

Beyond GDP and beyond Gini: the measurement of the inequality of opportunity

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Inequality at center stage...

...but what inequality?

1. *Increasing "inequality"* as a major political debate in a number of countries
2. Two recent key surprise electoral results (Brexit, Trump election) broadly attributed to *"inequality"* (and globalization!)
3. What inequality are we talking about ?
 - France is a moderately unequal country, yet populism (left and right) are close to 50% of the votes, very much on an inequality platform

Inequality cannot be reduced to the Gini coefficient of equalized disposable income

Other dimensions of inequality are crucial: the inequality of opportunity (i.e., access to income generating facilities)

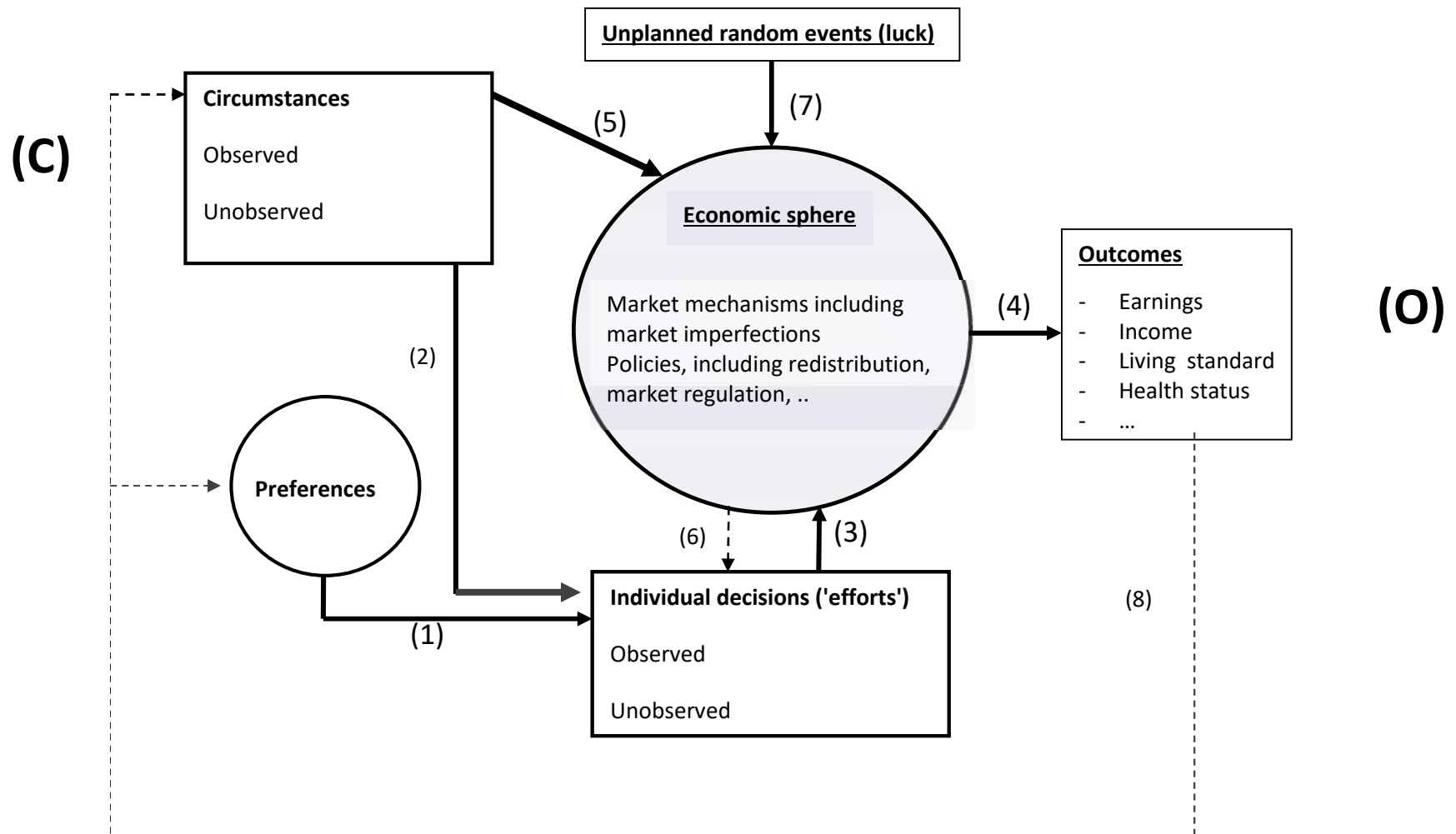
The importance of the inequality of opportunity

- Inequality of opportunities (I_{opp}) matters:
 - Per se (moral philosophy argument)
 - As a determinant of the inequality of outcomes
 - In a policy perspective
 - In a political perspective , through the perception people have of "inequality"
 - Example: the differentiated perception of social mobility: "Our children will not do as well as we did"
- Crucial to monitor the inequality of opportunity at the same time as that of outcomes
- How to monitor I_{opp} ? Measures and data requirements

Outline

1. The relationship between 'opportunities' and 'outcomes'
2. Outcome-based measures of I_{opp}
 - Non-parametric
 - Outcome means by type
 - Outcome distribution by type
 - Matrix representation
 - Parametric
 - Theoretical background
 - Intergenerational mobility elasticity (IGM)
 - Generalizing the IGM
 - Introducing cohort effects
3. Direct I_{opp} measures : the case of education
4. Conclusion

1. The relationship between individual circumstances, opportunities and outcomes



The logics of I_{opp} measurement

- I_{opp} based on inequality of the distribution of circumstances, or some particular dimension of **(C)**
- I_{opp} based on impact of the distribution of **(C)** on the distribution of a component of **(O)**
 - I_{opp} depends on the set of circumstances in **(C)** and the component of **(O)** being considered
 - Inequality in **(O)** as a 'metric' of I_{opp}

Note : I_{opp} will differ according to the component of (O) being considered (earnings, income, ... happiness)

2. Outcome-based measures of I_{opp}

a) Non-parametric measures

i. Measure based on outcome *means* by 'types'

- 'Type' = individuals facing a given set of circumstances
- Example: *'female, born in rural area of region X, from uneducated but wealthy parents'*
- Define the virtual outcome distribution where all individuals with a given type get the mean outcome of that type:

Type	Mean outcome	Number of people
1	m1	n1
2	m2	n2
etc..		

Measures based on outcome means by type

- *Absolute* I_{opp} = inequality measure of outcomes in the virtual distribution:

$$I_{opp} = I^0 = I(m_1: n_1 \text{ times}; m_2: n_2 \text{ times}; \dots; m_N: n_N \text{ times})$$

- Inequality measure $I()$ may be Gini, Theil, Log variance, ...
- *Relative* inequality of opportunity =
Absolute inequality of opportunity, I^0 /actual inequality of outcomes
- I^0 = Familiar *between group inequality component* of total outcome inequality in a partition of the population

Examples

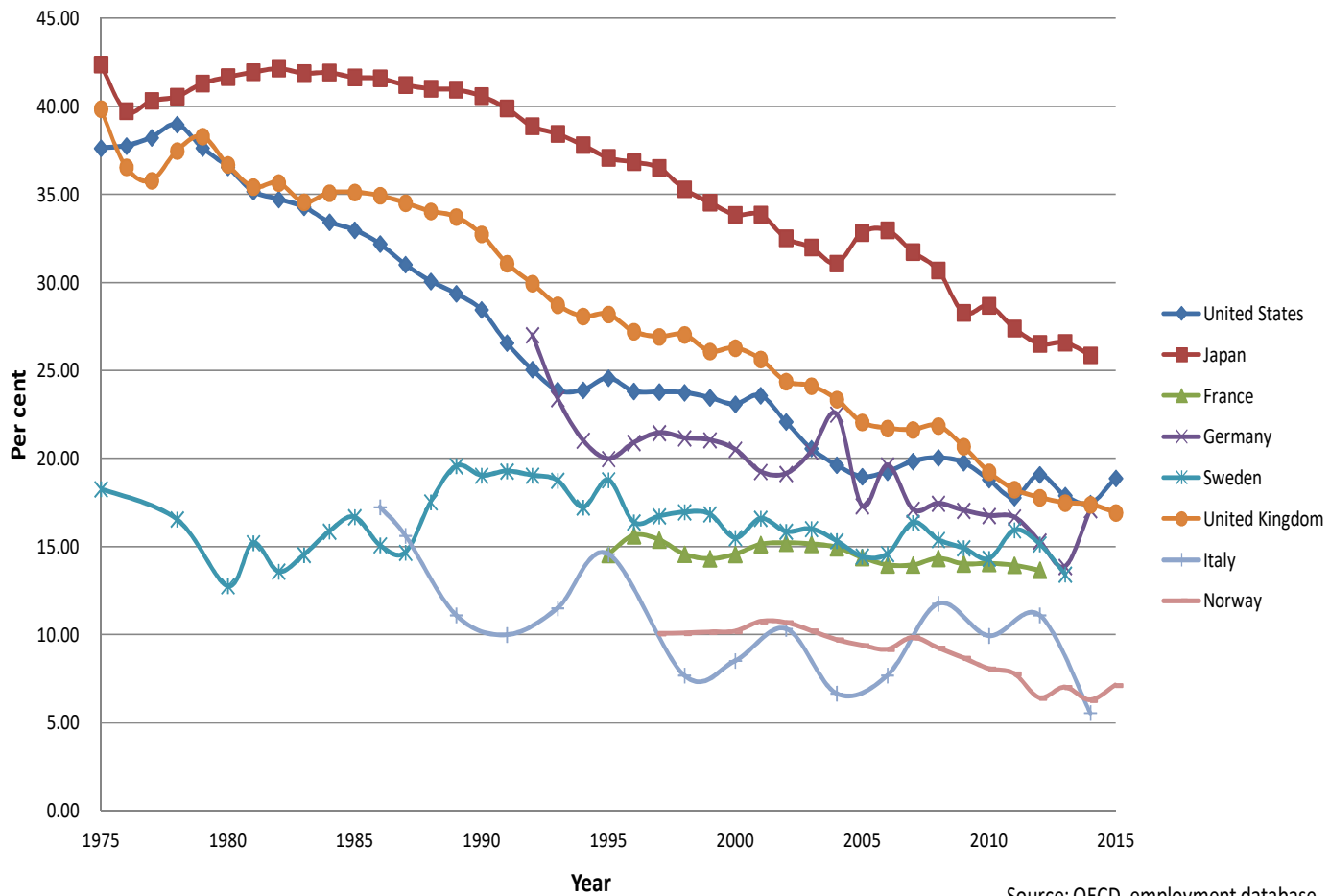
- Case of 2 types – i.e. gender
- Outcome = earnings

I_{opp} :

- Absolute gap = Male mean earnings - Female mean earnings
- Relative gap = Absolute gap / male (female) earnings

Figure 7. Gender wage gap in selected countries: 1975-2015

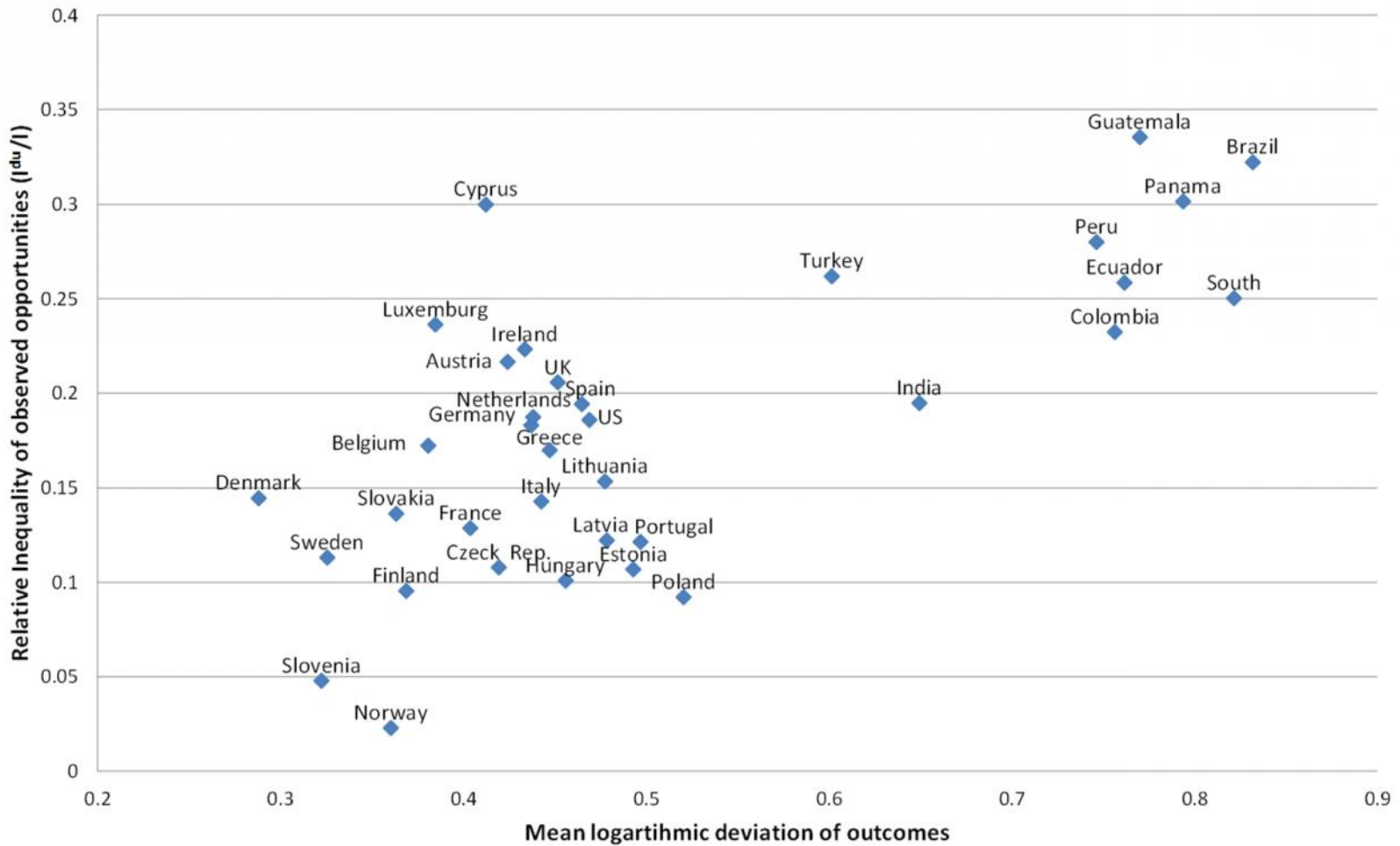
Percentage gap between median men and women earnings w.r.t. median mean earnings
(full-time workers)



Examples

- Types defined by: gender, education, region of birth, ...
- Outcome = earnings/household income per capita
- I_{opp} Between type inequality, mean logarithmic deviation
- Various countries

Figure 6. Inequality of outcomes and share due to observed dimensions of the inequality of opportunity: selected countries circa 2005 (Inequality measure = Mean logarithmic deviation)



Non-parametric measures

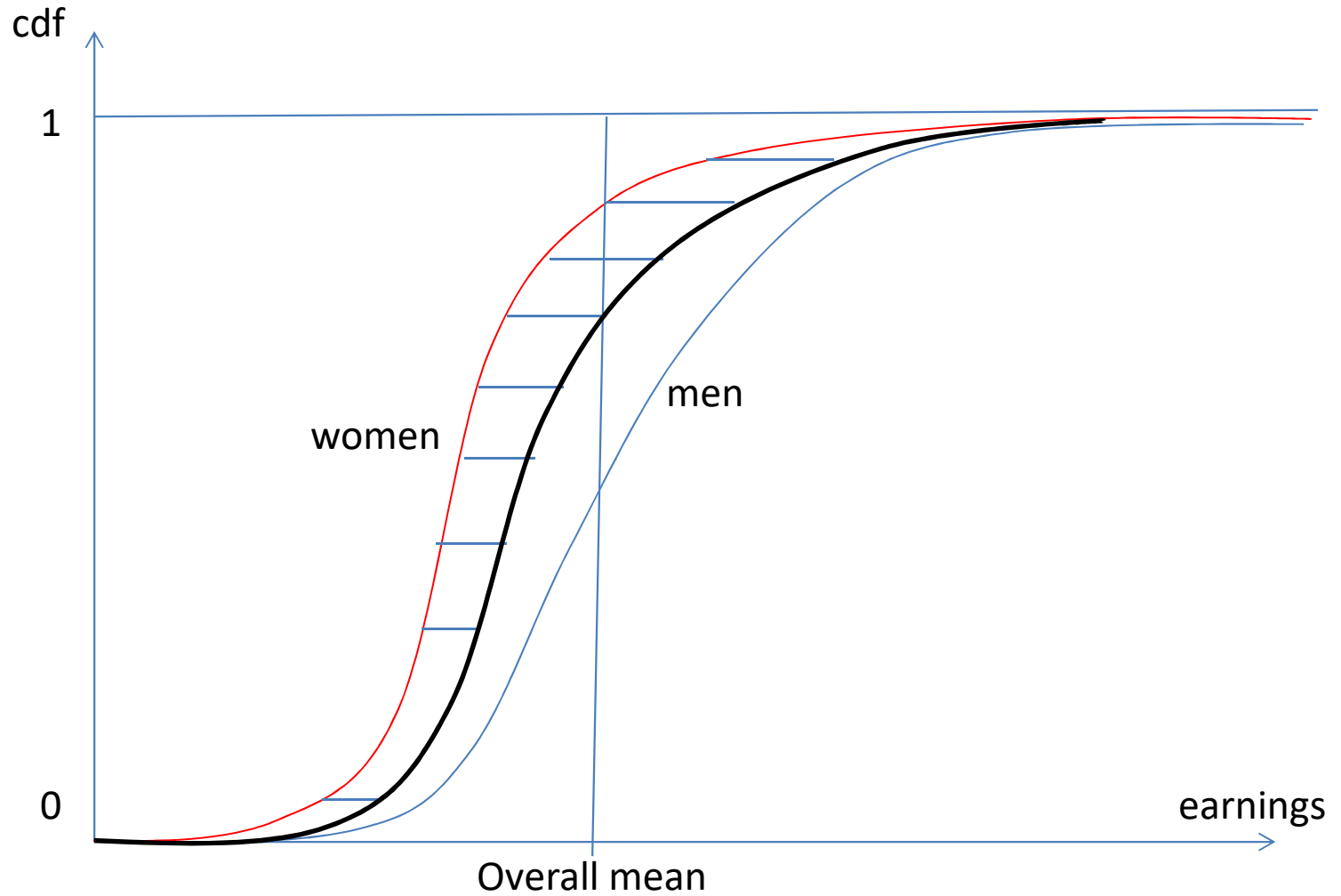
ii. Measure based on the outcome *distribution* by 'types'
(Roemer)

- Instead of measuring difference in outcome means across types, compare outcome quantiles
- I_{opp} based on aggregation of 'quantile gaps'
- Based on Roemer's criteria, I_{opp} could be defined as:

$$\int_0^1 [\bar{q}(f) - \text{Min}_t q_t(f)] df$$

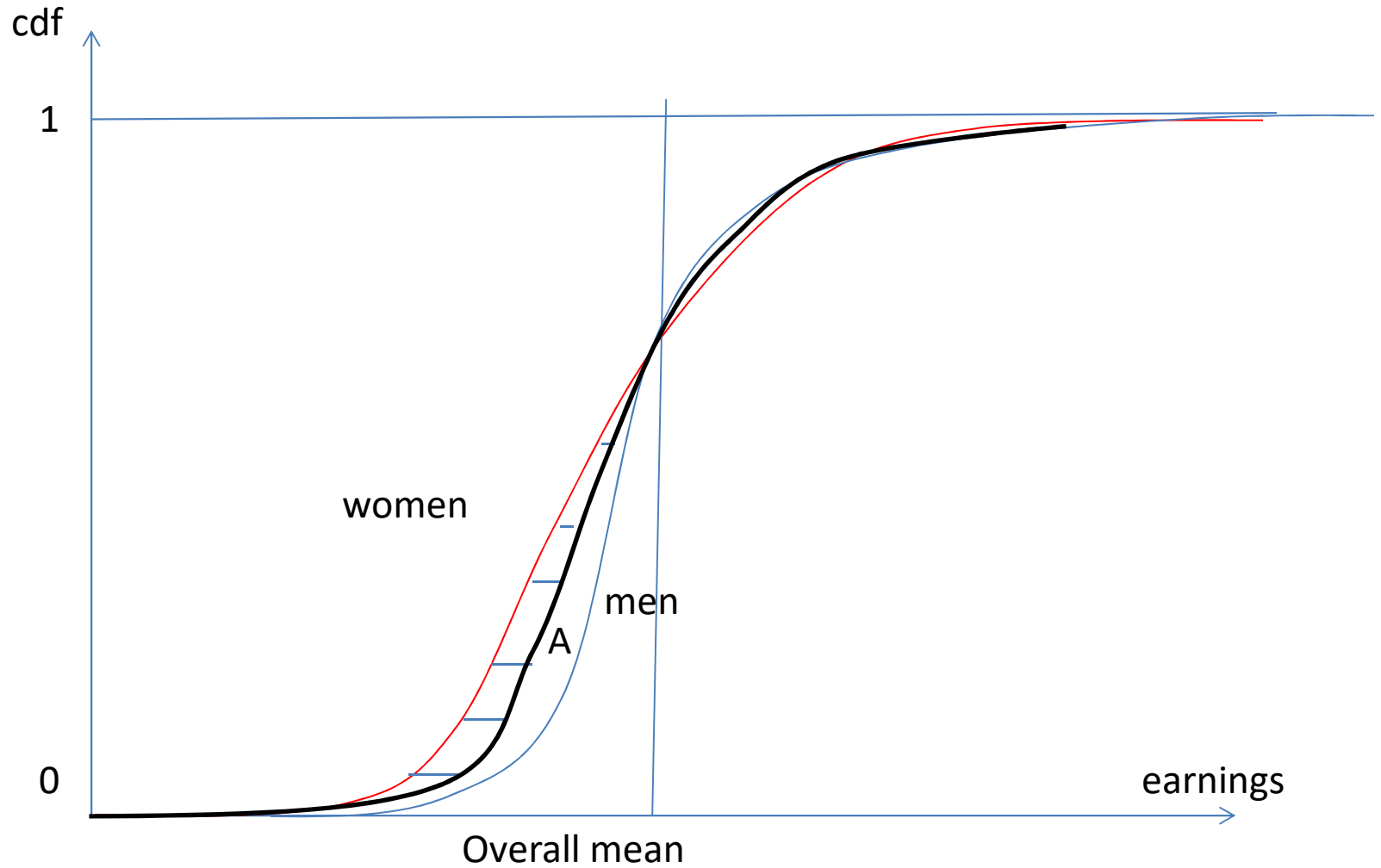
with $q_t(\pi) =$ quantile of order π for type t

Case of non-crossing cdf and two types:
 $I_{opp} = \text{area between upper cdf and mean cdf}$



Case of crossing cdf

I_{opp} = area between envelope of cdf and mean cdf



Non-parametric measures

iii. Matrix representation

Mostly used for inter(-intra) generational mobility

- '*Types*' = brackets of parental income/earnings /education
- *Outcome* = brackets of present generation distribution of income/earnings/wealth
- I_{opp} = measures based on the corresponding matrix (Huge literature on them + dominance criteria)

Matrix representation of the distribution of opportunities (intergenerational mobility case)

Figure 3. Earnings intergenerational transition matrix (P)

Fathers \ Sons	Y_1	Y_2	Y_3	...	Y_N	Total
Y_1	P_{11}	P_{12}	P_{13}		P_{1N}	$P_{1.}$
Y_2	P_{21}	P_{22}	P_{23}		P_{2N}	$P_{2.}$
Y_3	P_{31}	P_{32}	P_{33}		P_{3N}	$P_{3.}$
...
Y_N	P_{N1}	P_{N2}	P_{N3}		P_{NN}	$P_{N.}$
Total	$P_{.1}$	$P_{.2}$	$P_{.3}$...	$P_{.N}$	1

b) Parametric measures

i) Theoretical background

- Canonical model

$$y_i = f(C_i, e_i) + u_i \quad (1)$$

with C = 'circumstances', 'e' = efforts and u = unobserved circumstances and efforts

- Inequality of opportunity: direct unfairness

$$I^{O,du}(\tilde{e}) = I(\tilde{y}.); \quad \tilde{y}_i = f(C_i, \tilde{e}) + u_i \quad \text{close to } I^O \text{ above}$$

- Inequality of opportunity : Fairness gap

$$I^{O,fg}(\tilde{C}) = I(y.) - I(y_i^{norm}(\tilde{C})); \quad y_i^{norm}(\tilde{C}) = f(\tilde{C}, e_i) + u_i$$

Close to R^2 of OLS on (1) log-linear

Ambiguous status of 'residual' u

Parametric measures

ii) Intergenerational mobility, IGM

Model

$$\ln y_i = \alpha + \lambda \ln y_{iF} + u_i \quad y_{iF} = \text{income of father}$$

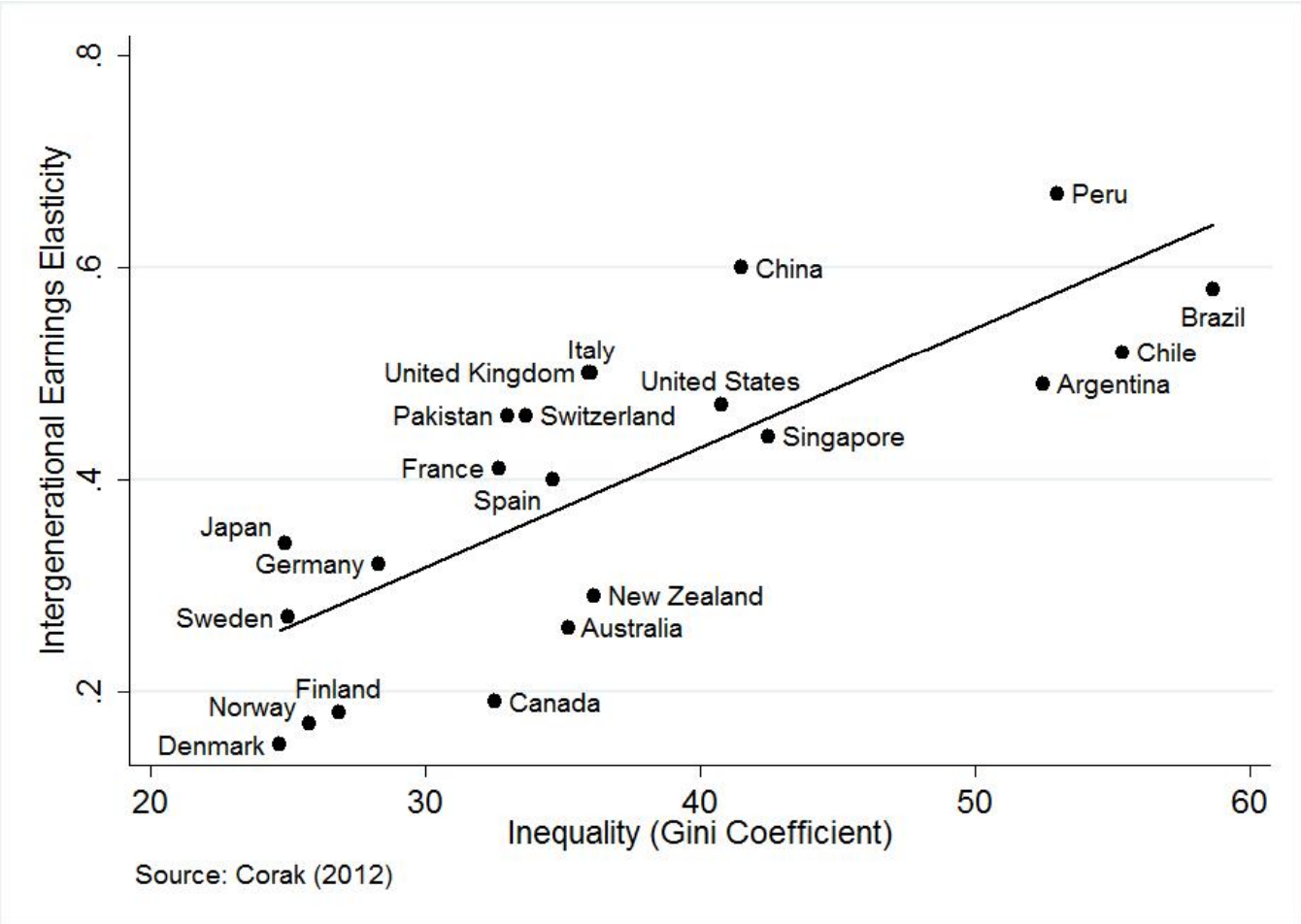
IGE = intergenerational elasticity = $\hat{\lambda}$

$$I^{o,fg} = R^2 = \hat{\lambda} \cdot \frac{\text{Var}(\ln y)}{\text{Var}(\ln y_F)}$$

I_{opp} = IGE if the variance of the (log) earnings does not change across generations

But, why considering the earnings of parents as the only circumstance affecting the earnings of children?

Illustration: the Great Gasby curve



Parametric measures

iii) Generalizing the IGM model

$$\text{Ln } y_i = S \cdot Z_i + u_i \quad (2)$$

where Z = all circumstances to be taken into account:

$$I^{o,fg} = R^2$$

Or:

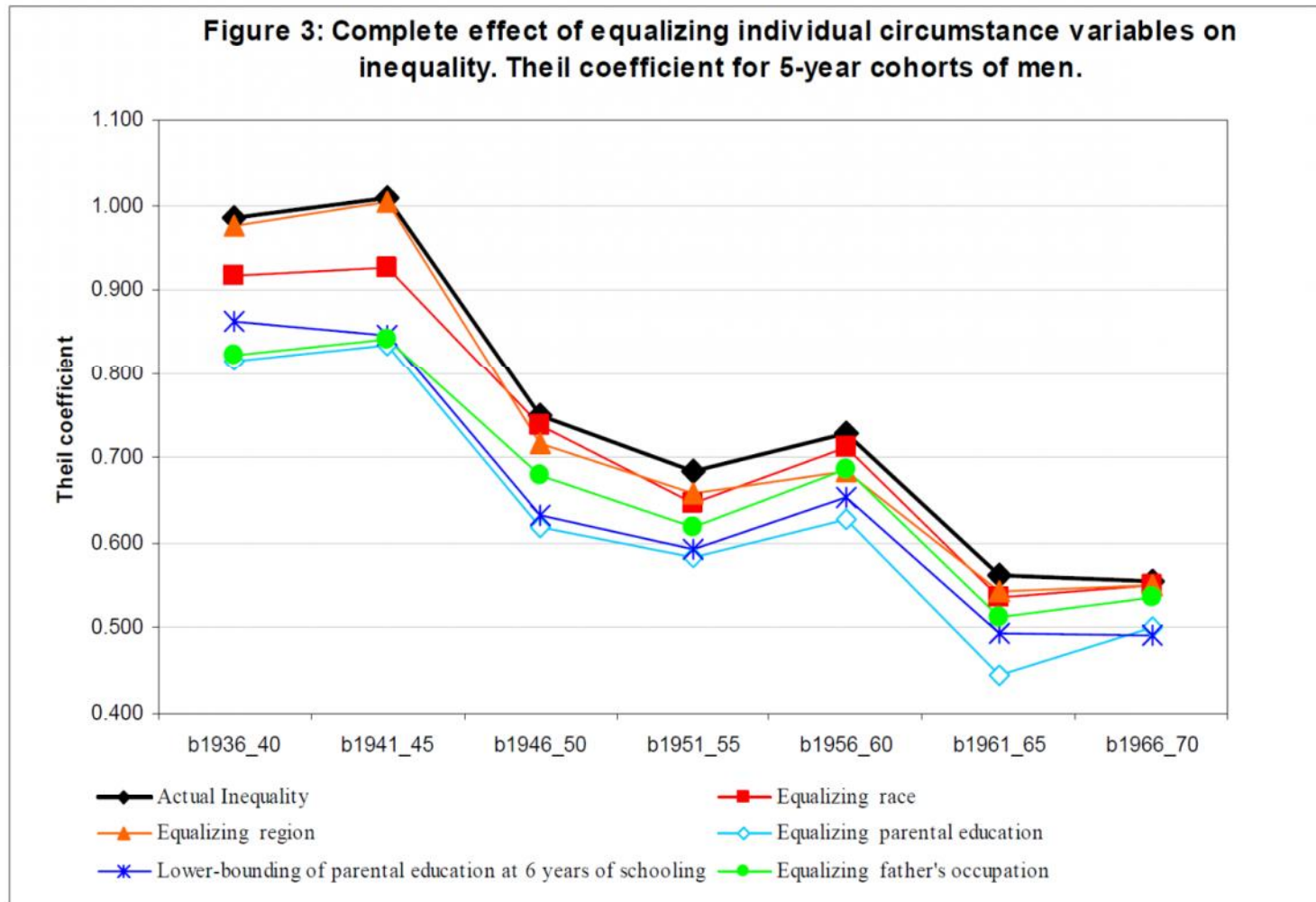
$$I^{O,fg} = I(y_i) - I(y_i^{norm}(\bar{Z})); \quad \text{Log } y_i^{norm}(\bar{Z}) = S \cdot \bar{Z} + u_i$$

$$I^{O,du} = I(\tilde{y}_i); \quad \text{Log } \tilde{y}_i = S \cdot Z_i$$

Note: Model (2) identical to model (1) (IGM) when y_{iF} is instrumented by Z (as in Aronsson and Mazumder, 2008)

Example: See Brunori, Ferreira, Peragine (2013)

Generalized IGM and cohort analysis



From Bourguignon, Ferreira, Menendez (2007)

Cohort analysis

- Distinguishing cohorts in regression analysis is essential as inequality of opportunities are likely to be age/cohort dependent
- Estimating model (2) at regular time intervals should allow to monitor I_{opp} over time.

A remark on gender gap

General model (2)

$$\ln y_i = \alpha .G_i + \beta .Z_i + u_i$$

with G = gender dummy (1 = male; 0 = female) and Z = personal characteristics (education, parental background, ...)

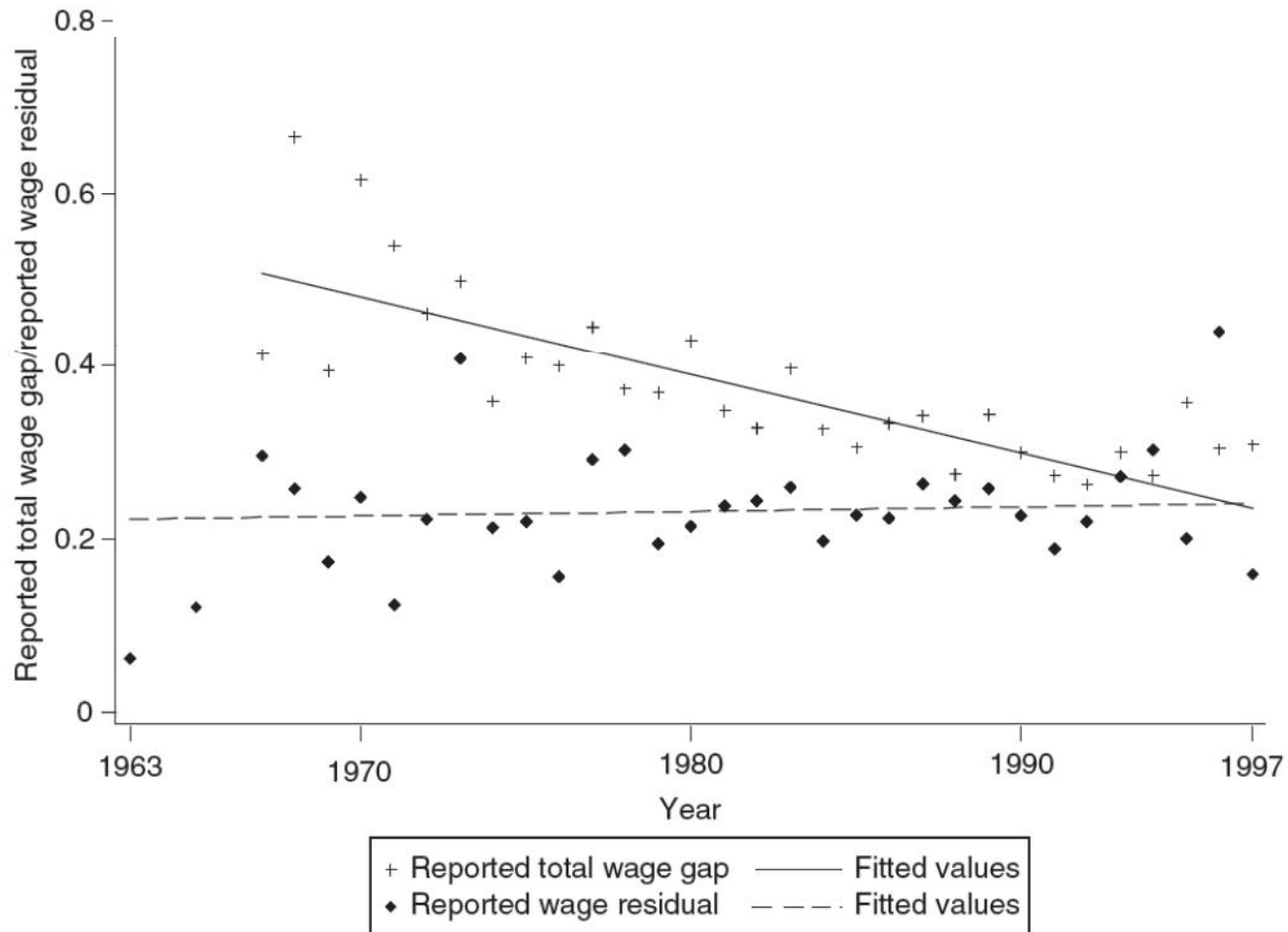
α = Residual gap (once full gap corrected for gender differences in Z) and differs from 'total wage gap'

Yet, it is the total gap that matters (if focusing on gender)

Note: to what extent should Z include variables like working time or job experience ?

Figure 8. Gender wage gap and residual wage gap in a meta-analysis of the wage gap decomposition literature

Source: Weichselbaumer and Winter-Ebmer (2005)

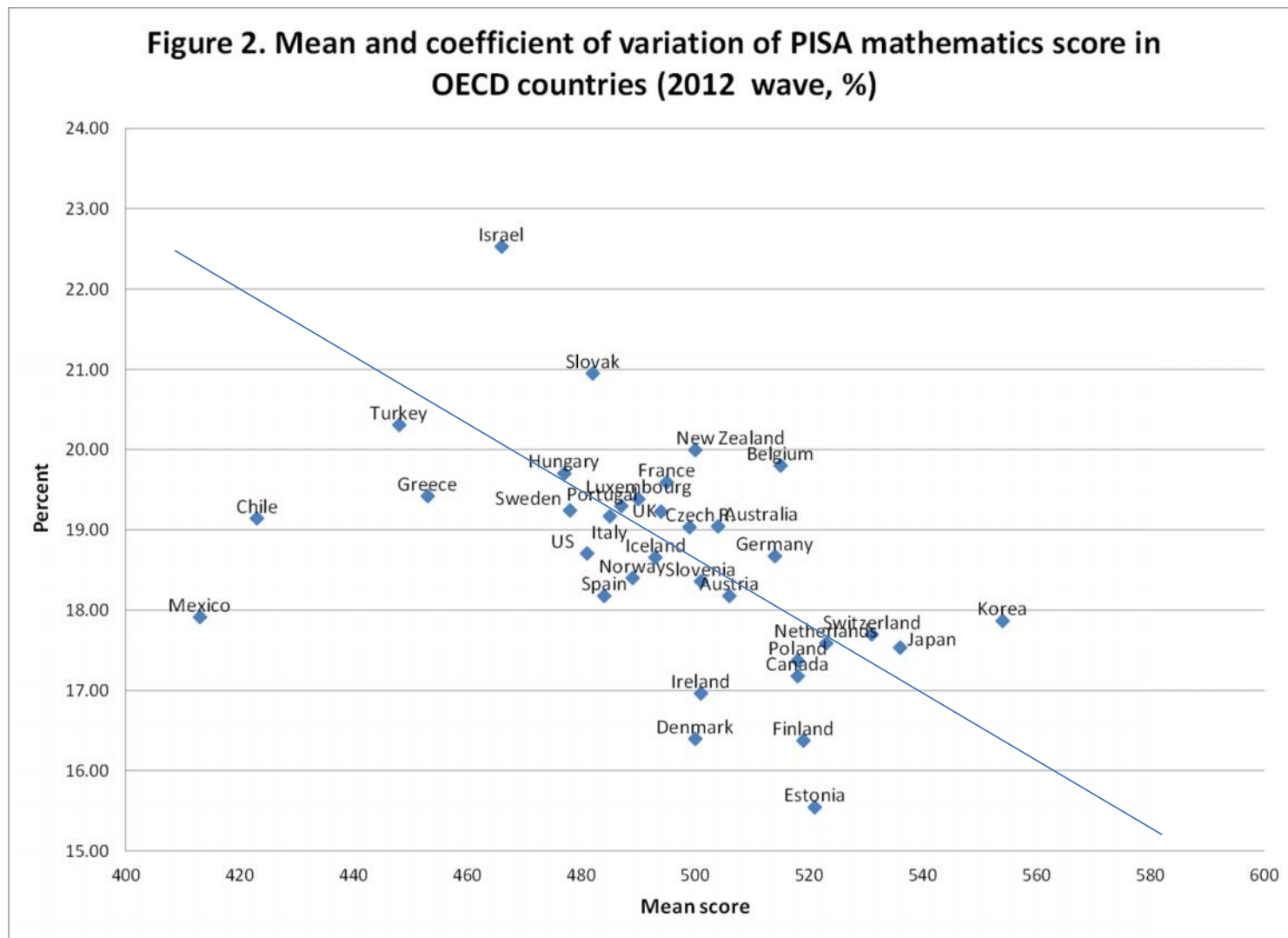


3) Direct measure of the inequality of opportunity

$$y = f(C,e) + u$$

- Measure directly the inequality of opportunity on C rather than through the effect of C on y
- Case of *educational achievement*: inequality measured by inequality of test scores PISA types
- Question: weight of that component of C in y? (less than 5% in Murnane, 2000)
- Educational as a circumstance or an outcome, explained by other circumstances

Inequality of Pisa scores



4. Conclusions

- Monitoring of the inequality of outcome (income, earning, wealth) *needs to be complemented* by that of opportunity
- Monitoring the inequality of opportunity = *monitoring of SOME observed circumstances* and/or their effect on the inequality of outcomes
- Even though a lower bound of the actual inequality of opportunity – which cannot be evaluated anyhow – monitoring it would respond to strong social demand
- Social mobility (IGM) as only one of components of the I_{opp} although possibly an important one

Priority statistics to monitor the inequality of opportunity

- The inequality of economic outcomes (earnings, income) arising from parental background and its share in total inequality of outcome.
- Variance analysis of scores in PISA and possibly surveys at younger ages
- (Gender (ethnic) inequality in employment and earnings)

THANK YOU