

Fundamentals of Mobility Measurement

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Outline

Background

- Basics

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- General considerations

- Status

- Measures: Intuitive and others

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- Basic structure

- Consistency

- Classes of measures

- Data issues

- Mobility indices: two classes

- Decomposition

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- desirable objective for social and economic policy?
- a policy tool?
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- improving data on intra- and inter-generational mobility
- convincing evidence needs appropriate measurement tools
- What is known about mobility?

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What do people know about mobility?

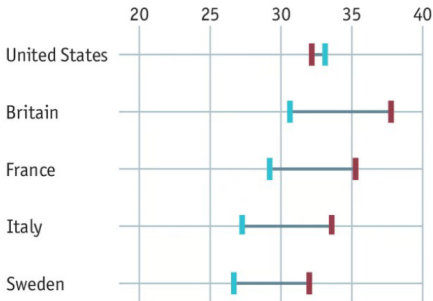
- do they value mobility?
- do they know it when they see it?

Perceptions and reality

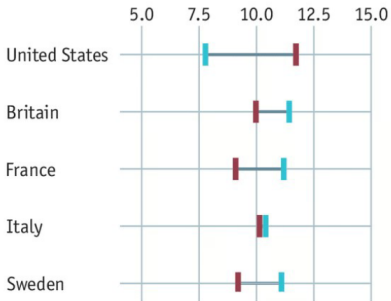
Actual and perceived social mobility of children, 2016

Actual Perceived*

Probability of remaining in the bottom quintile of earnings, %



Probability of moving from bottom to top quintile of earnings, %



Source: Alesina et al. (2018)

Academic approaches in the literature

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- **Variety of interpretation:** (Fields and Ok 1999a; Jäntti and Jenkins 2015)
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 2. long term / volatility

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- **Variety of temporal context:**
 1. inter / intra-generational
 2. long term / volatility
- **Variety of analytical context:**
 - in relation to a specific dynamic model
 - in relation to social-welfare issues
 - as an abstract distributional concept

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The approach

- Appropriate tools?
 - what makes a measure “suitable”?
 - base on simple principles concerning mobility
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- An abstract distributional concept
 - independent of value systems
 - application separated from principles
 - subject to practical limitations

The approach

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 - base on simple principles concerning mobility
 - several commonly-used techniques do not conform well
- An abstract distributional concept
 - independent of value systems
 - application separated from principles
 - subject to practical limitations
- This presentation
 - develops ideas in Cowell and Flachaire (2017, 2018)
 - shows how to give meaning of mobility comparisons

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Deal with mobility in the abstract

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- also “rank” mobility where underlying data are categorical
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Ingredients for a theory of mobility measurement:

1. a time frame
2. measure of individual status within society
3. aggregation of changes in status over the time frame

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Ingredient 1:

- Assume discrete time
- Focus on two periods: now (0) and the future (1)

Steps

- Separate the ingredients of problem
 1. time frame (two periods)
 2. status
 3. aggregation method

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- Set out general principles
 - essential
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 - check standard mobility measures against these
- Characterise an ordering
 - formulate principles as axioms
 - develop characterisation results

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Status: classes

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First step in an approach to “status”:

- define a finite set of K classes
- $n_k \geq 0$: # in class k , $k = 1, 2, \dots, K$
- exclusive and exhaustive
- $\sum_{k=1}^K n_k = n$, the size of the population

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- class k associated with attribute level x_k
($x_k < x_{k+1}$, $k = 1, 2, \dots, K - 1$)
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$k_0(i)$, $k_1(i)$: class occupied by person i at times t^0 and t^1

- mobility given by $(x_{k_0(1)}, \dots, x_{k_0(n)})$ and $(x_{k_1(1)}, \dots, x_{k_1(n)})$

Status: valuation

How to use the attribute movements to compute mobility?

- cardinal attribute: just aggregate the x s?
- don't have to use natural cardinalisation to value the x s
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- Revaluing the income classes: $N_1(x_k) := \sum_{h=1}^k n_{1h}$, $k = 1, \dots, K$

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- *static (2)*. $z_i = (\varphi(x_{k_0(i)}), \varphi(x_{k_1(i)}))$
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Exchange and structural mobility: (Van Kerm 2004, Tsui 2009)

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Statistical measures

- Many empirical studies use off-the-shelf tools
 - let income be y
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- Two widely used “statistical” methods:
 1. elasticity coefficient
 - linear regression of status-1 on status-0
 - $x_{1i} = \alpha + \beta x_{0i} + \varepsilon_i$
 - $1 - \hat{\beta}$ as a measure of mobility?

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 - $x_{1i} = \alpha + \beta x_{0i} + \varepsilon_i$
 - $1 - \hat{\beta}$ as a measure of mobility?
 2. correlation coefficient
 - use Pearson correlation coefficient $\hat{\rho}$
 - $1 - \hat{\rho}$ as a measure of mobility?

Statistical measures: elasticity coefficient

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Statistical measures: elasticity coefficient

- A high value of $1 - \hat{\beta}$ evidence of significant mobility?
- Low value does not necessarily imply low mobility
 - can have $1 - \hat{\beta} = 0$ where there is indeed mobility
 - since $\hat{\beta} = \frac{\text{cov}(\mathbf{x}_0, \mathbf{x}_1)}{\text{var}(\mathbf{x}_0)}$: $1 - \hat{\beta} = 0 \Leftrightarrow \text{cov}(\mathbf{x}_0, \mathbf{x}_1) = \text{var}(\mathbf{x}_0)$.

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- A difficulty:
 - take $\mathbf{x}_0 = (x_{01}, x_{01} + k, x_{01} + 2k)$, $\mathbf{x}_1 = (x_{11}, x_{12}, x_{11} + 2k)$
 - we have $1 - \hat{\beta} = 0, \forall x_{01}, x_{11}, x_{12}$
- Example:
 - $\mathbf{x}_0 = (1, 2, 3)$
 - $\mathbf{x}_1 \in \{(2, 0, 4), (2, 1, 4), (2, 1760, 4), (2100, 1, 2102), \dots\}$
 - zero mobility *in all cases*?

Statistical measures: correlation coefficient

- Both scale and translation independent:
 - if $x_1 = ax_0 + b$, then $\hat{\rho} = 1 \Leftrightarrow 1 - \hat{\rho} = 0$
 - so $\mathbf{x}_0 = (1, 2, 3)$ and $\mathbf{x}_1 = (0, 2, 4)$ imply $x_1 = 2x_0 - 2$; $1 - \hat{\rho} = 0$
 - Is this attractive?

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- Measure can behave strangely:
 - take equidistant status
 - $\mathbf{x}_0 = (x_{01}, x_{01} + k, x_{01} + 2k)$, $\mathbf{x}_1 = (x_{11}, x_{12}, x_{11})$
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- Example
 - $\mathbf{x}_0 = (1, 2, 3)$
 - $\mathbf{x}_1 \in \{(3, 2, 3), (3, 0, 3), (3, 100, 3), (1, 2, 1), (10, 1, 10), (2, 1, 2), \dots\}$
 - in all cases $1 - \hat{\rho} = 1$ and $1 - \hat{\beta} = 1$

Inequality-based measures

- Fields and Ok (1996) measure based on income differences:
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- Fields and Ok (1999b) measure based on log-income differences:
 - $FO_2 = \frac{1}{n} \sum_{i=1} |\log y_{1i} - \log y_{0i}|$

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 - $FO_1 = \frac{1}{n} \sum_{i=1} |y_{0i} - y_{1i}|$
- Fields and Ok (1999b) measure based on log-income differences:
 - $FO_2 = \frac{1}{n} \sum_{i=1} |\log y_{1i} - \log y_{0i}|$
- Shorrocks (1978) measures related to inequality:
 - $S_I = 1 - \frac{I(y_0+y_1)}{\frac{\mu_{y_0}}{\mu_{y_0+y_1}} I(y_0) + \frac{\mu_{y_1}}{\mu_{y_0+y_1}} I(y_1)}$
 - where $I(\cdot)$ is a predefined inequality measure

Directional measures

- Ray and Genicot (2023) upward mobility index (absolute) :
 - $RG_1 = -\frac{1}{\alpha} \log \left(\frac{\sum_{i=1}^n y_{1i}^{-\alpha}}{\sum_{i=1}^n y_{0i}^{-\alpha}} \right), \alpha > 0$

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- Ray and Genicot (2023) upward mobility index (relative):
 - $RG_2 = -\frac{1}{\alpha} \log \left(\frac{\sum_{i=1}^n y_{1i}^{-\alpha}}{\sum_{i=1}^n y_{0i}^{-\alpha}} \right) + \log \left(\frac{\sum_{i=1}^n y_{0i}}{\sum_{i=1}^n y_{1i}} \right)$

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- Bárcena and Cantó (2025) downward mobility index
 - $BC_D = \frac{1}{n} \sum_{i \in D} \left(\frac{y_{0i} - y_{1i}}{y_{0i}} \right)^\alpha, \quad \alpha \geq 0$

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 - $RG_1 = -\frac{1}{\alpha} \log \left(\frac{\sum_{i=1}^n y_{1i}^{-\alpha}}{\sum_{i=1}^n y_{0i}^{-\alpha}} \right), \alpha > 0$
- Ray and Genicot (2023) upward mobility index (relative):
 - $RG_2 = -\frac{1}{\alpha} \log \left(\frac{\sum_{i=1}^n y_{1i}^{-\alpha}}{\sum_{i=1}^n y_{0i}^{-\alpha}} \right) + \log \left(\frac{\sum_{i=1}^n y_{0i}}{\sum_{i=1}^n y_{1i}} \right)$
- Bárcena and Cantó (2025) downward mobility index
 - $BC_D = \frac{1}{n} \sum_{i \in D} \left(\frac{y_{0i} - y_{1i}}{y_{0i}} \right)^\alpha, \quad \alpha \geq 0$
- Bárcena and Cantó 2018 upward mobility index
 - $BC_U = \frac{1}{n} \sum_{i \in U} \left(\frac{y_{1i} - y_{0i}}{y_{0i}} \right)^\alpha, \quad \alpha \geq 0$

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Comparative performance: 3-person society

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	period			period				
	0	1 ^a	1 ^b	1 ^c	1 ^d	1 ^e	1 ^f	1 ^g
A	10	20	15	20	40	25	10	10
B	20	40	25	40	80	45	30	40
C	40	80	45	10	20	15	40	160

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	0	1 ^a	1 ^b	1 ^c	1 ^d	1 ^e	1 ^f	1 ^g
A	10	20	15	20	40	25	10	10
B	20	40	25	40	80	45	30	40
C	40	80	45	10	20	15	40	160
Elasticity	$1 - \hat{\beta}$	0	0.208	1.500	1.500	1.368	0	-1.000
Pearson correlation	$1 - \hat{\rho}$	0	0.001	1.500	1.500	1.465	0.053	0
Fields-Ok 1	FO_1	23.333	5.000	20.000	36.667	21.667	3.333	46.667
Fields-Ok 2	FO_2	0.693	0.249	0.924	1.155	0.903	0.135	0.693
Shorrocks 1	S_{Theil}	0	0.011	0.736	0.680	0.739	0.034	0.053
Shorrocks 2	S_{Gini}	0	0	0.500	0.444	0.500	0	0
Ray-Genicot absolute	RG_1	0.693	0.306	0	0.693	0.306	0.100	0.288
Ray-Genicot relative	RG_2	0	0.112	0	0	0.112	-0.033	-0.811
Bárcena-Cantó downward	BC_D	0	0	0.250	0.167	0.208	0	0
Bárcena-Cantó upward	BC_U	1.000	0.292	0.667	2.000	0.917	0.167	1.333

Comparative performance: China

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- Did it rise or fall around the millennium?
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	<i>1989-2000</i>	<i>2000-2011</i>
$1 - \beta$	0.7564	0.6928
$1 - \rho$	0.7947	0.7257
FO ₁	6506.5	16979.62
FO ₂	0.9619	1.1726

Principles: movement

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 1. mobility and unbalanced growth: (Bourguignon 2011)
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- Each captures a different concept of mobility :
 1. mobility and unbalanced growth: (Bourguignon 2011)
 2. inequality change and exchange mobility (Jäntti and Jenkins 2015; Kessler and Greenberg 1981, McClendon 1977)
- Essential for mobility measurement?
 - ensures a minimum-mobility property
 - situation with some movement registers higher mobility than a situation without movement

Principles: decomposition

- Applied to other aspects of distributional analysis
 - inequality
 - welfare evaluation

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 - inequality
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- Several aspects of decomposability may be desirable
 - decomposition by population characteristics
 - decomposition by region
- Special for mobility:
 - decompose by direction
 - mobility in terms of upward and downward movements (Bárcena and Cantó 2018; Bárcena and Cantó 2025)

Principles: consistency

- Consistency in comparisons:
 - comparing one bivariate distribution of (status-in-0, status-in-1) with another

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 - rescaling all the status values by a common factor?
 - translating the distributions by the same given amount?

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 - translating the distributions by the same given amount?
- Under such circumstances should each pair of distributions be ranked the same?

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Axiomatic approach

Axiomatic approach

- Basic concepts
 - status
 - individual observation
 - derived from distribution
 - Individual i 's status history $z_i = (u_i, v_i)$
 - profile: a list of histories $\mathbf{z} = (z_1, z_2, \dots, z_n)$

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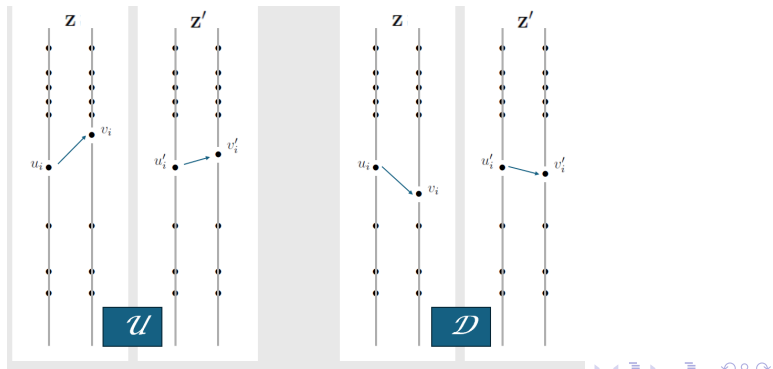
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- Use a priori axiomatisation
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 - gives a class of indices (Cowell and Flachaire 2017, 2018)
- Key axioms:
 - correspond to main principles
 - movement, decomposition consistency
 - do this in two stages

Monotonicity

[Monotonicity] Let $\mathbf{z}, \mathbf{z}' \in Z^n$ differ only in their i th history and $u'_i = u_i$ and define two conditions $\mathcal{U} := "v_i > v'_i \geq u_i"$ and $\mathcal{D} := "v_i < v'_i \leq u_i"$. If z_i satisfies either \mathcal{U} or \mathcal{D} then $\mathbf{z} \succ \mathbf{z}'$

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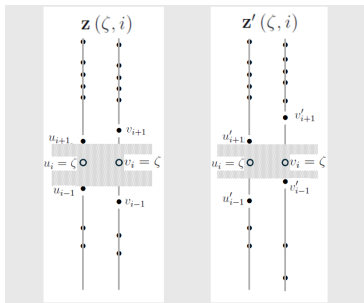


Independence

[Independence] Consider two profiles $\mathbf{z}, \mathbf{z}' \in Z^n$ where there is some $i \in \{2, \dots, n-1\}$ such that $u_{i-1} < u_i < u_{i+1}$, $v_{i-1} < v_i < v_{i+1}$, $u'_{i-1} < u_i < u'_{i+1}$, $v'_{i-1} < v_i < v'_{i+1}$. Let $\mathbf{z}(\zeta, i)$ denote the profile formed by replacing the i th history in \mathbf{z} by the history $\zeta \in Z$ and let $\hat{Z}_i := [u_{i-1}, u_{i+1}] \times [v_{i-1}, v_{i+1}]$. If $\mathbf{z} \sim \mathbf{z}'$ and $z_i = z'_i$ then $\mathbf{z}(\zeta, i) \sim \mathbf{z}'(\zeta, i)$ for all $\zeta \in \hat{Z}_i \cap \hat{Z}'_i$.

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A basic result

- **Monotonicity**] If $\mathbf{z}, \mathbf{z}' \in Z^n$ differ only in their i th history and $u'_i = u_i$ then, if $v_i > v'_i \geq u_i$, or if $v_i < v'_i \leq u_i$, $\mathbf{z} \succ \mathbf{z}'$
- **Independence**] Let $\mathbf{z}(\zeta, i)$ be profile found by replacing z_i by ζ and let $\hat{Z}_i := [u_{(i-1)}, u_{(i+1)}] \times [v_{(i-1)}, v_{(i+1)}]$. If $\mathbf{z} \sim \mathbf{z}'$ and $z_i = z'_i$ for some $i \in 2, \dots, n-1$ then $\mathbf{z}(\zeta, i) \sim \mathbf{z}'(\zeta, i)$ for all $\zeta \in \hat{Z}_i$

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- **[Continuity]** \succeq is continuous on Z^n
- **[Local immobility]** Let $\mathbf{z}, \mathbf{z}' \in Z^n$ where for some i , $u_i = v_i$, $v'_i = u'_i$ and, for all $j \neq i$, $u'_j = u_j$, $v'_j = v_j$. Then $\mathbf{z} \sim \mathbf{z}'$

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Theorem 1: given these axioms then $\forall \mathbf{z} \in Z^n$ the mobility ordering \succeq is an increasing monotonic transform of $\sum_{i=1}^n \phi_i(z_i)$

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Scale invariance (SI)

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$$\phi_i(u, v) = \begin{cases} A_i(v) [u^\alpha - v^\alpha] & \text{if OSI holds,} \\ A'_i(u) [v^\alpha - u^\alpha] & \text{if DSI holds,} \\ v^\beta h_i\left(\frac{u}{v}\right) & \text{if PSI holds,} \end{cases}$$

where A_i, A'_i, h are functions of one variable and α, β are constants.

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- Provides the basis for a class of mobility measures

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- Combining PTI with OTI or DTI :

$$\begin{aligned} & \text{either } \phi_i(u, v) = a_i[v - u], \\ & \text{or } \phi_i(u, v) = a_i[e^{\beta[v-u]} - 1] \end{aligned}$$

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- Then $\phi_i(u_i, v_i)$ must take the form

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- The term a_i can do a lot of work!

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- Normalise: $\beta = 1$, $c' = \frac{1}{\alpha[\alpha-1]}$, impose mobility property:
- Mobility is given by $\psi \left(\frac{1}{n} \sum_{i=1}^n u_i^\alpha v_i^{1-\alpha} - \theta(\mu_u, \mu_v), \mu_u, \mu_v \right)$
 - $\mu_u := \frac{1}{n} \sum_{i=1}^n u_i$, $\mu_v := \frac{1}{n} \sum_{i=1}^n v_i$, $\theta(\mu, \mu) = \mu$

Implications: Class 2

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Implications: Class 2

- Evaluation function: $\phi_i(u, v) = a_i[v - u]$
 - for any history $z_i = (v_i, u_i)$
 - write the (signed) distance $d_i := v_i - u_i$
- Overall mobility index $\sum_{i=1}^n a_i d_{(i)}$
 - $d_{(i)}$ is the i th component of vector (d_1, \dots, d_n) in ascending order
 - $d_{(1)} < 0$ is greatest downward mobility
 - $d_{(n)} > 0$ is greatest upward mobility
- Monotonicity?
 - $a_i < 0$ whenever $d_{(i)} < 0$
 - $a_i > 0$ whenever $d_{(i)} > 0$
- Population
 - a_i should be proportional to $1/n$
 - up to a change in scale we have $\frac{1}{n} \sum_{i=1}^n a_i d_{(i)}$

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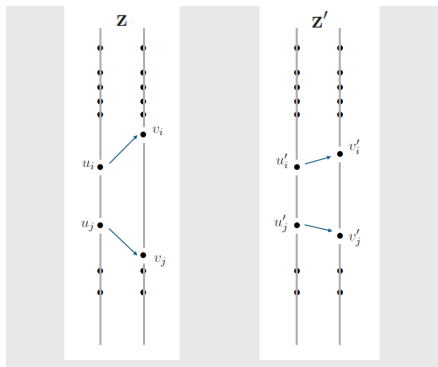
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- Status concept extends to ordinal data (Cowell and Flachaire 2017)
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- Now consider how to handle Interpretation 2 of “movement”

Mean-normalised data (1)

[Monotonicity-2] If \mathbf{z}, \mathbf{z}' differ only in components i, j : $u'_i = u_i$, $u'_j = u_j$, $v'_i - v_i = v_j - v'_j$; then, if $v_i > v'_i \geq u_i$ and if $v_j < v'_j \leq u_j$, $\mathbf{z} \succ \mathbf{z}'$

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- Mean-normalised version
 - divide each u_i by μ_u and each v_i by μ_v

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Mobility indices: Class 1

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- We have a *class* of aggregate mobility measures

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$$M_0 = -\frac{1}{n} \sum_{i=1}^n \frac{v_i}{\mu_v} \log \left(\frac{u_i}{\mu_u} / \frac{v_i}{\mu_v} \right)$$

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 - high $\alpha > 0$: M sensitive to downward movements
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- Concerned with *ranks* not *income levels*? Make status ordinal:
 - use estimated distribution function

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Mobility indices: Class 1 extended

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- The class can be extended by redefining status
- measure status from $-c$ rather than from 0
- $\frac{\theta(c)}{n} \sum_{i=1}^n \left[\left[\frac{u_i+c}{\mu_u+c} \right]^{\alpha(c)} \left[\frac{v_i+c}{\mu_v+c} \right]^{1-\alpha(c)} - 1 \right], \alpha(c) \neq 0, 1$

where $\theta(c) := \frac{1+c^2}{\alpha(c)^2 - \alpha(c)}$

Mobility indices: Class 1 extended

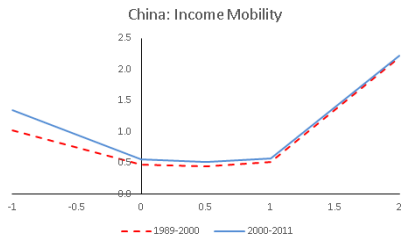
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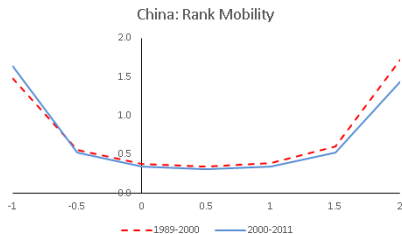
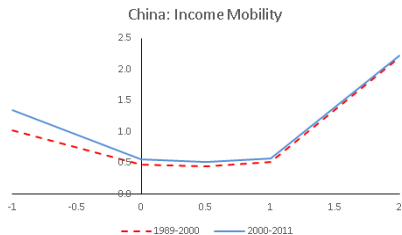
- In limit, as $c \rightarrow \infty$: $M'_\beta := \frac{1}{n\beta^2} \sum_{i=1}^n [e^{\beta[u_i - \mu_u - v_i + \mu_v]} - 1]$.
 - remains unchanged with absolute additions to status
 - β is a sensitivity parameter

M_α as a function of α : example

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Mobility measures: Class 2

- Focuses on aggregation of status differences $\sum_{i=1}^n a_i d_{(i)}$

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- **Case (a)**
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- **Case (b)**
 - make a_i sensitive to position i
 - $a_i = \phi\left(\frac{i}{n} - p - \frac{1}{2n}\right)$; $p := i^*/n$
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- **Special Case (b): linear ϕ**
 - weights are: $a_i = \frac{i}{n} - p - \frac{1}{2n}$
 - so $\Gamma_1 := \frac{1}{n} \sum_{i=1}^n \frac{i}{n} d(i) - \left[p + \frac{1}{2n}\right] \mu_d$,
 - $\Gamma_1 = 1/2G + \mu_d \left[\frac{1}{2} - p\right]$

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Decomposition: issues

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 - get different types of decomposition
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- Decomposition property follows from independence axiom
 - get different types of decomposition
 - depends on interpretation of the independence axiom
- Aspects in common with decomposition in other fields (social welfare, inequality):
 - by personal characteristics
 - by regions or countries
- Special for mobility:
 - Up and Down
 - similar to rich / poor decomposition for inequality
 - often treated as defining aspects of mobility
 - U: Ray and Genicot (2023); D: Bárcena and Cantó (2025)

Decomposition: Class 1

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 - $p_k \left[\frac{\mu_{u,k}}{\mu_u} \right]^\alpha \left[\frac{\mu_{v,k}}{\mu_v} \right]^{1-\alpha}$ weight on group k
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- Between group:
 - aggregation over mean changes of groups
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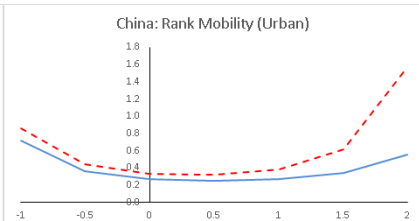
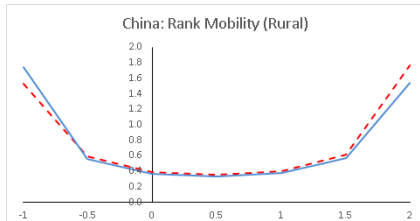
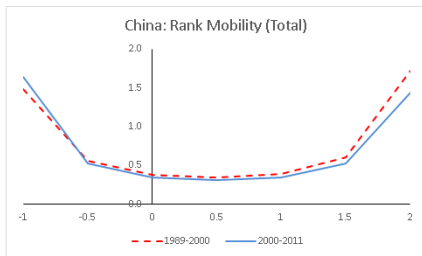
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- Partition population into “upward” U and “downward” D groups:
 - $M_\alpha = w^U M_\alpha^U + w^D M_\alpha^D + M_\alpha^{\text{btw}}$
 - compare Bárcena and Cantó (2025), Ray and Genicot (2023) eq (27)

Decomposition: Class 1 (Example)

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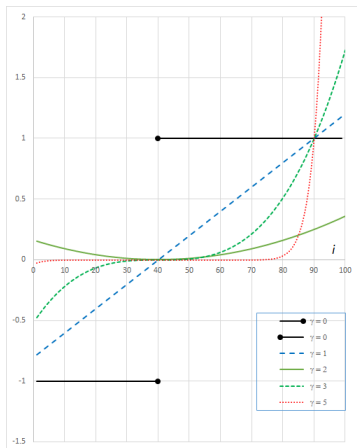
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- FO indices:
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- For general case $a_i = \phi\left(\frac{i}{n} - p - \frac{1}{2n}\right)$
 - $\Gamma_\gamma = p^{\gamma+1}\Gamma_\gamma^D + [1 - p]^{\gamma+1}\Gamma_\gamma^U$
 - $\phi(x) = x^\gamma$

Class 2: individual weights

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- Suppose the destination is an equal distribution

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 - $\frac{1}{n\beta^2} \sum_{i=1}^n \left[e^{\beta[u_i - \mu_u - v_i + \mu_v]} - 1 \right]$ become Kolm indices
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- Blackorby and Donaldson (1980, Bossert and Pfingsten (1990, Cowell and Flachaire (2021)

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Summary

- The approach:
 - separate “status” from “aggregation” issues
 - focus on meaning of mobility comparisons.
 - characterise “suitable” measures
- The results
 - two broad classes of mobility indices
 - each class satisfies the minimal set of requirements for mobility comparisons
 - each of these classes has a natural interpretation in terms of distributional analysis

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