

# Inequality of Opportunity in Educational Achievements

Cross-Country and Intertemporal Comparisons

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Inequality – Measurement, trend, impacts and policies

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# Outline

- 1 Aim & Motivation
- 2 Model
- 3 Data
- 4 Results
- 5 Conclusions
- 6 Appendix

# Research questions

- 1 Does the country ranking change when we switch the focus of the analysis from average test scores to fairness?
- 2 Is there any country that outperform in both the level and the degree of fairness?
- 3 There has been any change in the strength of the association between socio-economic characteristics and students' performances?

# Motivations

- Education influences labour market participation, civic engagement, health status, earnings, social mobility, etc. (Blau & Kahn, 2005; Hanushek & Woessmann, 2010; among others).
- Intergenerational persistence in educational achievements (Marks, 2005; Macdonald et al. 2010; Ermisch et al. 2012)
- Inequality in educational attainments (Thomas et al. 2001; Morrison & Murtin, 2007)
- Inequality in educational achievements (Brown et al., 2007; Micklewright et al. 2007)

## Existing evidences

- de la Vega & Lekuona (2013): PISA 2009
- Gamboa & Waltenberg (2011) PISA 2006 & 2009, LAC
- Ferreira & Gignoux (2011) PISA 2006

## What's new?

- 1 PISA 2012
- 2 Changes over time (PISA 2003, 2006, 2009, 2012)
- 3 How do the less advantaged students perform?

# Model

Adapt to our framework the idea of measuring fairness through an ordered pair (Roemer, 2013):

$$EduOpp = (W^{EEOp}, IEOp)$$

- $W^{EEOp}$ : focuses on worst-off students
- $IEOp$ : looks at the whole sample

# Outcome function

Test scores ( $s$ ) depend only on circumstances ( $c$ ) and effort ( $e$ )

$$s_i = f(c, e)$$

- $c$  used to partition students into  $K$  ( $j = 1, \dots, K$ ) types
- $e$  correspond to the rank  $\pi$  occupied by each student in its own type distribution of test scores
- $v^j(\pi)$ : level of  $s$  for individuals in type  $j$  occupying the rank  $\pi$

$$W^{EEOp} = \int_0^1 \min_j(\pi) d\pi \quad (1)$$

- Class-ranked situations:  $W^{EEOp}$  corresponds to the average score of the worst-off students (Roemer, 2013)
- Not class-ranked situations:  $W^{EEOp}$  corresponds to the left-hand envelope of the distribution of CDFs (Roemer, 2013)
- Empirically this involves the estimation, for each country, of each type-specific CDF and their envelopes



# IEOp

- Ex-Ante Approach: *IEOp* measured as between type inequality in mean outcome
- Parametric procedure

$$s_i = \beta k_i + \varepsilon_i$$

- Index of Inequality

$$IEOp = \frac{\text{var}(k, \hat{\beta})}{\text{var}(y)}$$

# Dataset

## OECD

- PISA 2003: 41 countries
- PISA 2006: 57 countries
- PISA 2009: 74 countries
- PISA 2012: 65 countries

## Domains

- Mathematics
- Science
- Reading

# Sample

2 stages sampling procedure

Students aged between 15 years and 3 months and 16 years and 3 months enrolled in grade 7 or higher

Raw test scores ( $s$ ) scaled by using IRT and then standardized

$$s_i = \hat{\mu} + \frac{\hat{\sigma}}{\sigma}(x_i - \mu)$$

where  $x_i$  is the test score of student  $i$ ,  $\hat{\mu} = 500$  and  $\hat{\sigma} = 100$  are the arbitrary (final) grand mean and SD

# Available Data

PISA contains information on:

- Schools' policies and practices
- Students' background
- Students' motivation
- Students' learning style

## Some empirical issues

$$EduOpp = (W^{EEOp}, IEOp)$$

- $W^{EEOp}$  focuses on the worst-off type: the omission of relevant circumstances determines a measure of “social welfare”  
 UPWARD biased
  - Intuition: when a new circumstance is added there is at least one additional type-distribution, conditional to a given value of the new circumstance, which is going to be at its left
- $IEOp$  looks at the whole population: the omission of relevant circumstances determines a measure of inequality which is DOWNWARD biased; some variation is left unexplained and attributed to effort.
- As # of  $K \uparrow \Rightarrow W^{EEOp} \downarrow$  and  $IEOp \uparrow$

# Variables

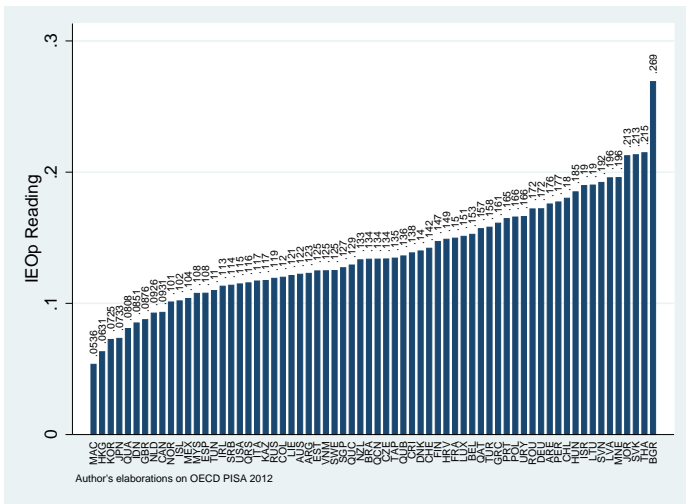
## Students' circumstances

- Gender
- Parental level of education
  - 1  $ISCED \leq 2$
  - 2  $3 \leq ISCED \leq 4$
  - 3  $ISCED \geq 5$
- Parental job classification
  - 1 White collar
  - 2 Blue collar

12types

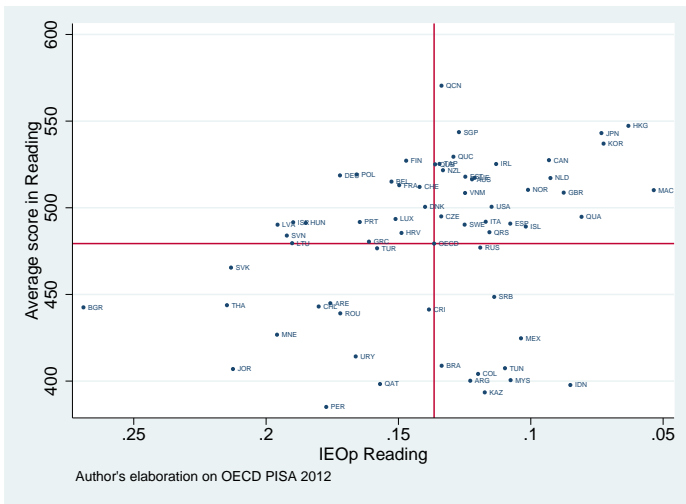


# IEOp in Reading, 2012

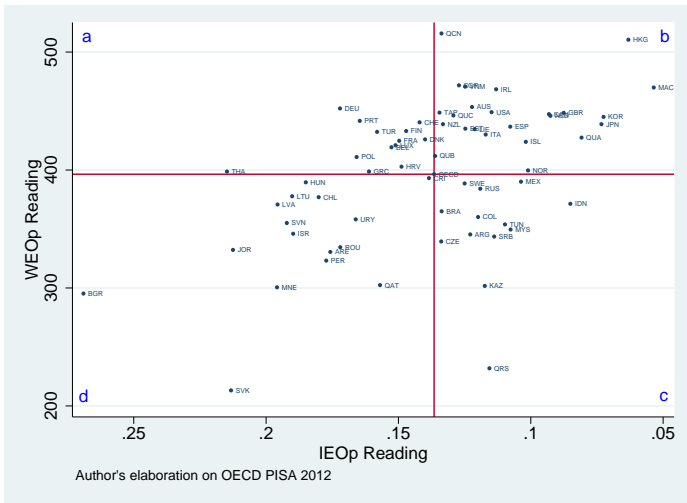




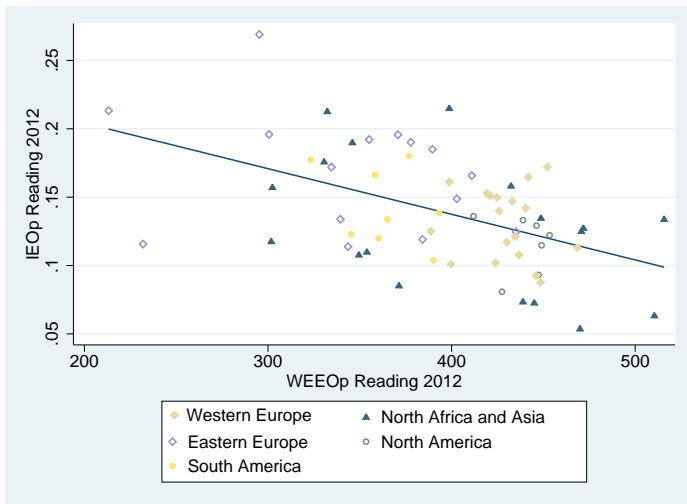
# Average performance and *IEOp* in Reading, 2012



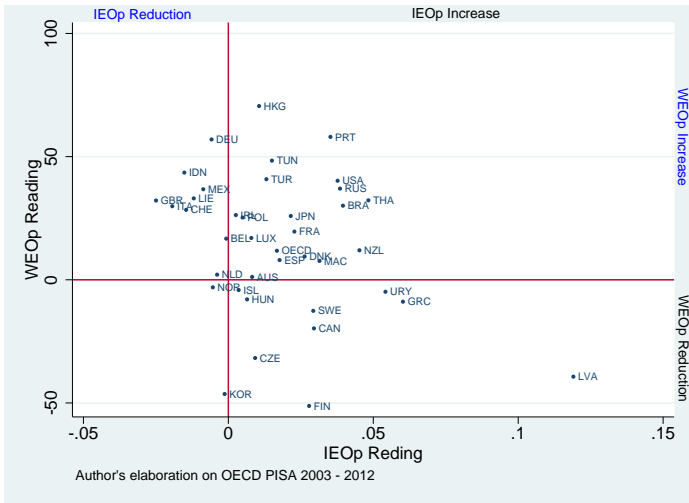
# Is there any “outperforming” country?



# Geographical Pattern



# Changes over time, 2003 – 2012



## Limits of the analysis

- Due to the omission of relevant circumstances the two components risk to be biased, so caution is necessary in interpreting the results.
- PISA involves only students who do not drop out and have not repeated too many grades.

With these caveats in mind...

# Conclusion

- High heterogeneity across countries in terms of both levels and degree of fairness in education
- The strength of the association between parental background and students' test scores tends to be higher in Reading than in Math and Science
- This association is, on average, lower in countries that perform better in average test scores
- There aren't countries that outperform in both dimensions of fairness
- $W^{EEOp}$  tends to be higher and  $IEOp$  lower in some Asiatic countries, in North America and in Western European countries where also variability is lower

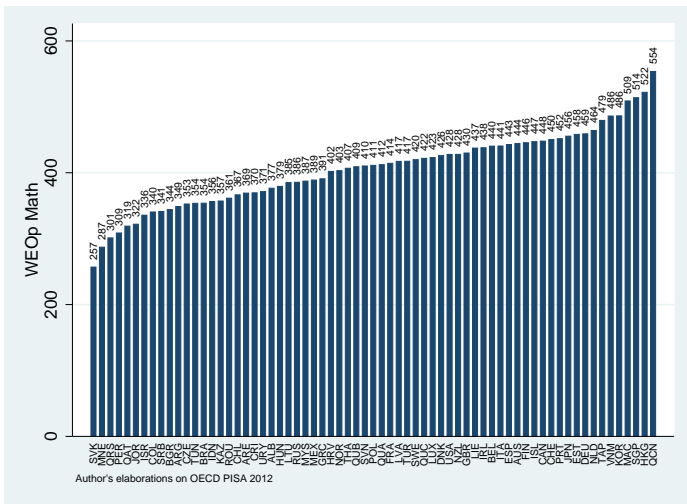
## Conclusions (ctd.)

- Eastern European countries occupy an intermediate position in terms of  $IEOp$
- Between 2003 and 2012  $\uparrow$  in  $W^{EEOp}$  has been accompanied by  $\uparrow$  in  $IEOp$
- Few countries moved toward lower degree of  $IEOp$  all the while improving the performances of the less advantaged students.
- Most of them, with the exceptions of Indonesia and Mexico, are Western European.

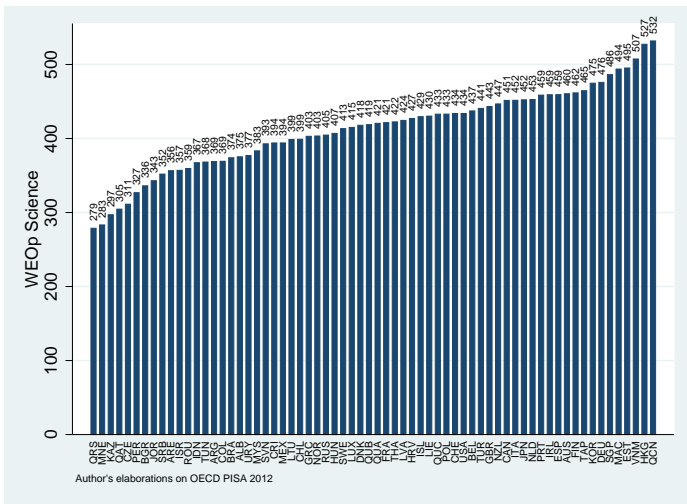
# THANK YOU!



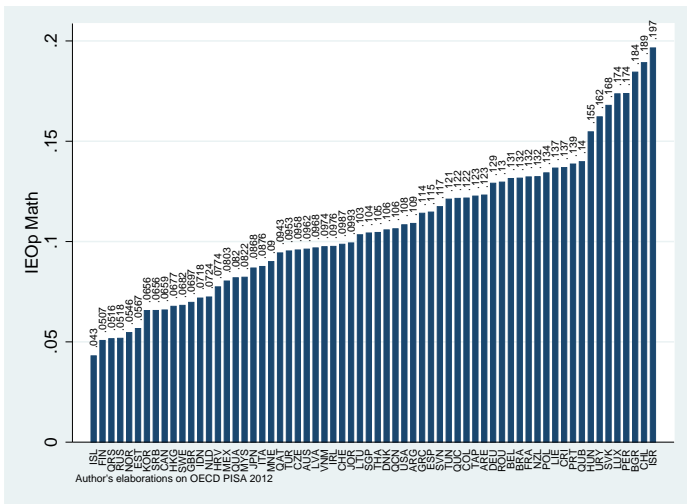
# W<sup>EEOp</sup> in Mathematics, 2012



# W<sup>EEOp</sup> in Science, 2012



# IEOp in Mathematics, 2012



# IEOp in Science, 2012

