

# Descriptive Units of Heterogeneity: An Axiomatic Approach to Measuring Heterogeneity

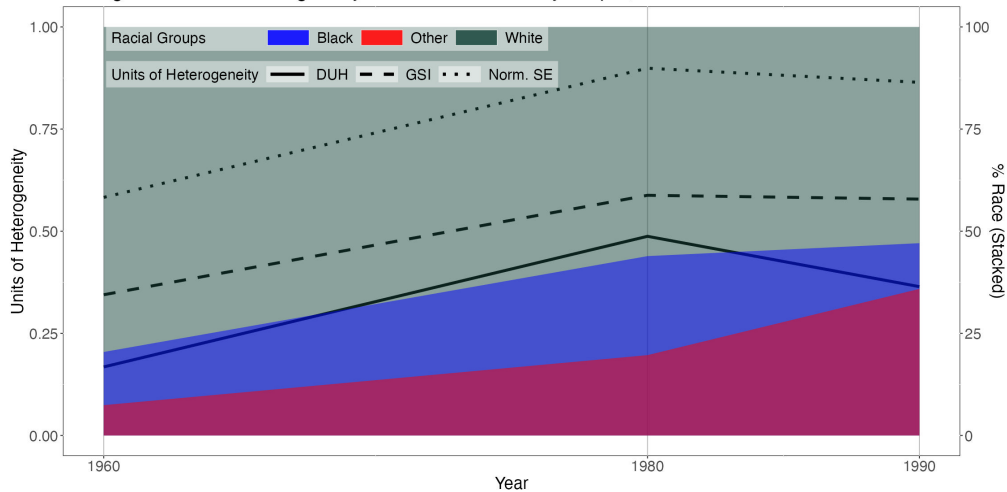
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# Empirical Motivation

Changes in Racial Heterogeneity in San Francisco City Proper, 1960-1990



Data: US Census Decennial Census, 1960-1990

# What is Heterogeneity in a System?

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$\forall s \in \mathcal{S}$ , denote

- The length of  $s$  as  $|s|$ .
- The “total population” of  $s$  as  $\|s\|_1 = \sum_{g=1}^{|s|} s_g$ .
- The mean group size of  $s$  as  $\mu(s) = \frac{\|s\|_1}{|s|}$ .

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A function  $\Phi : \mathcal{S} \rightarrow \mathbb{R}$  is a measure of heterogeneity if,  $\forall s, s' \in \mathcal{S}$ ,

$$\Phi(s) \geq \Phi(s') \iff s \text{ is weakly more heterogeneous } (\succsim) \text{ than } s'.$$

# Defining Maximum and Minimum Heterogeneity

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A system  $\underline{s}$  is minimally heterogeneous (or perfectly homogeneous) if

$$\exists k \in \mathbb{R}_{++}, \underline{s} = k \cdot (0, \dots, 0, k, 0, \dots, 0).$$

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**Problem:** If we treat heterogeneity as a distributional property without value-judgments, then there are cases when dispersion and deconcentration are insufficient.

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- Requires distributional assumptions (Gini).
- Require arbitrary normative judgments (Atkinson and Generalized Gini).
- Require implicit order of outcomes (Atkinson, Generalized Gini, and Gini).
- Lacks decomposability and the sensitivity to transfers (Hoover).

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Measures of concentration quantify heterogeneity by the non-dominance of one or a few groups in a system (deconcentration).

- Not sensitive to redistribution between small groups when one group is sufficiently large (HHI and SE) .
- Discard information provided by the presence of groups with 0 population (HHI and SE).

# Descriptive Units of Heterogeneity

Let  $\sigma$  of  $s$  be the permutation of  $s$  such that  $\sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_{|s|}$ .  $\sigma$  is called the *ordered system* of  $s$ .

Define

$$\hat{\sigma}_1 = \frac{\sigma_1}{||s||_1} \text{ and } \tilde{\sigma}_g = \frac{\sigma_g}{||s||_1 - \sigma_1}$$

The Descriptive Units of Heterogeneity (DUH) is defined as:

$$DUH_p(s) = \frac{\ln(\hat{\sigma}_1)}{\ln(|s|)} \cdot \left[ \left( \sum_{g=2}^{|s|} \left| \tilde{\sigma}_g - \frac{1}{|s| - 1} \right|^p \right)^{\frac{1}{p}} - 1 \right]$$

# Fundamental Axioms

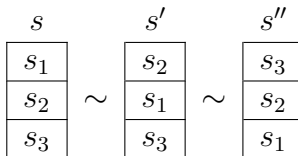
Axioms that induce partial ordering by pinning down

- 1 When two systems of  $|s|$  groups are equally heterogeneous
- 2 How to order the heterogeneity of two systems of  $|s|$  groups that are marginally different

# [GSYM] Group Symmetry

For any permutation  $\pi(s)$  of  $s$ ,  $\Phi(s) = \Phi(\pi(s))$ .

For example, take  $s, s', s'' \in \mathcal{S}$ ,



# [INV] Scale Invariance

A measure of heterogeneity  $\Phi$  satisfies the property of Scale Invariance if for any system  $s$  and a scalar  $\lambda \in \mathbb{R}_{++}$ ,  $\Phi(s) = \Phi(\lambda \cdot s)$ .

For example, take  $s, s' \in \mathcal{S}$ ,  $\lambda \in \mathbb{R}_{++}$ ,

$$\begin{array}{c|c} s & s' \\ \hline s_1 & \lambda s_1 \\ \hline s_2 & \lambda s_2 \\ \hline s_3 & \lambda s_3 \end{array} \sim$$



# [PT] Principle of Transfers

Let  $\sigma$  be the ordered system of  $s$ . Let  $e_i^{|s|}$  be an ordered tuple of length  $|s|$  such that its  $i^{th}$  element is some  $\varepsilon$  and the rest are 0. A measure of heterogeneity  $\Phi$  satisfies the Principle of Transfers if  $\forall i < j \leq |\sigma|$  and  $\varepsilon \in \mathbb{R}_+$ ,

$$\begin{cases} \sigma_i - \sigma_j \geq 2\varepsilon \\ \sigma_i - \sigma_{i+1} \geq \varepsilon \\ \sigma_{j-1} - \sigma_j \geq \varepsilon \end{cases} \quad \text{together imply } \Phi(\sigma) < \Phi(\sigma - e_i^{|s|} + e_j^{|s|}).$$

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$$\begin{array}{|c|} \hline \sigma \\ \hline \sigma_1 \\ \hline \sigma_2 \\ \hline \sigma_3 \\ \hline \end{array} \prec \begin{array}{|c|} \hline \sigma' \\ \hline \sigma_1 - \varepsilon \\ \hline \sigma_2 + \varepsilon \\ \hline \sigma_3 \\ \hline \end{array} \prec \begin{array}{|c|} \hline \sigma'' \\ \hline \sigma_1 - \varepsilon \\ \hline \sigma_2 \\ \hline \sigma_3 + \varepsilon \\ \hline \end{array}, \quad \text{i.e. } \Phi(\sigma) < \Phi(\sigma') < \Phi(\sigma''),$$

# Characterization Axioms

Axioms that can be combined to uniquely characterize measures with total ordering and cardinal interpretation.

# Two Determinants of Heterogeneity

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- $$Gini(s) = 1 - \frac{1}{|s|^2} \cdot \left( \frac{\sum_{g=1}^{|s|} (2g-1)\sigma_g}{\mu(s)} \right) = 1 - \sum_{g=1}^{|s|} \frac{(2g-1)}{|s|} \hat{\sigma}_g.$$

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- $$Gini(s) = 1 - \frac{1}{|s|^2} \cdot \left( \frac{\sum_{g=1}^{|s|} (2g-1)\sigma_g}{\mu(s)} \right) = 1 - \sum_{g=1}^{|s|} \frac{(2g-1)}{|s|} \hat{\sigma}_g.$$

GSYM can be satisfied by identical arithmetic treatment to the same “type” of groups in functional form!

# What Matters in Measuring Heterogeneity?

The heterogeneity of system  $s$  is measured by  $\Phi : S \rightarrow \mathbb{R}$  such that

$$\Phi(s) \geq \Phi(s') \iff s \text{ is weakly more heterogeneous than } s'.$$

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- Relative size(s) of the minority group(s):  $\tilde{\sigma}_g = \frac{\sigma_g}{||s||_1 - \sigma_1}, g \in \{2, \dots, |s|\}$

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A measure of heterogeneity can be decomposed as  $\Phi = \Phi(\phi, \psi)$  where  $\phi : \mathcal{S} \rightarrow \mathbb{R}$  reflects the dominance of the largest group and  $\psi : \mathcal{S} \rightarrow \mathbb{R}$  reflects the contribution of the minority group distribution.

Let  $\sigma$  be the ordered system of  $s$ . A measure of heterogeneity  $\Phi(s)$  satisfies Independence if it is a composite function of  $\phi : \mathcal{S} \rightarrow \mathbb{R}$  and  $\psi : \mathcal{S} \rightarrow \mathbb{R}$  such that  $\psi(s) = \psi(c, \sigma_2, \dots, \sigma_{|s|})$ ,  $\forall c \in \mathbb{R}_{++}$ .

# Using Evenness for $\psi$

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**Definition:** Take  $p \in [1, \infty)$ . The function  $\psi_p : \mathcal{S} \rightarrow \mathbb{R}$  defined as:

$$\psi_p(s) = 1 - \left( \sum_{g=2}^{|s|} \left| \tilde{\sigma}_g - \frac{1}{|s| - 1} \right|^p \right)^{\frac{1}{p}}$$

is a measure of evenness in the distribution of minority groups.

# Proposition 1

Let  $s$  be an arbitrary system such that  $|s| \geq 4$ . Let  $\Phi$  be a measure of heterogeneity that satisfies GSYM, INV, and IND. Holding  $\hat{\sigma}_1$  constant, if  $\Phi$  is strictly increasing in  $\psi_p$ , then  $\Phi$  satisfies PT if and only if  $p > 1$ .

# [PPT] Principle of Proportional Transfers

Let  $\sigma$  be the ordered system of  $s$ . Let  $e_i^{|s|}$  be an ordered tuple of length  $|s|$  such that its  $i^{th}$  element is some  $\varepsilon$  and the rest are 0. A measure of heterogeneity  $\Phi(s)$  satisfies the Principle of Proportional Transfers if  $\forall \varepsilon \in \mathbb{R}_+, \exists \alpha \in \mathbb{R}_{++}$

$$\begin{cases} \frac{\sigma_1 - \varepsilon}{\|s\|_1} = \left( \frac{\sigma_1}{\|s\|_1} \right)^\alpha \\ \sigma_1 - \varepsilon \geq \sigma_2 + \tilde{\sigma}_2 \cdot \varepsilon \end{cases} \quad \text{together imply } \Phi \left( \sigma - e_1^{|s|} + \sum_{g=2}^{|s|} \tilde{\sigma}_g \cdot e_g^{|s|} \right) = \alpha \cdot \Phi(\sigma).$$

In other words, *holding the order of groups constant*, a transfer from the largest group proportionally to the minority groups that reduces  $\hat{\sigma}_1$  to  $(\hat{\sigma}_1)^\alpha$  increases heterogeneity by a factor of  $\alpha$ .

# PPT Example

$$\begin{array}{c} \sigma \\ \hline \tilde{\sigma}_1 \\ \hline \hat{\sigma}_2 \\ \hline \hat{\sigma}_3 \end{array} \prec \begin{array}{c} \sigma' \\ \hline \hat{\sigma}_1^\alpha \\ \hline \hat{\sigma}_2 + \frac{\hat{\sigma}_2}{\hat{\sigma}_2 + \hat{\sigma}_3} (\hat{\sigma}_1 - \hat{\sigma}_1^\alpha) \\ \hline \hat{\sigma}_3 + \frac{\hat{\sigma}_3}{\hat{\sigma}_2 + \hat{\sigma}_3} (\hat{\sigma}_1 - \hat{\sigma}_1^\alpha) \end{array} .$$

A measure  $\Phi$  satisfying GSYM, INV, and PPT would yield:

$$\Phi(\sigma) < \alpha \Phi(\sigma) = \Phi(\sigma') .$$



# Proposition 2

If a measure of heterogeneity  $\Phi$  satisfies GSYM, INV, IND, and PPT, then it must be  $\Phi = \phi \cdot \psi_p$  where  $\phi(s) = \phi(\sigma) = -c \cdot \log_q(\hat{\sigma}_1)$ ,  $c, q \in \mathbb{R}_{++}$ .

# [CON] Contractibility

Let  $s$  be an arbitrary system. Let  $s'$  be the concatenation of  $s$  and the tuple  $(0)$  such that  $s' = (s, 0)$ . Let  $\sigma$  and  $\sigma'$  denote the ordered systems of  $s$  and  $s'$ . A measure of heterogeneity  $\Phi$  satisfies Contractibility if

$$\sigma_2 > 0 \Rightarrow \Phi(\sigma') < \Phi(\sigma).$$

Let  $\sigma$  be the ordered system of  $s$ . A measure of heterogeneity  $\Phi$  satisfies Unity if

$$\Phi(s) = 0 \iff \hat{\sigma}_1 = 1$$

and

$$\Phi(s) = 1 \iff \hat{\sigma}_1 = \hat{\sigma}_2 = \cdots = \hat{\sigma}_{|s|} = \frac{1}{|s|}$$

# Descriptive Units of Heterogeneity

Let  $\sigma$  be the ordered system of  $s$ . Denote  $\hat{\sigma}_1 = \frac{\sigma_1}{\|s\|_1}$  and  $\tilde{\sigma}_g = \frac{\sigma_g}{\|s\|_1 - \sigma_1}$ , where  $g \in \{2, \dots, |s|\}$  and  $p \in (1, \infty)$ . The family of descriptive units of heterogeneity (DUH) of the system  $s$  with  $|s| \geq 2$  is:

$$DUH_p(s) = \frac{\ln(\hat{\sigma}_1)}{\ln(|s|)} \cdot \left[ \left( \sum_{g=2}^{|s|} \left| \tilde{\sigma}_g - \frac{1}{|s| - 1} \right|^p \right)^{\frac{1}{p}} - 1 \right].$$

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*Theorem:*

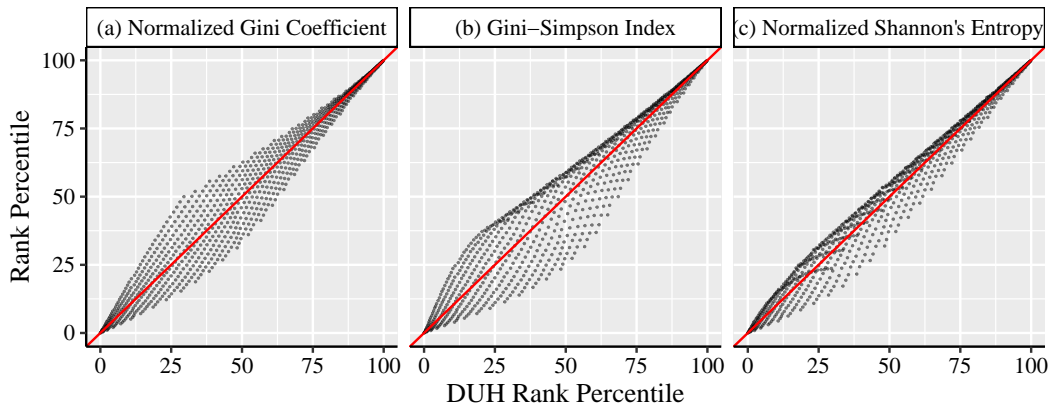
The descriptive units of heterogeneity constitute a uniquely determined family of measures—up to positive scalar multiplication—that incorporate evenness through the function  $\psi_p$ , and satisfy GSYM, INV, PT, IND, PPT, CON, and UNI.

# Existing Measures Satisfy Some Axioms

Type	Axiom	Gini	DUH	HHI	SE
<b>Fundamental</b>	Type Symmetry	✓	✓	✓	✓
	Scale Invariance	✓	✓	✓	✓
	Principle of Transfers	✓	✓	✓	✓
<b>Characterization</b>	Independence	×	✓	✓	✓
	Principle of Proportional Transfers	×	✓	×	×
	Contractibility	✓	✓	×	×
	Unity	×	✓	×	×
	Expandability	×	×	✓	✓
	Replication Principle	×	×	✓	×
	Shannon's Additivity	×	×	×	✓

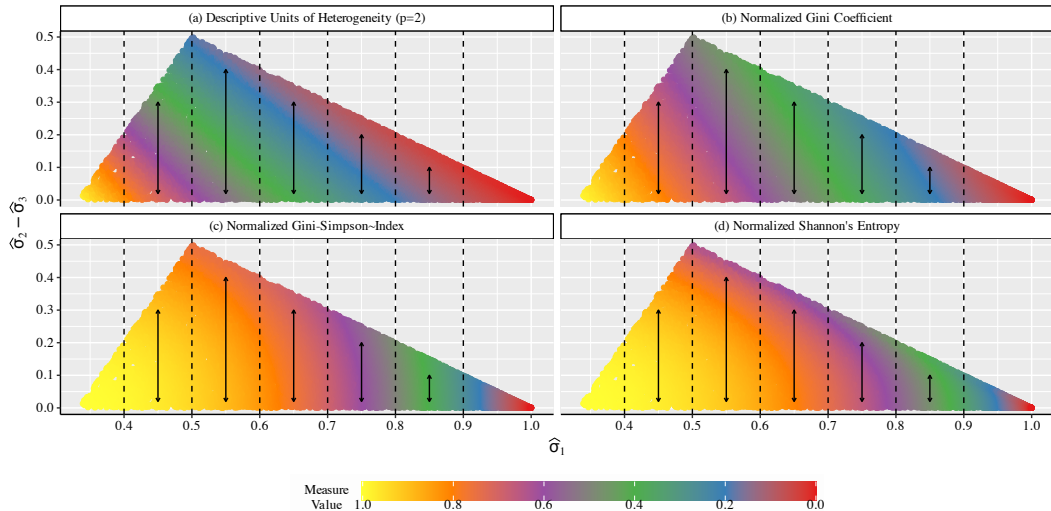
# Comparing Measures

Figure 1: Rank Correlation between Measures over Systems with  $|\sigma| = 3$  and  $\|\sigma\|_1 = 100$



# Comparing Measures

Figure 2: Comparison between DUH, Gini, GSI, and SE



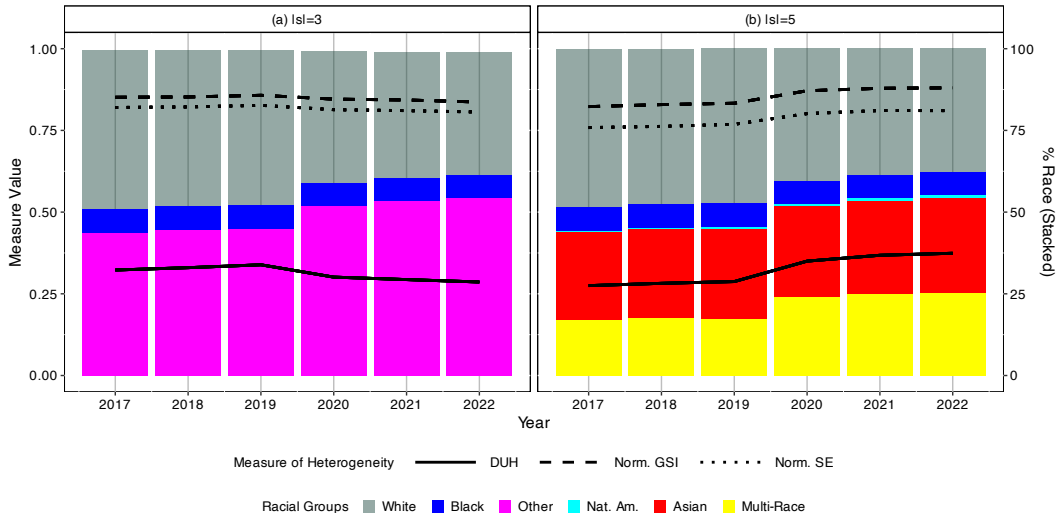


# Reasonable Group Labels

While the satisfaction of GSYM implies that exact group labels are independent of the heterogeneity of a system, the satisfaction of CON implies that two systems are only directly comparable if the labeling of elements reflect a reasonable and interpretable grouping scheme.

# Reasonable Group Labels

Figure 3: Different Group Labels Can Yield Opposite Inferences



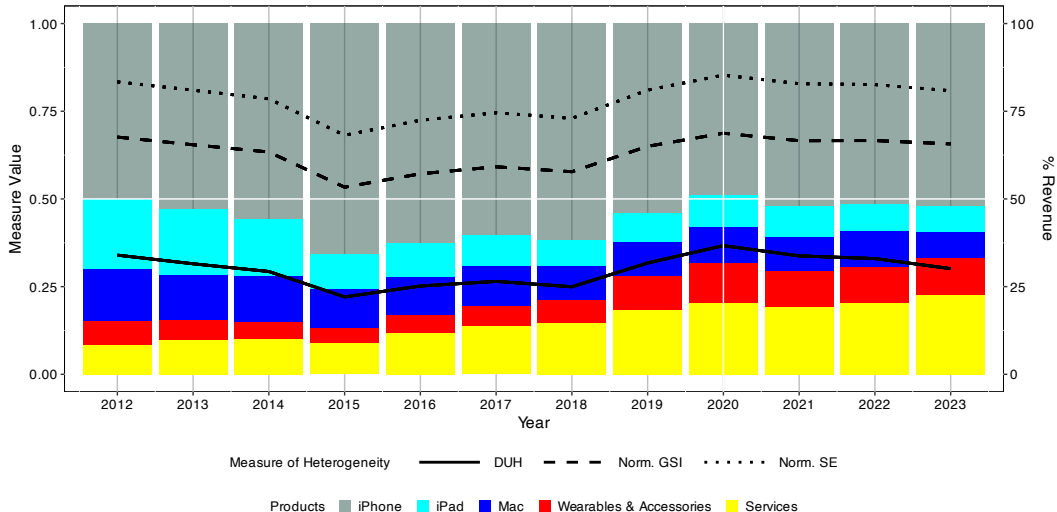
# Empirical Example: Racial Heterogeneity

**Table 1:** The Progression of Racial Composition and Racial Heterogeneity of a Hypothetical City

Share (%)	Decade			
	1	2	3	4
White	60%	65%	66%	69%
Black	34%	26%	22%	17%
Other	6%	9%	11%	14%
DUH	0.235	0.257	0.284	0.315
Gini	0.460	0.440	0.450	0.450
GSI	0.521	0.502	0.499	0.475
Norm. GSI	0.781	0.753	0.749	0.713
Norm. SE	0.767	0.771	0.778	0.758

# Empirical Example: Racial Heterogeneity

Figure 4: Revenue Heterogeneity using DUH, GSI, and SE



# Thank You!

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# Characterization of $DUH$

## Proof sketch

- 1 I show that any measure mapping an ordered system of group shares to the real numbers satisfies GSYM and INV.
- 2 I show that any  $\Phi$  that satisfies GSYM, INV, and PPT must be a positive monotonic transformation of  $\frac{1}{\hat{\sigma}_1}$ .
- 3 I show that if  $\Phi(\phi, \psi_p)$  satisfies INV, IND, and PPT, then  $\phi$  and  $\psi_p$  must be multiplicatively separable. In other words,  $\Phi = \phi \cdot \psi_p$ .
- 4 I show that, holding  $\hat{\sigma}_1$  constant and assuming GSYM, INV, and IND, the measure  $\Phi$ , using the measure of evenness  $\psi_p$  as defined, satisfies PT if and only if  $p > 1$  when  $|s| > 3$  and  $p \geq 1$  when  $|s| = 3$ .

# Characterization of $DUH$

- 5 I show that if  $\Phi(\phi, \psi_p)$  satisfies GSYM, INV, IND, and PPT, then  $\phi = -c \cdot \log_q(\hat{\sigma}_1)$ ,  $c, q \in \mathbb{R}_{++}$ .
- 6 For the case of  $p = 2$ , using the Euclidean distance for  $\psi_p$ , I show that the DUH family satisfies PT, in addition to GSYM, INV, IND, and PPT, by taking the derivative of the extreme case in which  $\hat{\sigma}_1$  is close to 1 and  $\psi_2 = 1$  with respect to a transfer from the largest group to the second-largest group.
- 7 I show that the DUH family satisfies CON and UNI, in addition to GSYM, INV, IND, PT, and PPT.

Adding any arbitrary number of zero-groups does not affect the measure  $\Phi$ .

$\Phi(n_1, \dots, n_G)$  satisfies *Expandability* if

$$\Phi(n_1, \dots, n_G) = \Phi(n_1, \dots, n_G, 0)$$



# [REP] Replication Principle

$\Phi(n_1, \dots, n_G)$  satisfies *Replication Principle* (for concentration) if  $\forall k \in \mathbb{N}$

$$\frac{1}{k} \Phi(n_1, \dots, n_G) = \Phi \left( \underbrace{\frac{n_1}{k}, \frac{n_1}{k}, \dots, \frac{n_1}{k}}_{\text{Sum to } n_1}, \frac{n_2}{k}, \frac{n_2}{k}, \dots, \underbrace{\frac{n_G}{k}, \dots, \frac{n_G}{k}}_{\text{Sum to } n_G} \right)$$

# [SADD] Shannon's Additivity

Define  $n_{gj} \geq 0$  such that  $n_g = \sum_{j=1}^{m_g} n_{gj}$ ,  $\forall g \in \{1, \dots, G\}$ ,  $\forall j \in \{1, \dots, m_g\}$

$\Phi(n_1, \dots, n_G)$  satisfies *Shannon's Additivity* if

$$\Phi(n_{11}, \dots, n_{Gm_G}) = \Phi(n_1, \dots, n_G) + \sum_{g=1}^G \frac{n_g}{n_S} \cdot \Phi\left(\frac{n_{g1}}{n_g}, \dots, \frac{n_{gm_g}}{n_g}\right)$$

which implies (by setting  $m_g = 1$ ,  $\forall g \in \{1, \dots, G-1\}$  and  $n_{G'} = n_G + n_{G+1}$ ),

$$\begin{aligned} \Phi(n_1, \dots, n_G, n_{G+1}) &= \Phi(n_1, \dots, n_{G'}) \\ &\quad + \frac{n_{G'}}{n_1 + \dots + n_{G-1} + n_{G'}} \cdot \Phi\left(\frac{n_G}{n_{G'}}, \frac{n_{G+1}}{n_{G'}}\right) \end{aligned}$$