a set Ω of k moments such that the Gini coefficient is Ω -aggregative. Suppose Y_1 and Y_2 are distinct distributions that have the same moments and the same Gini coefficients. Then Y_1 and Y_2 have different Lorenz curves. Consider a population partitioned into subgroups such that each subgroup has distribution Y_1 and denote the set of k subgroup moments of Y_1 by Ω_1 . Then, the aggregate Gini coefficient of the population is given by $G = F(G(Y_1), \dots, G(Y_1), \Omega_1, \dots, \Omega_1; \pi_1, \dots, \pi_n)$. By Proposition 6, we have

$$F(G(Y_1), \dots, G(Y_1), \Omega_1, \dots, \Omega_1; \pi_1, \dots, \pi_n) = \left(\sum_{i=1}^n \sqrt{\pi_i \theta_i G(Y_1)} \right)^2 = G(Y_1),$$

since subgroup Lorenz curves are homothetic and $\theta_i = \pi_i$ for all i .

Next, consider another population partitioned into subgroups, where one subgroup has distribution Y_1 and the other subgroups have distribution Y_2 , and let \tilde{G} denote the aggregate Gini coefficient of the population. Since the subgroup Gini coefficients and moments are the same as before, we have

$$\tilde{G} = F(G(Y_1), \dots, G(Y_1), \Omega_1, \dots, \Omega_1; \pi_1, \dots, \pi_n) = G(Y_1).$$

But since the subgroup Lorenz curves are not all homothetic, then Proposition 6 theorem implies

$$\tilde{G} > G(Y_1),$$

which is a contradiction. Therefore, the Gini coefficient is Ω -nonaggregative.

A.8 Proof of Proposition 8

Let us first show that the decomposition with within-group inequality term (4) satisfies all the axioms. Since the within-group inequality term satisfies theorem 1 with $\alpha = \frac{1}{2}$ and $f(x) = x^{\frac{1}{2}}$, then (4) satisfies axioms 1-6.

As the Gini coefficient is Ω -nonaggregative, subgroup distributions are Ω -equivalent only if they are the same distribution. In this case, all subgroups have equal Gini coefficients and income share of each subgroup equals its population share. Then, both aggregate Gini coefficient and the within-group inequality term in (4) are equal to the common subgroup Gini coefficient. Thus, between-group inequality is equal to zero and axiom 7 holds.

Bhattacharya and Mahalanobis (1967) show that the aggregate Gini coefficient can

be written as

$$G(Y) = \sum_i \pi_i \theta_i G(Y_i) + G(\bar{Y}) + R, \quad (12)$$

where \bar{Y} is derived from Y by replacing each individuals income by the relevant subgroup mean and R depends on the amount of overlap between distributions and is zero if and only if there is no overlap in incomes between subgroups. Hence, the condition in axiom 8 holds only if there can be no overlap in subgroup incomes. It is possible to create overlap with within-group transfers that keep subgroup Gini coefficients constant whenever two distributions have both positive population share and income share. Thus, Gini coefficient is means-aggregative only in the case of only one group, j , having both non-zero population share and non-zero income share. In this special case, equation (12) reduces to $G(Y) = \pi_j \theta_j G(Y_j) + G(\bar{Y})$ and the within-part in equation (4) reduces to $G^W(Y) = \pi_j \theta_j G(Y_j)$. Therefore, we get that $G^B(Y) = G(\bar{Y})$, which is invariant under within-group transfers. Hence, axiom 8 holds.

We next show that (4) is the unique within-group inequality term for the Gini coefficient that satisfies axioms 1-8. Suppose W is a within-part for the Gini coefficient that satisfies all eight axioms. First, by theorem 1, $W(G_1, \dots, G_k, n_1, \dots, n_k, \bar{y}_1, \dots, \bar{y}_k) = f^{-1}(\sum_{i=1}^k \pi_i^{1-\alpha} \theta_i^\alpha f(G_i))$, for all $(G_i, n_i, \bar{y}_i)_{i=1}^k \in D^k$, where f is a strictly monotonic function with $f(0) = 0$ and $\alpha \in \mathbb{R}$. Since cf gives the same W as f for any $c \neq 0$, we can assume $f(1) = 1$ without loss of generality.

Now, if there is only one group, i , with both non-zero population share and non-zero income share, then by axiom 8 and equation (12), we get

$$W = f^{-1}(\pi_i^{1-\alpha} \theta_i^\alpha f(G(Y_i))) = \pi_i \theta_i G(Y_i)$$

for all $\pi_i, \theta_i, G(Y_i) \in [0, 1]$. By taking f of both sides, we get

$$\pi_i^{1-\alpha} \theta_i^\alpha f(G(Y_i)) = f(\pi_i \theta_i G(Y_i)) \quad (13)$$

for all $\pi_i, \theta_i, G(Y_i) \in [0, 1]$. Further, by setting $G(Y_i) = 1$ we get

$$\pi_i^{1-\alpha} \theta_i^\alpha = f(\pi_i \theta_i) \quad (14)$$

for all $\pi_i, \theta_i \in [0, 1]$. By setting $\pi_i = 1$ we get $f(x) = x^\alpha$ for all $x \in [0, 1]$ and by setting $\theta_i = 1$ we get $f(x) = x^{1-\alpha}$ for all $x \in [0, 1]$. Thus, we have $\alpha = 1 - \alpha$ which implies $\alpha = \frac{1}{2}$ and $f(x) = x^{\frac{1}{2}}$ for all $x \in [0, 1]$. Therefore, the decomposition with the

within-part $G^W(Y) = \left(\sum_{i=1}^n \sqrt{\pi_i \theta_i G(Y_i)} \right)^2$ is the unique decomposition for the Gini coefficient that satisfies axioms 1-8.

A.9 Proof of Proposition 9

Randomly permuting the group affiliations of a random sample of an infinite population is equivalent to replacing the incomes of that sample with a random draw from the aggregate income distribution. Equivalency between these two operations is a direct consequence of the anonymity property of inequality indices. That is, $I(P(y_1, \dots, y_n)) = I(y_1, \dots, y_n)$ for any permutation P . To prove proposition 9, it is useful to first prove the following two lemmas.

Lemma 6. *In an infinite population partitioned into k subgroups, drawing a random sample (with mass greater than zero) from the population and assigning the sampled observations to a new subgroup increases the within-group inequality term of the Gini coefficient (while keeping the aggregate Gini coefficient constant). That is,*

$$\begin{aligned} & G^W((G_1, (1-\alpha)\pi_1, (1-\alpha)\theta_1), \dots, (G_k, (1-\alpha)\pi_k, (1-\alpha)\theta_k), (G, \alpha, \alpha)) \\ & \geq G^W((G_1, \pi_1, \theta_1), \dots, (G_k, \pi_k, \theta_k)) \end{aligned}$$

for all G_i, π_i, θ_i and $\alpha \in (0, 1]$.

Proof. Using the definition of the within-part of the Gini coefficient in equation (4) and the Brunn-Minkowski theorem, we get

$$\begin{aligned} & G^W((G_1, (1-\alpha)\pi_1, (1-\alpha)\theta_1), \dots, (G_k, (1-\alpha)\pi_k, (1-\alpha)\theta_k), (G, \alpha, \alpha)) \\ & = \left((1-\alpha) \sum_{i=1}^k \sqrt{\pi_i \theta_i G(Y_i)} + \alpha \sqrt{G(Y)} \right)^2 \\ & \geq \left((1-\alpha) \sum_{i=1}^k \sqrt{\pi_i \theta_i G(Y_i)} + \alpha \sum_{i=1}^k \sqrt{\pi_i \theta_i G(Y_i)} \right)^2 \\ & = \left(\sum_{i=1}^k \sqrt{\pi_i \theta_i G(Y_i)} \right)^2 \\ & = G^W((G_1, \pi_1, \theta_1), \dots, (G_k, \pi_k, \theta_k)). \end{aligned}$$

□

Lemma 7. *Splitting a subgroup into $m \geq 2$ subgroups weakly reduces the within-group inequality of the Gini coefficient. That is,*

$$\begin{aligned} & G^W((G_1, \pi_1, \theta_1), \dots, (G_k, \pi_k, \theta_k)) \\ & \geq G^W((\tilde{G}_1, a_1\pi_1, a_1\theta_1), \dots, (\tilde{G}_m, a_m\pi_1, a_m\theta_1), (G_2, \pi_2, \theta_2), \dots, (G_k, \pi_k, \theta_k)), \end{aligned}$$

for all G_i, π_i, θ_i , where $G_1 = G(Y_1)$, $\tilde{G}_i = G(\tilde{Y}_i)$ for $i = 1, \dots, m$ with $Y_1 = \tilde{Y}_1 \cup \tilde{Y}_2 \cup \dots \cup \tilde{Y}_m$, and $\sum_{j=1}^m a_j = 1$. Equality holds if and only if $\tilde{Y}_1, \tilde{Y}_2, \dots, \tilde{Y}_m$ are identical distributions.

Proof. For the ease of notation, we prove the lemma for the case $m = 2$. Other cases are shown analogously.

Let Y_1 be the income distribution of subgroup 1 and let $\tilde{Y}_1 \cup \tilde{Y}_2$. Let $\tilde{\Lambda}_1$ and $\tilde{\Lambda}_2$ be the Lorenz areas of \tilde{Y}_1 and \tilde{Y}_2 . Now,

$$\text{Vol}(\tilde{\Lambda}_1 \oplus \tilde{\Lambda}_2) = \text{Vol}(\Lambda_1) = (\tilde{\pi}_1 + \tilde{\pi}_2)(\tilde{\theta}_1 + \tilde{\theta}_2)G(\tilde{Y}_1 \cup \tilde{Y}_2),$$

where $\tilde{\pi}_i = \frac{a_i\pi_1}{\pi_1+\pi_2}$ and $\tilde{\theta}_i = \frac{a_i\theta_1}{\theta_1+\theta_2}$ for $i = 1, 2$. By the Brunn-Minkowski inequality

$$\text{Vol}(\tilde{\Lambda}_1 \oplus \tilde{\Lambda}_2) \geq \left(\sqrt{\tilde{\pi}_1\tilde{\theta}_1\tilde{G}_1} + \sqrt{\tilde{\pi}_2\tilde{\theta}_2\tilde{G}_2} \right)^2,$$

where the inequality holds as equality if and only if the Lorenz areas of groups 1 and 2 are homothetic, or if the distributions are identical. Thus,

$$\begin{aligned} & G^W((G_1, \pi_1, \theta_1), (G_2, \pi_2, \theta_2)) \\ & = \left(\sqrt{\pi_1\theta_1G_1} + \sqrt{\pi_2\theta_2G_2} \right)^2 \\ & = \left(\sqrt{(\tilde{\pi}_1 + \tilde{\pi}_2)(\tilde{\theta}_1 + \tilde{\theta}_2)G(\tilde{Y}_1 \cup \tilde{Y}_2)} + \sqrt{\pi_2\theta_2G(Y_2)} \right)^2 \\ & = \left(\sqrt{\text{Vol}(\tilde{\Lambda}_1 \oplus \tilde{\Lambda}_2)} + \sqrt{\text{Vol}(\Lambda_2)} \right)^2 \\ & \geq \left(\sqrt{\tilde{\pi}_1\tilde{\theta}_1\tilde{G}_1^{\frac{1}{2}}} + \sqrt{\tilde{\pi}_2\tilde{\theta}_2\tilde{G}_2^{\frac{1}{2}}} + \sqrt{\pi_2\theta_2G_2^{\frac{1}{2}}} \right)^2 \\ & = G^W((\tilde{G}_1, a_1\pi_1, a_1\theta_1), (\tilde{G}_2, a_2\pi_2, a_2\theta_2), (G_2, \pi_2, \theta_2)), \end{aligned}$$

with equality holding if and only if subgroups 1 and 2 have identical distributions. □

Let G denote the aggregate Gini coefficient. Now, by the subgroup replication

principle, we have

$$\begin{aligned} G^W &((G_1, (1 - \alpha)\pi_1, (1 - \alpha)\theta_1), \dots, (G_k, (1 - \alpha)\pi_k, (1 - \alpha)\theta_k), (G, \alpha, \alpha)) \\ &= G^W((G_1, (1 - \alpha)\pi_1, (1 - \alpha)\theta_1), \dots, (G_k, (1 - \alpha)\pi_k, (1 - \alpha)\theta_k), (G, \alpha\pi_1, \alpha\pi_1), \dots, (G, \alpha\pi_k, \alpha\pi_k)). \end{aligned}$$

Let \tilde{G}_i denote the resulting Gini coefficient after $1 - \alpha$ sample of subgroup i 's distribution is merged with $\alpha\pi_i$ sample of aggregate distribution. By lemma 7, merging any two groups must strictly increase within-group inequality unless the two groups have the same distribution of income, Thus, we get

$$\begin{aligned} G^W &((G_1, (1 - \alpha)\pi_1, (1 - \alpha)\theta_1), \dots, (G_k, (1 - \alpha)\pi_k, (1 - \alpha)\theta_k), (G, \alpha\pi_1, \alpha\pi_1), \dots, (G, \alpha\pi_k, \alpha\pi_k)) \\ &\leq G^W((\tilde{G}_1, \pi_1, \theta_1), \dots, (\tilde{G}_k, \pi_k, \theta_k)) \end{aligned}$$

Equality holds only if all subgroup have the same distribution of income. As the aggregate Gini coefficient is unchanged while the within-group inequality term increases, between-group inequality must decrease.

A.10 Proof of Proposition 10

We consider a redistribution scheme in which any income y_i is replaced by an income $\tilde{y}_i = y_i - \alpha(y_i - \mu)$ for $\alpha \in [0, 1]$. Let \tilde{G} denote the Gini coefficient after this operation. First, note that the redistribution scheme reduces the aggregate Gini coefficient by α percent.

$$\begin{aligned} \tilde{G} &= \frac{\mathbb{E}[|\tilde{y}_i - \tilde{y}_j|]}{2\mu} \\ &= \frac{\mathbb{E}[|y_i - \alpha(y_i - \mu) - (y_j - \alpha(y_j - \mu))|]}{2\mu} \\ &= (1 - \alpha) \frac{\mathbb{E}[|y_i - y_j|]}{2\mu} \end{aligned}$$

Now, it is easy to see that within group inequality also decreases by α percent under this redistribution scheme.

$$\begin{aligned}
\tilde{G}^W &= \left(\sum_m \sqrt{\pi_m \tilde{\theta}_m \tilde{G}_m} \right)^2 \\
&= \left(\sum_m \sqrt{\pi_m \tilde{\theta}_m \frac{\mathbb{E}_m[|\tilde{y}_i - \tilde{y}_j|]}{2\tilde{\mu}_m}} \right)^2 \\
&= \left(\sum_m \pi_m \sqrt{\frac{\mathbb{E}_m[|\tilde{y}_i - \tilde{y}_j|]}{2\mu}} \right)^2 \\
&= \left(\sum_m \pi_m \sqrt{(1-\alpha) \frac{\mathbb{E}_m[|y_i - y_j|]}{2\mu}} \right)^2 \\
&= (1-\alpha) \left(\sum_m \sqrt{\pi_m \theta_m G_m} \right)^2 \\
&= (1-\alpha) G^W
\end{aligned}$$

where $\tilde{\theta}_m$ is subgroup m 's income share after the replacement of incomes. Finally, since both the aggregate Gini coefficient as well as the within group term decrease by α percent, it must be the case the the between group terms also decreases by α percent.

A.11 Proof of Proposition 11

Note that the within group inequality term can be written as follows.

$$\begin{aligned}
G^W &= \left(\sum_m \sqrt{\pi_m \theta_m G_m} \right)^2 \\
&= \left(\sum_m \sqrt{\pi_m \theta_m \frac{\mathbb{E}_m[|y_i - y_j|]}{2\mu_m}} \right)^2 \\
&= \left(\sum_m \pi_m \sqrt{\frac{\mathbb{E}_m[|y_i - y_j|]}{2\mu}} \right)^2
\end{aligned}$$

But since lump-sum transfers between groups affect neither the mean absolute difference within groups nor the mean income in the population, it follows that such transfers do not affect within group inequality.

A.12 Proof of Proposition 12

Scale invariance follows directly from the scale invariance of the within-group inequality term and the inequality measure itself. To show translation invariance, note first that due to the arithmetic definition of the Gini coefficient in (5), adding some fixed amount z to each income in a population with average income equal to μ decreases the Gini coefficient by a factor of $\mu/(\mu + z)$. Let \tilde{G}_i and $\tilde{\theta}_i$ denote the Gini coefficient and the income share of subgroup i after translation. Within-group inequality after translation can then be written as follows

$$\begin{aligned}
\tilde{G}^W &= \left(\sum_i \sqrt{\pi_i \tilde{\theta}_i \tilde{G}_i} \right)^2 \\
&= \left(\sum_i \sqrt{\pi_i \left(\pi_i \frac{\mu_i + z}{\mu + z} \right) \left(\frac{\mu_i}{\mu_i + z} G_i \right)} \right)^2 \\
&= \left(\sum_i \pi_i \sqrt{\frac{\mu_i}{\mu + z} G_i} \right)^2 \\
&= \frac{1}{\mu + z} \left(\sum_i \pi_i \sqrt{\mu_i G_i} \right)^2 \\
&= \frac{1}{\mu + z} \left(\sum_i \pi_i \sqrt{\frac{\theta_i}{\pi_i} \mu G_i} \right)^2 \\
&= \frac{\mu}{\mu + z} \left(\sum_i \sqrt{\pi_i \theta_i G_i} \right)^2 \\
&= \frac{\mu}{\mu + z} G^W.
\end{aligned}$$

Within-group inequality also decreases by a same factor of $\mu/(\mu + z)$. Hence, the ratio of within-group inequality to between-group inequality remains unchanged.