# Effects of demographic and educational changes on the labor markets of Brazil and Mexico

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#### **Research questions**

- Main objective: estimate the impact of demographic and educational changes on earnings in Brazil and Mexico:
  - What are the effects of changing age and educational compositions on male earnings at the aggregate level?
  - How does the concentration of skilled workers affect the private and social returns to education at the **individual level**?
- Within the labor force (15–64 years of age), the population is getting older and better educated with regional variation.
- Age and education increase earnings.
- Larger proportion of older and more educated males causes:
  - Negative impacts on earnings of competing workers.
  - Greater knowledge and economic dynamism.

#### **Cohort size**

- Human capital: schooling and work experience have positive impacts on earnings (Mincer 1974).
- Baby boom: large cohorts of better educated individuals entered the U.S. labor market, decreasing their relative earnings.

(Berger 1985; Bloom and Freeman 1986; Bloom, Freeman, and Korenman 1987; Easterlin 1978; Freeman 1979; Sapozknikov and Triest 2007; Welch 1979)

– Larger cohorts also had positive impacts on labor outcomes. (Autor, Katz, and Krueger 1998; Katz and Autor 1999; Katz and Murphy 1992; Shimer 2001)

 Effects of cohort size on the labor market have been estimated for several developed countries.
(Biagi and Lucifora 2008; Borjas 2003; Brunello 2010; Korenman and Neumark 2000; Skans 2005)

### **Concentration of human capital**

- Social returns to education: concentration of well-educated people benefits everyone else in the population.
  (Acemoglu 1996; Glaeser 2011; Moretti 2011)
- Other positive impacts: concentration also generates greater knowledge and economic dynamism.
  (Moretti 2004a, 2004b; Glaeser 2011; Berry and Glaeser 2011)

### Several studies for developed countries, but much less is known about developing countries. (Queiroz and Golgher 2008; Amaral et al. 2013; Rigotti 2006)

#### Main contribution

- Few studies have addressed how demographic and educational compositions affect earnings, as well as private and social returns to education in **developing countries**.
- Contributes to the literature on demographic change in developing countries by predicting earnings using:
  - Variations in age-education composition.
  - Regional differences.
- This project is part of a broader research agenda dealing with the effects of population changes on demographic, social, and economic outcomes.

#### **Brazil & Mexico**

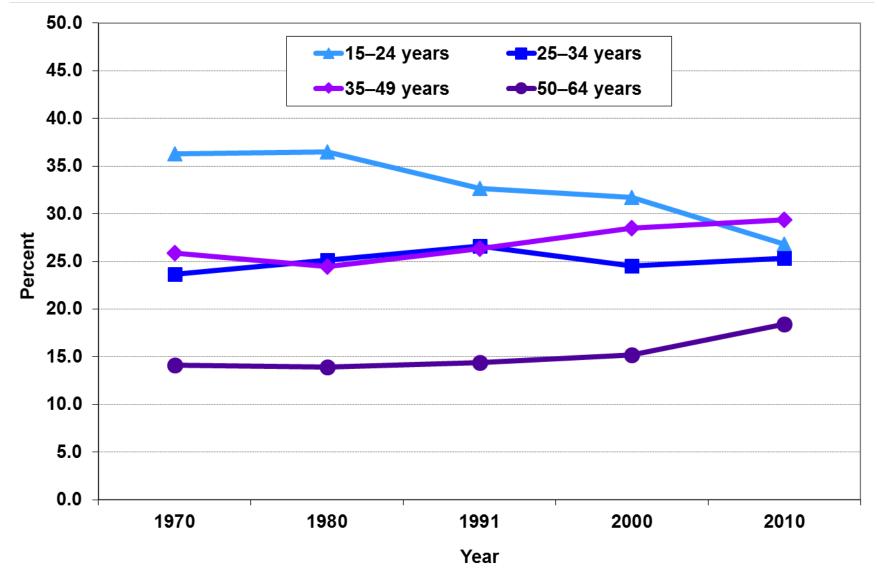
# - Fertility decline is contributing to changes in age composition (IBGE 2012; CONAPO 2004, 2014).

Total Fertility Rate	1970	2010
Brazil	5.8	1.9
Mexico	6.8	2.3

- Educational expansion began late and has a long way to go (Barro and Lee 2001; Marcílio 2001, 2005; Rios-Neto and Guimarães 2010).

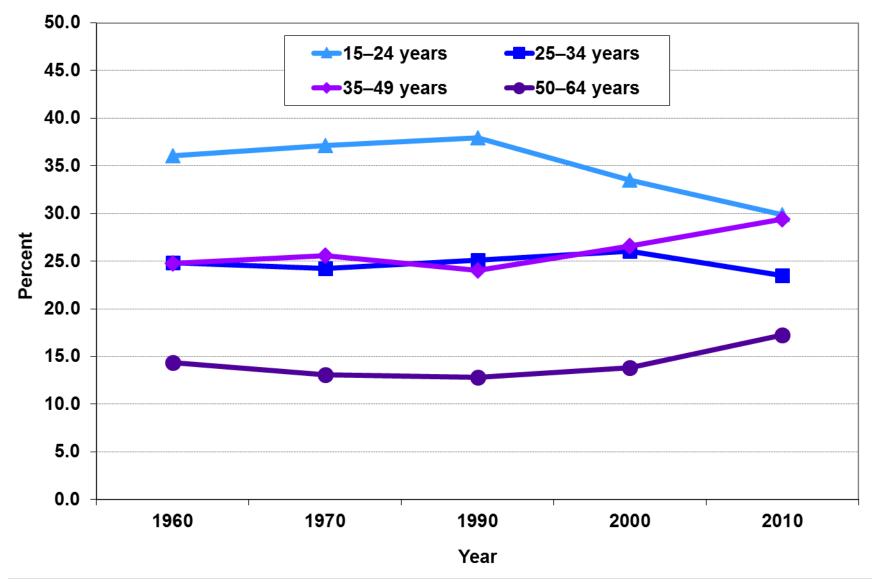
- Improvement in educational attainment coincides with decline in family size and school-age children (Lam and Marteleto 2005, 2008).
- These countries have data that captures information on:
  - Population aging.
  - Educational improvement.
  - Geographic variation.

#### Male age composition Brazil, 1970–2010



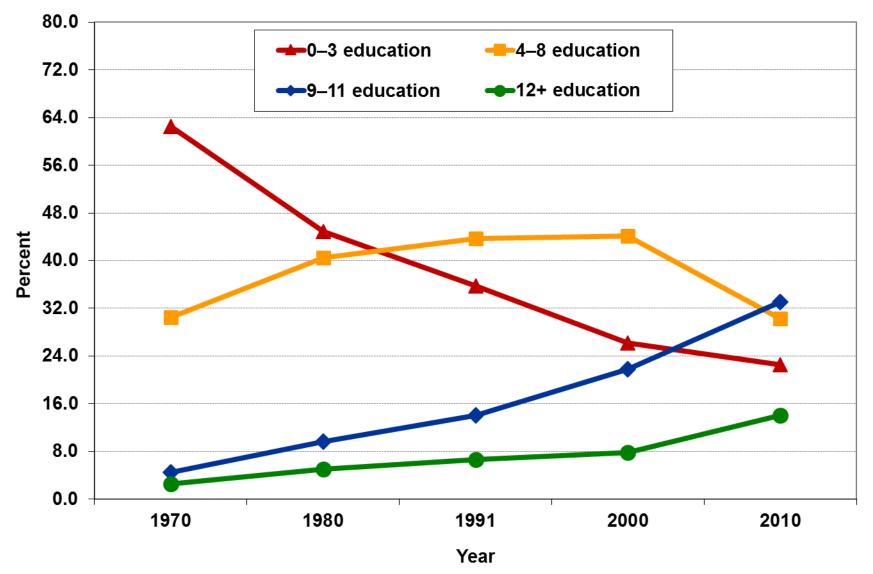
Source: 1970, 1980, 1991, 2000, and 2010 Brazilian Demographic Censuses.

#### Male age composition Mexico, 1960–2010



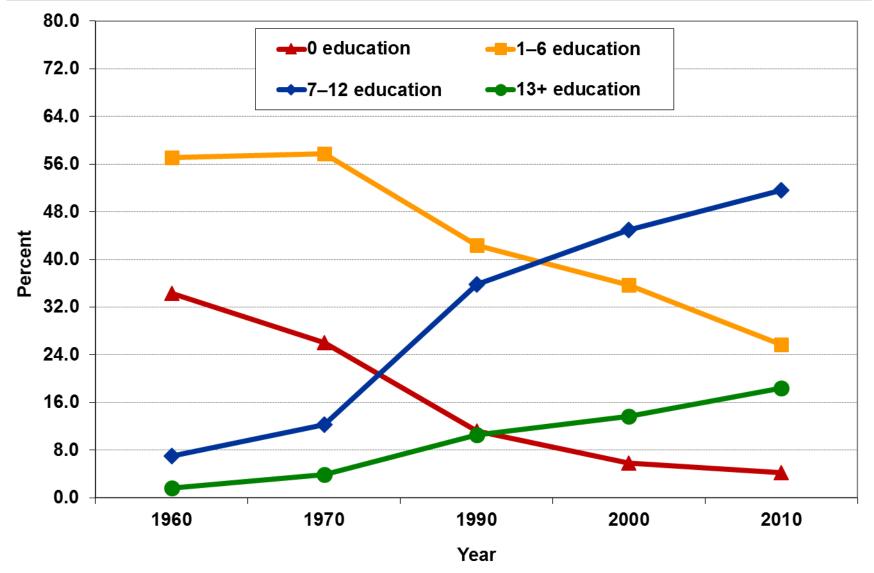
Source: 1960, 1970, 1990, 2000, and 2010 Mexican Demographic Censuses.

#### Male educational composition Brazil, 1970–2010



Source: 1970, 1980, 1991, 2000, and 2010 Brazilian Demographic Censuses.

#### Male educational composition Mexico, 1960–2010



Source: 1960, 1970, 1990, 2000, and 2010 Mexican Demographic Censuses.

#### Brazilian micro-data

- Brazilian Censuses: 1970, 1980, 1991, 2000, and 2010.
- **Minimum comparable areas:** 502 micro-regions.
- **Age** in years is categorized into four groups:
  - Youths (15–24).
  - Young adults (25–34).
  - Experienced adults (35-49).
  - Older adults (50-64).
- Education: four groups indicating years of schooling:
  - Incomplete first phase of primary school (0–3).
  - No further than primary school (4–8).
  - Secondary school (9–11).
  - At least some university (12+).
- **Earnings** from main occupation: converted to Jan. 2002.

#### Mexican micro-data

- Mexican Censuses: 1990, 2000, and 2010.
- **Minimum comparable areas:** 2,456 municipalities.
- **Age** in years is categorized into four groups:
  - Youths (15–24).
  - Young adults (25–34).
  - Experienced adults (35-49).
  - Older adults (50-64).
- Education: four groups indicating years of schooling:
  - No education (0).
  - Primary school (1-6).
  - Secondary school (7–12).
  - At least some university (13+).
- **Earnings** from all occupations.

# What are the effects of changing age and educational compositions on male earnings at the aggregate level?

### **Aggregate-level data**

- Database is aggregated by census years, micro-regions, and age-education groups:
  - Brazil: 5 years \* 502 micro-regions \* 16 age-education groups.
  - Mexico: 3 years \* 2,456 municipalities \* 16 age-education groups.
- Cells with less than 25 people receiving income were excluded:
  - Brazil: 32,201 observations remained.
  - Mexico: 82,604 observations remained.
- Only male population: labor force participation is not driven by level of earnings, fertility decline, and changes in educational attainment.

#### Data setup

Year	Area	Age- education group G11–G44	Log of mean earnings log(Y <sub>git</sub> )	Distr. of male pop. P11–P44	P11	P12	P13	P14	 P44	Num. of obs.
1970	110006	15–24 years & 0–3 educ.	5.80	0.221	0.221	0	0	0	 0	2,016
1970	110006	15–24 years & 4–8 educ.	6.02	0.102	0	0.102	0	0	 0	927
1970	110006	15–24 years & 9–11 educ.	6.57	0.007	0	0	0.007	0	 0	62
1970	110006	15–24 years & 12+ educ.	7.58	0.001	0	0	0	0.001	 0	11
1970	110006	50–64 years & 12+ educ.	7.91	0.002	0	0	0		 0.002	15

#### **Fixed effects models**

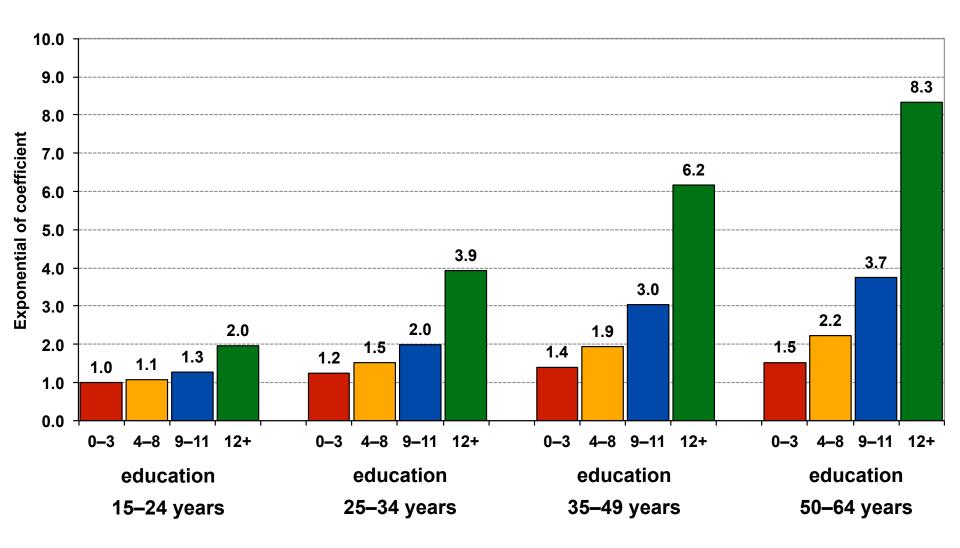
	Baseline model	Composition model
Dependent variable		
Logarithm of the mean real monthly earnings by age-education group, area, and time	log(Y <sub>git</sub> )	log(Y <sub>git</sub> )
Independent variables		
16 age-education indicators * time	(G <sub>11</sub> –G <sub>44</sub> ) * θ <sub>t</sub>	(G <sub>11</sub> –G <sub>44</sub> ) * θ <sub>t</sub>
Distribution of male population into 16 age- education groups * time		(P <sub>11</sub> –P <sub>44</sub> ) * θ <sub>t</sub>
Area-time fixed effects	$\alpha_{it}$	$\alpha_{it}$

# Estimating the impacts of relative group size on male earnings

- Baseline model:
  - Effects of age-education indicators ( $G_{11}$ - $G_{44}$ ).

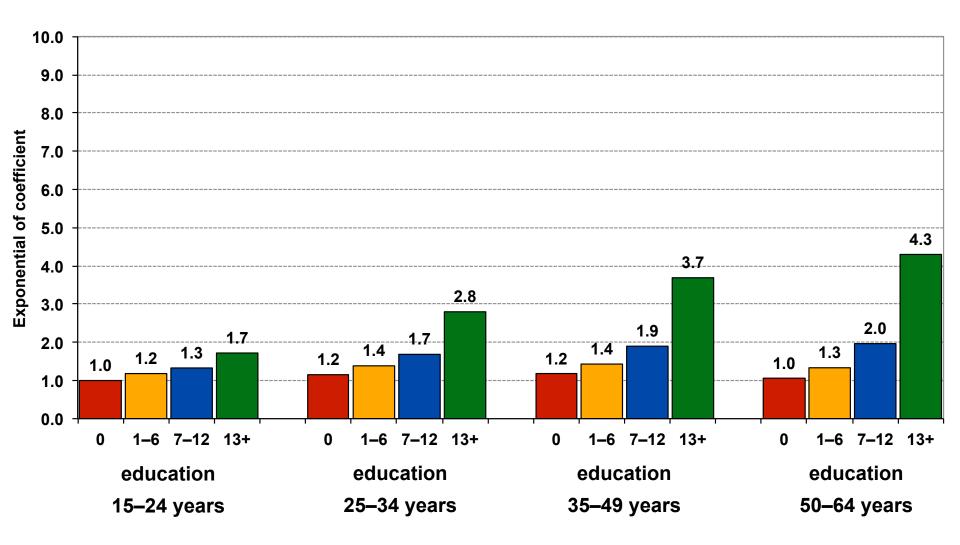
- Composition model:
  - Effects of age-education indicators  $(G_{11}-G_{44})$ .
  - Effects of age-education-group proportions ( $P_{11}-P_{44}$ ).

#### Effects of age-education indicators (G<sub>11</sub>–G<sub>44</sub>)<sup>18</sup> Baseline model, Brazil, 2010



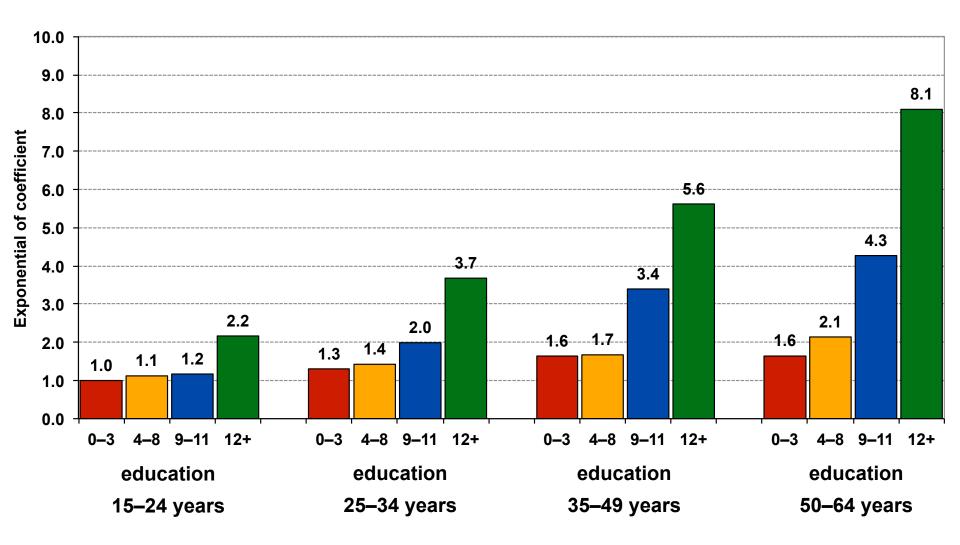
Source: 1970, 1980, 1991, 2000, and 2010 Brazilian Demographic Censuses.

#### Effects of age-education indicators $(G_{11}-G_{44})^{19}$ Baseline model, Mexico, 2010



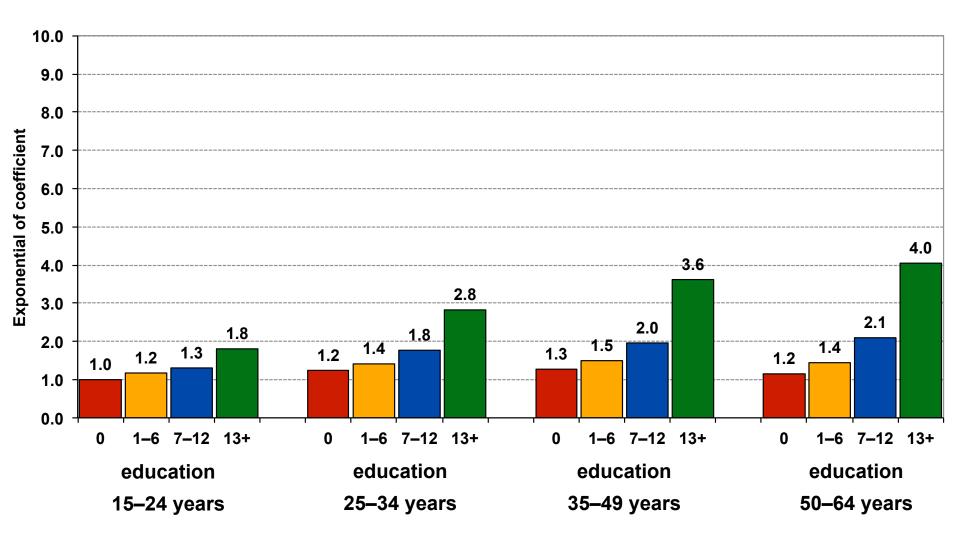
Source: 1990, 2000, and 2010 Mexican Demographic Censuses.

#### Effects of age-education indicators (G<sub>11</sub>–G<sub>44</sub>)<sup>20</sup> Composition model, Brazil, 2010



Source: 1970, 1980, 1991, 2000, and 2010 Brazilian Demographic Censuses.

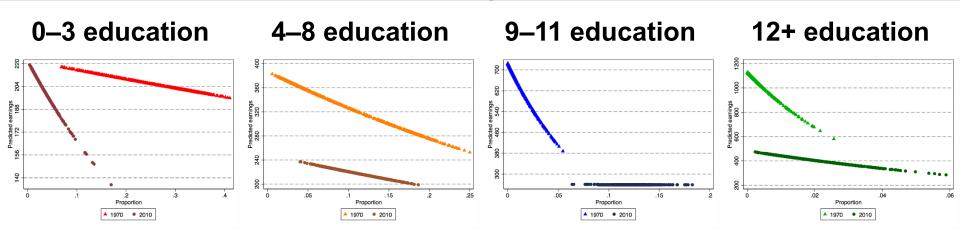
#### Effects of age-education indicators (G<sub>11</sub>–G<sub>44</sub>)<sup>21</sup> Composition model, Mexico, 2010



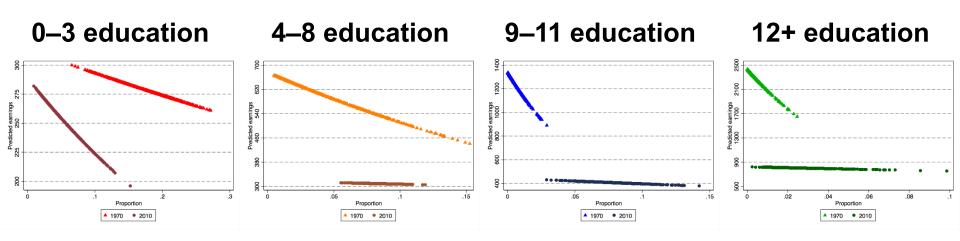
Source: 1990, 2000, and 2010 Mexican Demographic Censuses.

# Effects of group proportions in 502 areas $^{22}$ (P<sub>11</sub>–P<sub>24</sub>), Brazil, 1970 and 2010

15-24 years



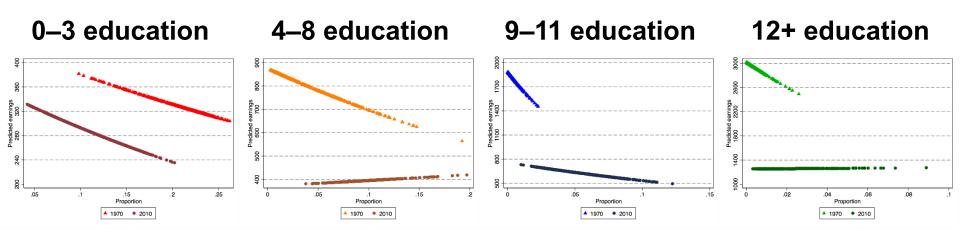
25–34 years



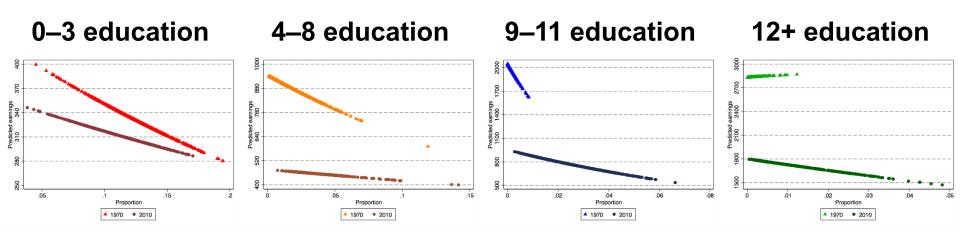
Source: 1970, 1980, 1991, 2000, and 2010 Brazilian Demographic Censuses.

# Effects of group proportions in 502 areas <sup>23</sup> (P<sub>31</sub>–P<sub>44</sub>), Brazil, 1970 and 2010

35-49 years



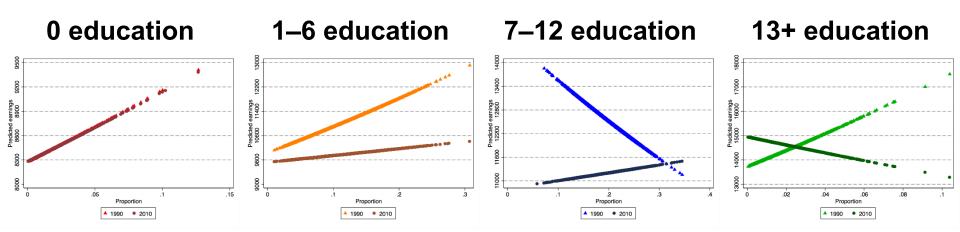
50–64 years



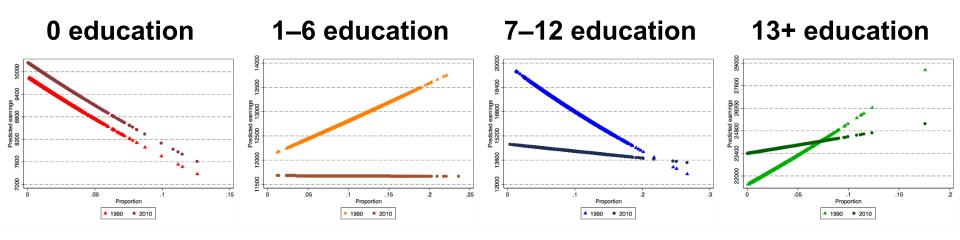
Source: 1970, 1980, 1991, 2000, and 2010 Brazilian Demographic Censuses.

### Effects of group proportions in 2,456 areas <sup>24</sup> (P<sub>11</sub>–P<sub>24</sub>), Mexico, 1990 and 2010

15-24 years



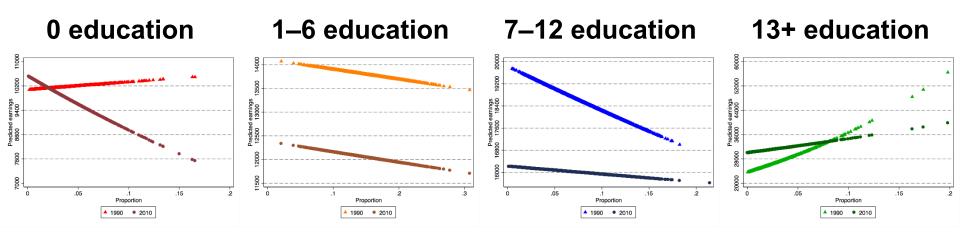
25–34 years



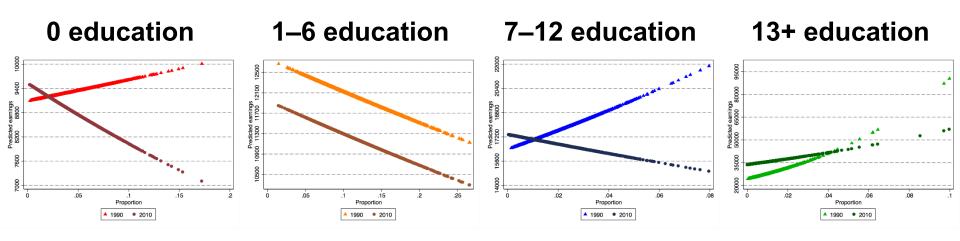
Source: 1990, 2000, and 2010 Mexican Demographic Censuses.

# Effects of group proportions in 2,456 areas <sup>25</sup> (P<sub>31</sub>–P<sub>44</sub>), Mexico, 1990 and 2010

#### 35-49 years



50–64 years



Source: 1990, 2000, and 2010 Mexican Demographic Censuses.

#### **Robustness checks**

- Extra models included as independent variables:
  - Cross effects.
  - Population size of micro-regions.
  - Female workers.
  - Internal migration.
- Original impacts of distribution of males into age-education groups (P<sub>11</sub>–P<sub>44</sub>) remained negative and significant.

# How does the concentration of skilled workers affect the private and social returns to education at the individual level?

#### Individual data analysis

- Males in the labor force: working or looking for a job.
- Two sets of analysis: aged 15–60 (shown here) and 30–50 (prime age adults).
- Education: (1) less than primary; (2) primary completed; (3) secondary completed; (4) university completed.
- Dependent variable: logarithm of individual earnings.
- Variables of interest:
  - Years of schooling: measure private returns to education.
  - Concentration of educated workers (undergraduates): measures social returns to education.

#### **Estimation procedure**

 The spatial distribution of the more educated population is associated with unobserved factors which in turn can be correlated with the level of income (Moretti 2004a, 2004b):

– The level of education becomes endogenous.

 The alternative needed to solve this problem is to use instruments to estimate the stock of skilled labor in localities:

- Lagged explanatory variables.

 The models are estimated for the overall population, as well as by income quantiles (25th, 50th, 75th).

#### Two-stage least squares model

#### Estimating the proportion of skilled workers by area:

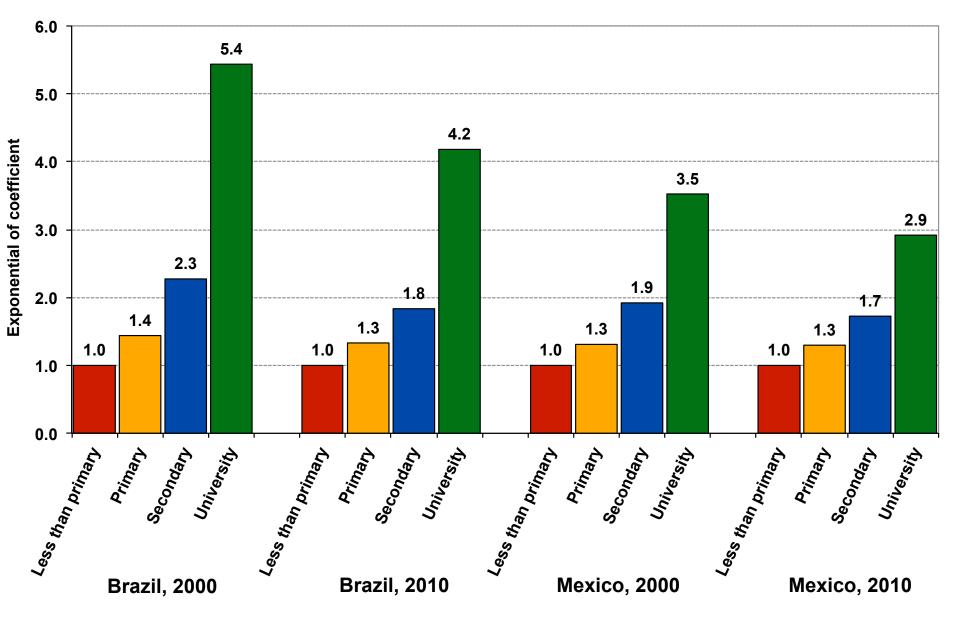
 $P_{(t)} = \beta_0 + \beta_1 L_{1(t-n)} + \beta_2 L_{2(t-n)} + \beta_3 L_{3(t-n)} + e$ 

- $P_{(t)}$ : proportion of workers with high educational level (undergraduates) in time *t* for each investigated area.
- $L_{1(t-n)}$ : enrollment rate in high school in the previous census.
- $L_{2(t-n)}$ : young-age-dependency ratio in the previous census.
- $L_{3(t-n)}$ : local average earnings in the previous census.
- Estimating private and social returns to education:

$$log(Y) = \beta_0 + \beta_n X_n + e$$

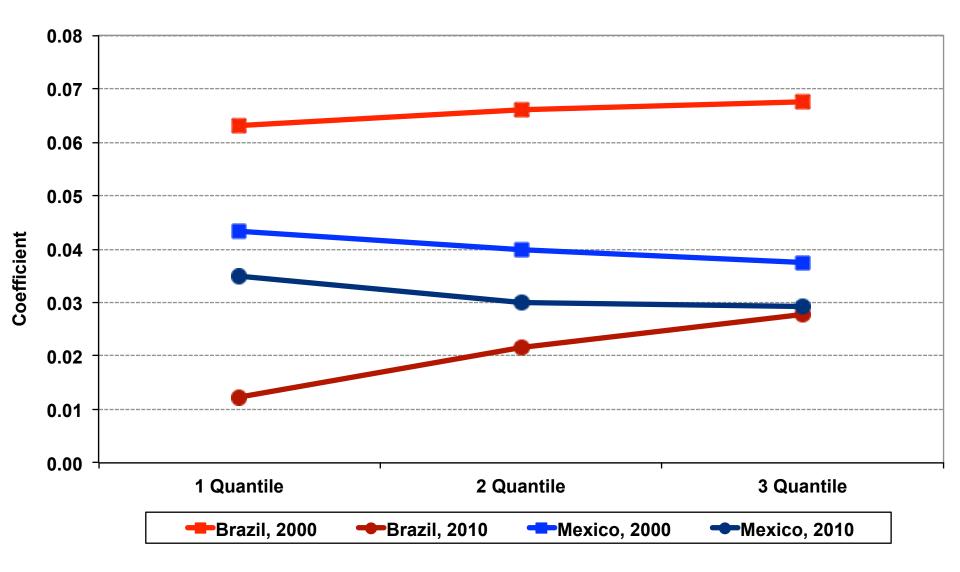
- log(Y): logarithm of individual earnings.
- $X_n$ : years of schooling; proportion of undergraduates; age; migrant; urbanization rate; unemployment rate; region.

#### Private returns to education, 2000 and 2010



Source: Brazilian and Mexican Demographic Censuses.

# Social returns to education by income quantile, 2000 and 2010



Source: Brazilian and Mexican Demographic Censuses.

#### **Cohort size**

- In line with previous studies: larger cohort-education size generally depresses earnings.
- Men with low education: these groups are decreasing over time, but their earnings are not increasing.
- Secondary school: groups are increasing over time and experiencing negative impacts on earnings.
- Time: effects are becoming less negative over the years.
  - However, effects for secondary-school groups are more negative in Brazil in 2010, compared to 2000.

### **Concentration of human capital**

- Positive effects of the concentration of skilled workers on earnings:
  - Decrease for **Mexico** along the income distribution.
  - Increase for **Brazil** along the income distribution.
- Time: in both countries, effects decreased from 2000 to 2010, which might be related to educational progress.
- Income inequality: might increase in Brazil, because the concentration of human capital is more beneficial to the highest income quantile than lower quantiles:
  - In the U.S., concentration of human capital has been more beneficial to lower income quantiles.

### Implications

#### Reduction in income inequality:

- More better-educated men: negative impacts reduced differentials in relation to lower-educated men.
- Fewer younger men: smaller negative impacts on their earnings prevented greater disparities in relation to older men.

#### Increase in income inequality:

 Concentration of human capital: higher positive impacts on the highest quantile might be a consequence of educational improvement in certain localities.

#### – Public policies:

- Demand for education: improve educational levels in areas that still have large proportions of people with low-education.
- Decentralize college education: recent Brazilian policies might generate positive impacts for the whole country.

#### **Research agenda**

- Other countries (IPUMS-International): India, Indonesia, South Africa, Chile, and Argentina.
- Models by sectors: estimate impacts of composition on earnings of workers with:
  - Formal employment.
  - Informal employment.
  - Self employment.
- Occupational profile and labor force participation: analyze how adults and elderly labor supply are changing over time and across regions in Mexico and Brazil.