

# Measuring inequality

LECTURE 13

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## Outline for final lectures

- Once datasets have been finalized, it is time to produce results, with the aim of representing the patterns emerging from the data.
- In practice?
  - Inequality this lecture
  - Poverty next lecture
- Basic summary statistics on household demographics, education, access to services, etc.
- Average expenditures and incomes final lecture

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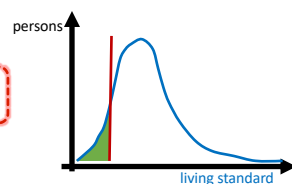
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## Inequality and poverty measurement

- 1) a measure of living standards
- 2) high-quality data on households' living standards
- 3) a distribution of living standards (inequality)
- 4) a critical level (a poverty line) below which individuals are classified as "poor"
- 5) one or more poverty measures



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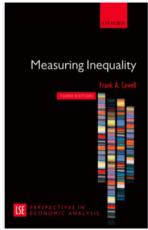
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Cowell (2011)



99.9% of this lecture is explained with better words in Cowell's work: this book and other (countless) journal articles



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## Warning

- During the course we paid attention to distinguish between different concepts: living standard, income, expenditure, consumption, etc.
- In this lecture we make an **exception** and use these terms **interchangeably** – we focus on measuring inequality of “a distribution”
- Similarly, I will **not** make a distinction between income per **household**, per **capita**, or per **adult equivalent**
- For once, and for today only, we will be (occasionally) inconsistent



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## Basic concepts

- Economists make a distinction between:
  - **Functional** distribution of income  
distribution among **factors of production**  
land (**rent**), labor (**wages**), and capital (**profits**)
  - **Personal (or size)** distribution of income  
distribution among **persons**, irrespective of their economic function
- We focus on the latter.



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## Functional vs Personal distribution of income

Average factor shares in Indian Economy, 1960-61 to 1991-1992

### Functional Distribution of National Income in India

Abstract: This study attempts to measure the functional distribution of national income in India for the period 1960-61 to 1991-92. The study is based on the data of the Central Statistical Organisation (CSO) and the National Accounts Statistics (NAS). The study shows that the share of labour in national income has increased from 56.42% in 1960-61 to 60.69% in 1991-92. The share of land has decreased from 30.30% in 1960-61 to 15.21% in 1991-92. The share of capital has increased from 13.28% in 1960-61 to 24.10% in 1991-92. The study also shows that the share of labour in national income has increased in all sectors, while the share of land has decreased in all sectors. The share of capital has increased in all sectors.

Sector	Labour	Land	Capital
Primary sector	56.42	30.30	13.28
Secondary sector	67.68	3.47	28.85
Tertiary sector	61.57	3.74	34.69
All sectors	60.69	15.21	24.10
Public sector	86.15	0.83	13.02
Private sector	56.53	17.76	25.71

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## Focus on the term 'inequality'

- “When we say **income inequality**, we mean simply **differences in income**, without regard to their desirability as a system of reward or undesirability as a scheme running counter to some ideal of equality” (Kuznets 1953: xxvii)
- In practice, how can we appraise the inequality of a given income distribution? Three main options:
  - ① Tables
  - ② Graphs
  - ③ Summary statistics

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## The distribution of income and taxable income

2018 Tax statistics, National Treasury and the South African Revenue Service




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Tax year		2017			
Income group	Number of taxpayers	Income before deductions (R million)	Deductions allowed (R million)	Taxable income (R million)	
<= 0	116 998	-15 096	11	-15 107	
1 – 70 000	401 447	14 835	565	14 270	
70 001 – 350 000	2 689 263	543 389	58 380	485 009	
350 001 – 500 000	764 197	317 965	44 483	273 483	
500 000 +	926 660	898 846	109 690	789 157	
<b>Total</b>	<b>4 898 565</b>	<b>1 759 939</b>	<b>213 128</b>	<b>1 546 811</b>	

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### Tables: an assessment

- In general, tables are **not recommended** when the focus is inequality
- Difficult to get a clue of the extent of inequality in the distribution by looking at a table, plus income brackets are arbitrary
- Does putting income distribution into a graph (diagram) **help** to represent inequality?

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### Histograms

- Let the interval  $[x^-, x^+]$  denote the **range** of the data.
- Partition  $[x^-, x^+]$  into  $m^*$  non-overlapping **bins** (intervals) of equal **width**  $h = (x^+ - x^-)/m^*$ .
- A histogram estimate of the **density**  $f(x)$  is the fraction of observations falling in the bin containing  $x$ , divided by the bin width  $h$ :

$$\hat{f}(x) = \frac{\text{(fraction of sample obs. in same bin as } x)}{h}$$

- The area of each bar ( $= h \times \hat{f}(x)$ ) is interpreted as the fraction of sample observations within the bin. All bar areas sum up to unity.

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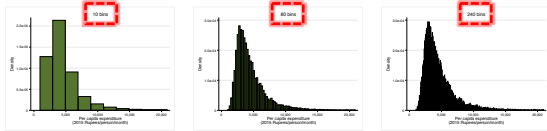
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## Histograms?



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## Histograms: an assessment

- The position and number of bins is **arbitrary**
- Inherently **lumpy**: discontinuities at the edge of each bin
- Can provide very **different pictures** of the **same distribution**
- Read Cowell, Jenkins and Litchfield (1996) for more.

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## Beyond histograms

- A **probability density function (PDF)** is the 'continuous version' of a histogram
- A convenient way to introduce the PDF is by starting from the **cumulative distribution function (CDF)**

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The cumulative distribution function (CDF)

- The **cumulative distribution function (CDF)** is defined as follows:

$$F(x) = \int_0^x f(x)dX$$

- $F(x)$  is the **proportion of individuals** having  $X$  less than or equal to  $x$ .
- If  $X$  is income and, say,  $x = 2,000$  Rps., then  $F(x) = \Pr(X < 2,000)$ , that is the fraction of people with less than 2,000 Rps.



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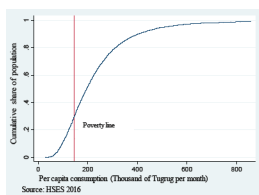
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The empirical cumulative distribution function (ecdf)

Mongolia HSES 2016, Cumulative distribution of per capita consumption (p.10)



- Pick up any income level on the **x-axis**, and the curve **F(x)** will tell you the percentage of individuals in the population having a level of income **lower than x**.



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The probability density function (pdf)

- The **probability distribution function (pdf)** is the derivative of the CDF:

$$f(x) = \frac{dF(x)}{dx}$$

- By definition of derivative:

$$f(x) = \lim_{h \rightarrow 0} \frac{F(x+h) - F(x)}{h}$$



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## The probability density function (pdf)

- Now drop the limit (and replace = by  $\approx$ ):

$$f(x) \approx \frac{F(x+h) - F(x)}{h}$$

$$h \cdot f(x) \approx F(x+h) - F(x)$$

$$h \cdot f(x) \approx \Pr(X \leq x+h) - \Pr(X \leq x) \approx \Pr(x \leq X \leq x+h)$$

- The PDF  $f(x)$  is not a probability measure, but a scaled version of it: it is the probability of  $X$  falling in the interval  $(x, x+h)$  divided by the length  $h$  of such an interval.

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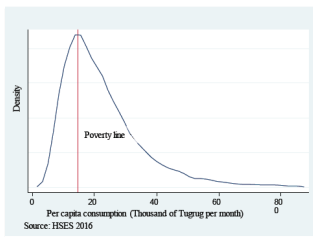
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## Mongolia, 2016

PDF of per capita consumption (p.11)




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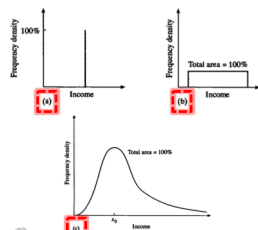
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## Interpretation



Two extremes:

- perfect equality: everyone is concentrated at one particular income value
- uniform density: income is spread uniformly from the poorest to the richest individual – significant inequality
- in-between, typical case.

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Pdfs: an assessment

- The bandwidth is **arbitrary**
- In most cases, they require some **trimming** of top values to avoid looking "squished" and being unreadable
- In general, it does not show what is going on in the **upper tail** very clearly



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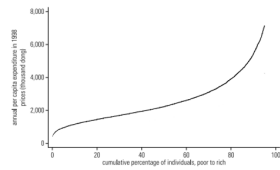
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The quantile function

- Let  $p = F(x)$  be the proportion of people in the population with income **lower than**  $x$ .
- The **quantile function**  $Q(p)$  is defined as:  
 $F[Q(p)] = p$  or  $Q(p) = F^{-1}(p)$   
cdf                      inverse cdf
- $Q(p)$  is the **income level** below which we find a proportion  $p$  of the population.



Source: Imbens and Wainer (2005).  
See [http://www.stat.columbia.edu/ksimon/papers/05\\_01\\_05.pdf](http://www.stat.columbia.edu/ksimon/papers/05_01_05.pdf)



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\*The Parade of Dwarfs

Pen (1971)

- Assume that everyone in the population has **height** proportional to **income**.
- Line people up in order of height, and let them **march**.
- After some time, the shape of such a parade will be represented by the curve called **Parade of Dwarfs** (and a Few Giants).



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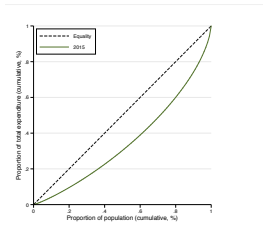
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## The Lorenz Curve (1905)

Picture and intuition



- **Horizontal axis:** cumulative % of **population** (individuals ordered poorest to the richest)
- **Vertical axis:** cumulative % of **income** received by each cumulative % of population.
- **45-degree line:** Lorenz curve if perfect equality.
- The overall distance between the 45-degree line and the Lorenz curve is indicative of the amount of **inequality** present in the population.

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## The Lorenz Curve (1905)

Mathematically

- The **Lorenz curve**  $L(p)$  is defined as follows:

$$L(p) = \frac{\int_0^p Q(q) dq}{\int_0^1 Q(q) dq}$$

- The **numerator** sums the incomes of the poorest  $p\%$  of the population;
- The **denominator** sums the incomes of all.
- The **ratio**  $L(p)$  indicates the cumulative % of total income held by a cumulative proportion  $p$  of the population.
- **Example:** if  $L(0.5) = 0.3$ , then we know that the **50%** poorest individuals hold **30%** of the total income in the population.

C4D2 TRAINING

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## Quantile function and Lorenz curve: an assessment

- These graphical tools emphasize the **ranking** of shares of the population on the basis of income
- The Lorenz curve clearly shows how far the distribution is from perfect equality
- Still, no graph is as straightforward and easily comparable as a **scalar measure** of inequality

C4D2 TRAINING

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## Recap and next steps

- Not all **graphs** are OK to represent inequality
- **Lorenz curve** is the most popular
- A better conceptual understanding comes from constructing inequality **measures** from first principles.
- The most straightforward approach: inequality measures as **pure statistical measures of dispersion**.

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## Inequality indicators

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## Measures of dispersion

- **Range**  $R = x_{\max} - x_{\min}$ 
  - ▲ **PRO:** Easy to compute and communicate
  - ▼ **CON:** Insensitive to changes between extremes (can we really know min and max?)
- **Variance**  $\sigma^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2$ 
  - ▲ **PRO:** Easy to compute, additively decomposable
  - ▼ **CON:** not robust (outliers), depends on the scale of measurement
- **Coefficient of Variation**  $CV = \frac{\sqrt{\sigma^2}}{\mu}$ 
  - ▲ **PRO:** Scale invariant
  - ▼ **CON:** not robust (outliers), properties?

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## Quantiles, Quintiles, Quartiles, ...

- The **p-quantile** of a distribution of values is a number  $x_p$  such that a proportion  $p$  of the population values are less than or equal to  $x_p$ .
- For example, if  $p = 0.5$ , then the 0.5-quantile  $x_{0.5}$  is any value such that  $F(X < x_{0.5}) = 0.5$ .
- Certain quantiles have special names:
  - The 0.5-quantile  $x_{0.5}$  is the **median**, or **50-th percentile**.
  - The 0.1-quantile is the **first decile**, or **10-th percentile**.
  - The 0.2-quantile is the **first quintile**, or **20-th percentile**.
  - The 0.25-quantile is the **first quartile  $Q_1$** , or **25-th percentile**.
  - etc. etc.

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## Quantile ratios

- A **quantile ratio** measures the **gap** between the rich and the poor.
- It is defined as the ratio of two quantiles  $Q(p_2)/Q(p_1)$  using percentiles  $p_1$  and  $p_2$ .
- Three popular indices are:
  - The **quintile ratio** ( $p_2 = 80$  and  $p_1 = 20$ ):  
 $QR = Q(p_{80})/Q(p_{20})$
  - the **decile ratio** ( $p_2 = 90$  and  $p_1 = 10$ ):  
 $DR = Q(p_{90})/Q(p_{10})$

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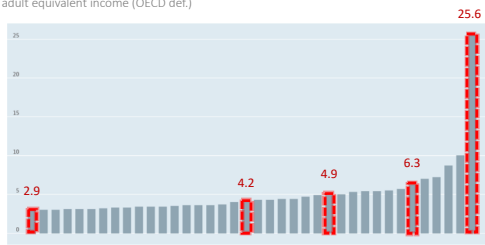
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## The decile ratio

per adult equivalent income (OECD def)




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### Quantile share ratios

- Let  $S_{20}$  denote the share of (equivalised disposable) income received by the **bottom 20%** of the population, and  $S_{80}$  the income share received by the **top 20%** of the population.
- The **quintile share ratio** is defined as follows:  

$$S_{80-20} = S_{80}/S_{20}$$
- The quintile share ratio is the level-1 Laeken indicator, chosen by the EU to monitor income distribution.




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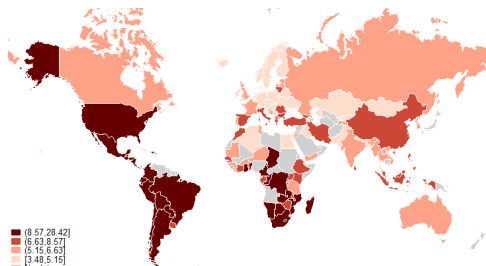
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### QSR around the world



Source: 1993 income share held by highest 20% over income share held by lowest 20%, last available 2002-2017 35

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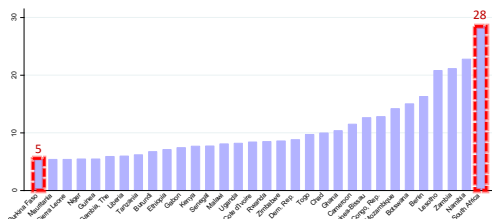
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### $S_{80}/S_{20}$ in Sub Saharan Africa



Source: 1993 income share held by highest 20% over income share held by lowest 20%, last available 2002-2017 36

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### The Gini Coefficient

A definition

- Yitzhaki (1997) counts more than a dozen formulas available for the Gini index.
- A classic definition of the Gini coefficient:

$$G = \frac{1}{2n^2\mu} \sum_{i=1}^n \sum_{j=1}^n |x_i - x_j|$$

- The Gini coefficient ranges from 0 (all recipients have the same income: full equality), to 100 (all income is received by one recipient: maximum inequality).




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### The Gini Coefficient

Interpretation – Pyatt 1976: 244

- The Gini index “is the average gain to be expected, if each individual has the choice of being himself or some other member of the population drawn at random, expressed as a proportion of the average level of income”
- E.g., if the Gini index for an Italian is 0.30, we can say that the expected gain from playing the experiment of exchanging income with someone else randomly chosen in the Italian population, is 30% of average income.




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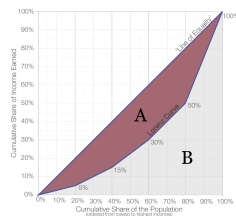
### The Gini Coefficient

graphical interpretation

- The Gini index is two times the area A between the Lorenz curve and the equality diagonal:

$$\begin{aligned} Gini &= \frac{A}{(A+B)} \\ &= 2A \\ &= 2\left(\frac{1}{2} - B\right) = 1 - 2B \end{aligned}$$

$$\begin{aligned} Gini &= 2 \int_0^1 [p - L(p)] dp \\ &= 1 - 2 \int_0^1 L(p) dp \end{aligned}$$




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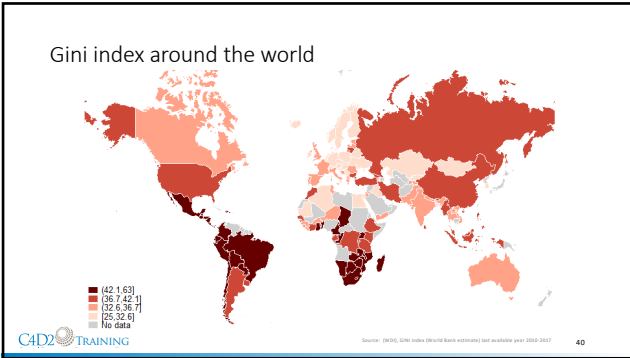
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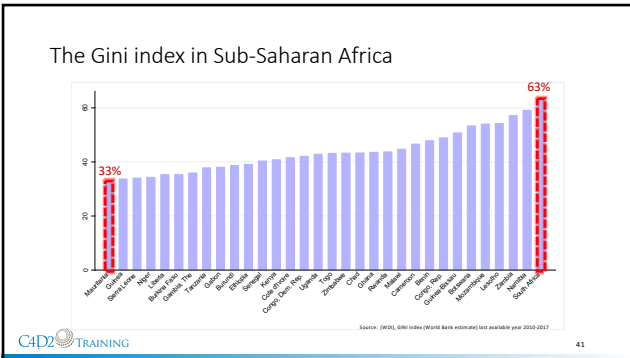
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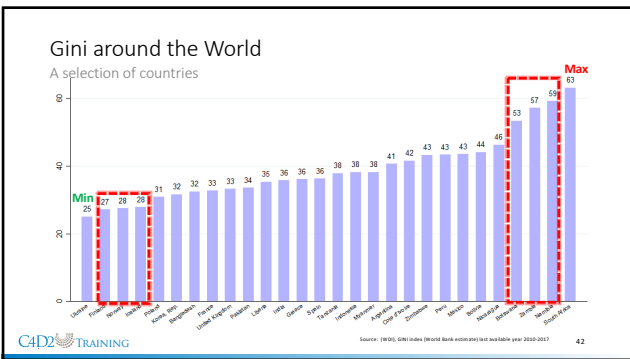
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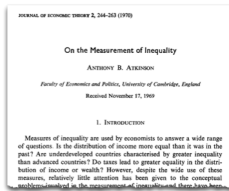
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## \*Atkinson's paper

### The paper



### Tony Atkinson (1944-2017)



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## Recap

- Quantile ratios, quantile share ratios, Gini, are all popular inequality measures
- They do a fine job at representing inequality with a number
- **Problem**  
they do not always have all the **properties** that we would want for an inequality measure
- **Solution**  
solve the problem backwards. First lay out some desirable properties, then construct a measure that complies with them

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## Deriving inequality measures from axioms

- **Axiom**: a statement accepted as true as the basis for argument or inference.
- The **axiomatic approach** allows us to “custom-build” inequality measures that fit our needs:
  1. We define a set of elementary properties (axioms) that we think inequality measures ought to have
  2. We obtain a mathematical formula that delivers a class of inequality measures satisfying the axioms

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Five axioms of inequality measures

- A. **Anonymity (or Symmetry)**  
*Who* is earning the income does not matter
- B. **The Population Principle**  
*Population size* does not matter
- C. **Scale Invariance (or Relative Income Principle)**  
*Income levels* do not matter
- D. **The (Pigou-Dalton) Principle of Transfers**  
Rank-preserving rich-to-poor transfers reduce inequality
- E. **Decomposability (or Subgroup Consistency)**  
The measure is additively decomposable

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\*Five axioms of inequality measures

(A) **Anonymity (or Symmetry)**

- If income distribution X is any permutation of income distribution Y, then  $I(X) = I(Y)$ .
- In short, it does not matter who is earning the income.

(P) **The Population Principle**

- When one income distribution is an n-fold replication of another, the two are distributionally equivalent.
- The population size does not matter: all that matters are the proportions of the population who earn different levels of income.

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\*Five axioms of inequality measures

(S) **Scale Invariance (or Relative Income Principle)**

- If everyone's income changes by the same proportion, then inequality does not change.
  - $X = (x_1, x_2, \dots, x_n)$
  - $Y = (\lambda x_1, \lambda x_2, \dots, \lambda x_n)$
  - $I(X) = I(Y)$
- Inequality should not depend on whether income is measured in PKR or €.
- *Income levels*, in and of themselves, have no meaning as far as inequality measurement is concerned.

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\*Five axioms of inequality measures

(T) The (Pigou-Dalton) Principle of Transfers

- If one distribution is obtained from another by transferring a positive amount of income  $\delta$  from a relatively rich person to a relatively poor person, without altering their ranks in the distribution, then inequality must decrease.
- $X = (x_1, x_2, \dots, x_i, \dots, x_j, \dots, x_n)$
- $Y = (x_1, x_i + \delta, \dots, x_j - \delta, \dots, x_n)$ , with  $\delta > 0$
- $I(Y) \leq I(X)$




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\*Five axioms of inequality measures

(D) Decomposability (or Subgroup Consistency)

- An **additively decomposable** inequality measure is one which can be expressed as a **weighted sum** of the inequality values calculated for population groups plus the contribution of differences between group means.

$$I = \sum_{k=1}^K \omega_k I_k + I(\bar{x}_1, \dots, \bar{x}_K), \quad \sum_{k=1}^K \omega_k = 1$$

where  $I_k$  is the inequality index calculated within the k-th group, and  $\omega_k$  are the population shares.




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Generalized Entropy Indices (GEI)

Shorrocks (1980)

- Inequality measures that satisfy all axioms (A to E), **must** have the following form:

$$GE(\theta) = \frac{1}{\theta^2 - \theta} \left[ \frac{1}{n} \sum_{i=1}^n \left( \frac{x_i}{\bar{x}} \right)^\theta - 1 \right]$$

where  $\theta$  is a parameter that may be given any value (positive, zero or negative).




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## The Generalized Entropy Indices

- Depending on the value of the  $\theta$  parameter:

$\theta = 0 \rightarrow$  Mean Logarithmic Deviation

$$GE(0) = MLD = \frac{1}{n} \sum_{i=1}^n \log\left(\frac{x_i}{\bar{x}}\right)$$

$\theta = 1 \rightarrow$  Theil Index

$$GE(1) = THEIL = \frac{1}{n} \sum_{i=1}^n \frac{x_i}{\bar{x}} \log\left(\frac{x_i}{\bar{x}}\right)$$

$\theta = 2 \rightarrow$  Half Coefficient of Variation Squared

$$GE(2) = \frac{\sigma^2}{2}$$

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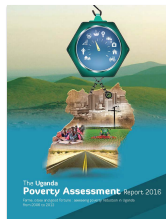
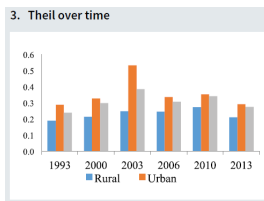
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## Uganda

National Household Survey 2012/2013




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## Inequality decomposition

- Inequality decompositions are typically used to estimate the extent to which the **heterogeneity** of the population affects overall inequality. Two popular techniques are:

- Decomposition by **population sub-group**
- Decomposition by **income source**

- We focus on the former:

- Societies can often be partitioned into groups (e.g. North-South). We would like to be able to decompose total inequality into two components, namely the inequality **within** the constituent groups, and inequality **between** the groups:

$$I_{TOTAL} = I_{WITHIN} + I_{BETWEEN}$$

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## Inequality decomposition

- The most popular additively decomposable inequality index is the **Mean Logarithmic Deviation**.
- Partition the population into  $k = 1, \dots, K$  groups. Then:

$$MLD = \underbrace{\sum_{k=1}^K v_k MLD_k}_{\text{WITHIN}} + \underbrace{\sum_{k=1}^K v_k \log\left(\frac{x}{x_k}\right)}_{\text{BETWEEN}}$$

where  $v_k$  are population shares.

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## Botswana, 2009/10

household income and expenditure survey

	2010		
	GE(0)	GE(1)	GE(2)
Total	0.669	0.823	3.206
<b>Urban / rural</b>			
Between-group inequality	0.063	0.068	0.077
Between as a share of total	0.094	0.083	0.024
Within-group inequality	0.606	0.754	3.130
<b>Region</b>			
Between-group inequality	0.060	0.068	0.081
Between as a share of total	0.090	0.083	0.025
Within-group inequality	0.609	0.754	3.125

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## Lessons learned

- Many ways to **describe** inequality, some more effective than others
- **Graphs**: most notable are quantile functions and Lorenz curves
- **Measures**: different inequality measures lead to different results. Based on their properties, the recommended choice is GEI (generalized entropy indices), and in particular the MLD (mean log deviation). However, Gini remains extremely popular in practice

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## References

### Required readings

Cowell, F. (2011). *Measuring inequality*. Oxford University Press. (Chapter 1 & 2)

### Suggested readings

Atkinson, A. B. (1970). On the measurement of inequality. *Journal of economic theory*, 2(3), 244-263.

Cowell, F., Jenkins, S. P., and Litchfield, J. A. (1996). *The changing shape of the UK income distribution: kernel density estimates* (pp. 49-75). Cambridge University Press.

Farris, F. A. (2010). The Gini index and measures of inequality. *The American Mathematical Monthly*, 117(10), 851-864.

Haughton and Khandker (2009). *Handbook on Poverty and Inequality*, Chapter 6.

Pen, J. (1971). *Income Distribution: facts, theories, policies*. Praeger.

Pyatt, G. (1976). On the interpretation and disaggregation of Gini coefficients. *The Economic Journal*, 86(342), 243-255.

Shorrocks, A. F. (1980). The class of additively decomposable inequality measures. *Econometrica: Journal of the Econometric Society*, 613-625.

Thank you for your attention

Homework

### Exercise 1 – Engaging with the literature



Considering equations (10) to (12) in Farris (2010) give a brief interpretation of a Gini index of 63% for South Africa



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### Exercise 1 - Engaging with the literature

A solution

- The Gini index shows “how the lower of two randomly chosen incomes compares, on average, to mean income”.
- E.g., if the Gini index for South African family income is 0.63, “we conclude that the lower of two South African family incomes, chosen at random, is about 37%  $[(1-0.47)*100]$  of the mean; on the average, the poorer of two families earns only over one third of the national mean”.



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### Exercise 2 - Inequality in South Asia



- Turn to page 2 of this report (see next slide)
- What criticisms would you make to this chart?



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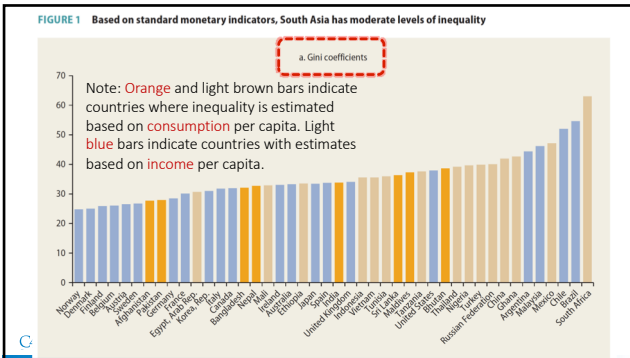
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### Exercise 3 - Functional vs Personal distribution of income

- The nature of the relationship that links the evolution of income shares to income inequality is complex and still widely debated among researchers.
- In that context, comment on Figure 19 of the ILO Global Wage Report 2016/2017.

Global Wage Report 2016/17  
Wage inequality on the workplace

**Global Wage Report**

C4D2 TRAINING 65

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