

Educational Inequality and Intergenerational Mobility in Latin America: A New Database*

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Abstract. Causes and consequences of the intergenerational persistence of inequality are a topic of great interest among different fields in economics. However, issues of data availability have restricted a broader and cross-national perspective on the topic so far. Based on rich sets of harmonized household survey data, we contribute to fill this gap providing time series for several indexes of relative and absolute intergenerational education mobility for 18 Latin American countries over 50 years. In this paper, we introduce this new database and describe the observed patterns. We find that on average intergenerational mobility has been rising in Latin America. This pattern seems to be driven by high upward mobility of children from low-educated families, while there is substantial immobility at the top of the distribution. Significant cross-country differences can be observed which are associated with the degree of income inequality, poverty, economic growth and public educational expenditures.

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Panel data set can be downloaded at <https://sites.google.com/site/mobilitylatam/>

1 Introduction

One of the oldest and most argued topics in economics are the causes and consequences of economic inequality. On the one side, the high levels of inequality experienced by most developed and developing countries during the last century, attracted special attention by researchers and policy makers. On the other side, differing views on the beneficial, detrimental or neutral impact of economic inequality coexist and it seems still difficult to make clear causal statements on the way in which inequality affects economic performance and vice versa. For instance, influential theoretical models and empirical analyses on the relationship between inequality and growth yield opposite results so far ([Banerjee and Duflo, 2003](#)).¹ Hence, scholars argued that answers to questions related to inequality require taking into account that the observed level of income inequality displays both, the rewards obtained by individuals for their efforts, as well as the returns to given circumstances that cannot be influenced by them, like the socioeconomic status of their parents ([Roemer, 2000](#)).² It has, therefore, been suggested to extend distributional analysis to the degree of intergenerational mobility in a society ([Corak, 2013](#)).

The evaluation of intergenerational mobility allows to address one important concern: for a given level of inequality, how likely is it that families persist at the top or bottom of the distribution over the course of time. Analyzing the subject across multiple countries and periods further helps to understand which factors are associated with this likelihood. However, comparing estimates for different countries deriving from different studies raises the question on whether the uncovered cross-country differences are real or due to differences in data and measurement issues ([Solon, 2002](#)). Therefore, studying the subject in a harmonized framework is necessary to deepen our knowledge on the factors associated with the intergenerational transmission of socioeconomic status.

¹See [Furman and Stiglitz \(1998\)](#) for an overview of the consequences of inequality for growth. Recently, [Neves et al. \(2016\)](#) review the empirical literature on the inequality-growth nexus and perform a Meta-Analysis. Their results point at non-significant results on average with a high amount of heterogeneity in effect sizes across countries.

²For instance, [Marrero and Rodríguez \(2013\)](#) show that across US states inequality of opportunity has a negative impact on growth, while the impact of income inequality based on merit and effort is positive.

Furthermore, while large data sets with multiple and comparable measures of economic inequality and even historical time series are available for a multitude of countries, the same does not apply to intergenerational mobility. The subject has been extensively analyzed within countries, for instance recently for the US (Chetty et al., 2014b,a) and India (Azam and Bhatt, 2015), but research on this topic still suffers from the lack of comparable estimates across multiple countries and over long time spans. Our study (and the associated database) contributes to fill this gap by estimating trends of relative and absolute intergenerational mobility of educational attainment in Latin America using novel sets of harmonized household survey data.

Our estimates constitute a panel of comparable summary indicators of intergenerational mobility for 18 countries over more than 50 subsequent cohorts that we make available to be used in future research. The aim of the present study is to introduce this new data set giving a comprehensive analysis of the observed trends of intergenerational mobility in Latin America, as well as their association with macroeconomic and institutional characteristics. This way, our study extends and enriches in several dimensions the existing evidence on intergenerational mobility in Latin America, recently reviewed by Torche (2014), and the cross-country study on intergenerational mobility by Hertz et al. (2007) that includes estimates for seven Latin American countries. First, including more countries and a longer time span in a harmonized framework. Second, providing preciser estimates that rely on several survey waves and higher numbers of observations. Third, obtaining estimates from two independent sources for nine of the 18 countries in our sample. Fourth, computing several indexes that fulfill different axioms and measure different dimensions of relative and absolute mobility. Fifth, computing estimates for father-son and mother-daughter pairs, as well as for the degree of assortative mating. Sixth, designing the resulting panel data appositely to be used in future research.

The paper is structured as follows: Section 2 describes the data sources and harmonization procedure used to obtain our estimates. Section 3 explains the applied methodologies. Section 4 comprises all our results: First, we describe the uncovered cross-country patterns, trends, heterogeneity by gender, and degrees of assortative mating. Then, we show the association of our

intergenerational mobility estimates with economic performance and institutional characteristics. Section 5 concludes.

2 Data

2.1 Description of Data Sources

The sources of information used to obtain our estimates derive from two sets of harmonized household survey data. The selection criteria to include a survey in our analysis is the availability of retrospective information on the parental educational background of adult individuals. To avoid co-residency bias in our estimates, we focus on surveys that include retrospective questions about parental education in the questionnaire, and did not include those where information on parental characteristics might be retrieved because parents and children reside in the same household.³

The first harmonized survey data set used in this study derives from the annual opinion survey *Latinobarómetro*. *Latinobarómetro* records individual and household characteristics of a nationally representative sample of adult respondents in 18 Latin American countries since 1995, including questions about own and parental education since 1998.⁴ The survey comprises every year a sample of 1000 to 1200 individuals per country, representing more than 600 million inhabitants. It is carried out by local firms under technical supervision of the *Latinobarómetro* Corporation, a private non-profit organization based in Santiago (Chile).⁵ For the present study, we use the survey waves that include retrospective questions on parental education (1998 to 2015). The second data set is retrieved through an ex-post harmonization of selected national household surveys that are mainly conducted by national statistical offices. All estimates with both data sets (henceforth

³For a recent analysis of co-residency bias in intergenerational mobility estimates, see [Emran et al. \(2016\)](#).

⁴The Dominican Republic was included for the first time in 2004. The representativeness of the survey has varied over time reaching 100 % of the total population in all countries around the year 2000.

⁵The study receives financing from Latin American and non-Latin American governments, the private sector, and international organizations. Among others: IADB (Inter-American Development Bank), UNDP (United Nations Development Program), AECI (Agencia Española de Cooperación Internacional), SIDA (Swedish International Development Cooperation Agency), CIDA (Canadian International Development Agency), CAF (Corporación Andina de Fomento), OAS (Organization of American States), United States Office of Research, IDEA International, UK Data Archive.

Latinobarómetro and National Household Surveys) are obtained weighting each observation by the inverse probability of selection, normalizing the weights over the different survey waves. All the surveys used in our analysis are listed and described in Appendix A (Supplemental Material).

Latinobarómetro has the great advantage to be harmonized ex-ante and is developed particularly to be suitable for cross-country studies. The other household surveys are not uniform across Latin American countries. Therefore, we made all possible efforts to make statistics comparable across countries and over time by using similar definitions of variables in each country and survey year, and by applying consistent methods of processing the data. In particular, the inclusion of retrospective questions is not a universal characteristic that applies to all household surveys. Thus, while with the sample retrieved from Latinobarómetro we estimated the indexes for 18 countries, with the National Household Surveys estimates for 9 countries could be obtained. The advantage of most surveys included in the latter is the substantially higher number of observations. Furthermore, the survey structure allows to estimate father-son, and mother-daughter associations while Latinobarómetro only includes information on the parent with the highest educational degree.

2.2 Restriction criteria

We draw the same sample in each country and survey. The sample comprises individuals born between 1940 and 1990 who were at least 23 years old when interviewed in the household survey. The age limit ensures that individuals have a higher likelihood to have completed their educational career avoiding biased estimates. Since parental education is retrieved through retrospective questions, it does not matter whether the individual and her parents reside together in the same household or not to be included in our sample. The main restriction criteria is therefore the availability of information on own and parental education. Our final samples over all countries and cohorts comprise 211,401 individuals in the Latinobarómetro survey and 1,078,445 individuals in the National Household Surveys.

The amount of missing information about parental educational background is relatively small in Latinobarómetro – on average about 12 % of all individuals in the survey with available information

on own education. In some of the National Household Surveys the amount is much higher, ranging from 12 % in Guatemala to 61 % in Peru and 84 % in Brazil. In order to proof if selectivity issues bias our intergenerational mobility estimates, we compare the average years of education of all individuals in the household survey with the sample of individuals with available information on parental educational background. Differences are negligible in both data sets, counting at most 0.3 years of schooling, and in most countries not statistically significant. Furthermore, no clear pattern hints at a specific direction of a possible selectivity bias (e.g. in Peru, the average of the sample used to compute our estimates is 0.2 years lower than the unrestricted sample, while in Brazil the mean of our sample is 0.3 years higher).

2.3 Measurement of educational attainment

In Latinobarómetro the recorded information on parental education regards only the parent with highest education among the two. In the National Household Surveys, mostly the education of both, mother and father, is available. Also in this case, to obtain our baseline estimates we use the parent with the highest educational degree, as usually done in the literature ([Black and Devereux, 2011](#)).

In order to improve the comparability of the completed years of education, which is our main result variable, in processing the National Household Surveys we follow the same coding used by Latinobarómetro. That is, we truncate the years of education at the university level since the degree of heterogeneity is greater at that level. In this way, completed years of education range from 0 to 15. Furthermore, Latinobarómetro has the same variable measuring the education of individuals and their parents. Most other surveys record years of formal education for individuals who are interviewed, but not that precisely for their parents. In those cases we impute the years of education which are necessary to complete the obtained degree and follow the same scheme used in the Latinobarómetro survey.⁶

⁶Detailed information on the codification of educational attainment for parents and children in each country are contained in the Supplemental Material.

Figure 1 shows the mean and coefficient of variation of completed years of education in our samples, comparing the statistics obtained from Latinobarómetro and the National Household Surveys. The cohorts always refers to the cohorts of the children's generation. It is evident that in most countries the two harmonized survey sets yield very similar statistics in trends and levels. Throughout the cohorts, educational attainment of individuals in Latin America increased steadily, while there is certain heterogeneity in the levels of schooling among countries. In the youngest cohort, we find Guatemala, Honduras, and Nicaragua on the one end with around six years of education on average, and Argentina, Chile and Colombia on the other with around 12 years.

In order to give an idea on how educational attainment are related to economic well-being, Figure 2 shows the mean income levels for six broad educational categories and the returns to education – measured by the ratio of incomes achieved by high and low educated people – for two different cohorts in each Latin American country. This analysis helps to read our intergenerational education mobility estimates and put the results in the right context.⁷ We see, that although substantial differences between countries exist, higher educational degrees are clearly associated with higher level of income. Furthermore, despite the educational expansions experienced in all countries, returns to education are rather similar for people of different ages. Thus, apart of the intrinsic value of educational mobility itself, our measures are meaningful indicators for intergenerational mobility of (material) well-being.

3 Estimated Mobility Indexes

Pioneering works by Becker and Tomes (1979) and Solon (1992) conceptualize the mechanisms and transmission channels that explain the observed degree of persistence between the economic outcomes of parents and children. However, especially in cross country comparisons, different indexes measuring intergenerational mobility may yield very different pictures. Researchers should

⁷As shown by Blanden (2013), there is a meaningful correlation between estimates of intergenerational income mobility and educational mobility across countries.

therefore adopt the measurement which fulfill the needs of the dimension they have in mind to analyze and the questions they want to answer.⁸

In the context of educational mobility, some questions might need absolute mobility measures, for instance to capture educational expansions (structural mobility), while others might need to neglect this dimension and focus on positional changes of families within the distribution (exchange mobility). In this study, and with the creation of the associated database, we try to offer an exhaustive panorama of absolute and relative indexes and to show the overall picture of intergenerational mobility in Latin America from its different angles.

Future research using our estimates should choose the indexes which fit the requirements of the research question regarding two key aspects: i) what is the intuition behind the phenomena that has to be analyzed, and ii) which axioms have to be fulfilled. In what follows, we describe the computed indexes. The key variables are always referring to educational outcomes of parents (y^p) and children (y^c) measured either in completed years of education or the obtainment of a certain educational degree. The indexes are estimated for each cohort j and country k separately.⁹

3.1 Slope coefficient and intergenerational correlations

The most widely used mobility index in the intergenerational mobility literature is the slope coefficient from a linear regression of children's on parents' outcomes.¹⁰ Here, we regress the years of education of the child from family i belonging to cohort j in country k on the years of education of his parent with the highest educational attainment among the two:

$$y_{ijk}^c = \alpha_{jk} + \beta_{jk} \cdot y_{ijk}^p + \gamma_{jk} X_{ijk} + \varepsilon_{ijk}. \quad (1)$$

⁸For conceptual and methodological reviews on intergenerational mobility, see [Black and Devereux \(2011\)](#); [Jäntti and Jenkins \(2015\)](#); [Piketty \(2000\)](#).

⁹[Neidhöfer \(2016\)](#) develops a method to transform the educational outcomes of parents and children in a way that makes them more appropriate as a proxy measure for socioeconomic status and more comparable across time (see also [Neidhöfer and Stockhausen, 2016](#)). However, since the analysis is performed for each cohort separately, the correction is not necessary in this case. Proper methods are applied to standardize the estimated coefficients ex-post, as explained below.

¹⁰The specification of the model displayed here simplifies to one child per family.

In this equation, α is a constant, X a vector of control variables for age and sex and ε the error term. The slope coefficient can furthermore be standardized to take differences in the distributions of children's and parents' outcomes into account:

$$r_{jk} = \beta_{jk} \frac{\sigma_{jk}^p}{\sigma_{jk}^c}. \quad (2)$$

If no control variables are included in the regression, the standardization yields an index equal to Pearson's correlation coefficient.

β and r are measures for positional mobility that capture both dimensions, structural mobility as well as exchange mobility, and reflect the degree of regression to the population mean between two generations. Its wider use in the literature has the advantage of comparability between these and other estimates for the same or other countries. Hereby, r "corrects" β by the changes in inequality in the marginal distributions of the outcome of interest. Scholars still argue about which of the two is more suitable for cross-country (and cross-cohort) comparisons (see [Jäntti and Jenkins, 2015](#)). Therefore, it seems important to report both.

An index which fully controls for the marginal distributions – and not only for the changes in inequality – and captures the pure positional change aspect of mobility, is Spearman's rank correlation coefficient:

$$\rho_{jk} = \frac{\text{cov}(\text{rank}_{jk}^c, \text{rank}_{jk}^p)}{\sigma_{jk,rank}^c \sigma_{jk,rank}^p}. \quad (3)$$

Whether these corrections are necessary or not depends on the research question. As stated before, the intergenerational transmission of inequality could be an important dimension which would get lost if measuring mobility by (2) and (3). However, if exchange mobility is the only important aspect to be accounted for, (1) might not be the suitable index to rely on.

The outcome which is most likely available for two subsequent generations and is furthermore comparable across countries are educational attainment measured in completed years of education.

Thus, these indexes have one important feature in common: All give a broad and intuitive picture of the overall educational persistence experienced by a certain cohort in a given country.¹¹

3.2 Transition probabilities

Another insightful measure in terms of intergenerational mobility is the probability of children facing different circumstances, measured by parental educational background, to afford a certain minimum level of education. We compute two different indicators:

The *probability of bottom upward mobility*

$$BUM_{jk} = Prob(y_{ijk}^c \geq s | y_{ijk}^p < s), \quad (4)$$

and the *probability of upper class persistence*

$$UCP_{jk} = Prob(y_{ijk}^c \geq s | y_{ijk}^p \geq s). \quad (5)$$

The indicators yield the probabilities of children to achieve at least a secondary educational degree – measured by a Probit regression on a dummy variable being one if the individual completed at least secondary education (s) and zero otherwise – conditional on their parents' education. Parent's education is hereby measured by two different types: i) low parental education, i.e. less than completed secondary education. ii) high parental education, i.e. at least a secondary school degree. In terms of social mobility and equality of opportunity these probabilities measure upward mobility for people at the bottom of the distribution and class persistence at the top, respectively.

¹¹Equation (1) might be also estimated on the logarithm of the outcome of interest, e.g. years of education. In this case, the slope coefficient is an elasticity measuring marginal changes in children's education associated with marginal changes in their parent's education. The intuitive difference between the educational persistence explained above and the intergenerational education elasticity (not discussed in this paper but included in the database) lies mainly in the functional form assumed to underlie the intergenerational transmission of education and social status.

3.3 Absolute and directional mobility

The measures described above cover the relative and absolute dimensions of intergenerational mobility understood as movement of families within the distribution over time. However, they do not give comparable information about the size of those movements. Two further indexes – initially developed by [Fields \(1996\)](#) and mostly applied to measure individual income movements in an intragenerational context – are therefore computed which measure the per capita movements in years of education:

$$M1_{jk} = \frac{1}{N_{jk}} \sum_{i=1}^{N_{jk}} |y_{ijk}^c - y_{ijk}^p|. \quad (6)$$

$$M2_{jk} = \frac{1}{N_{jk}} \sum_{i=1}^{N_{jk}} (y_{ijk}^c - y_{ijk}^p), \quad (7)$$

$M1$ shows the average difference between the two generations within the same families, regardless of the direction of the change. Upward and downward movements are summed up to one summary measure. In contrast, $M2$ measures the average directional change between two generations. High values of $M2$ can for example be a sign for educational expansions. Together, $M1$ and $M2$ give also insightful information on the degree of downward movements: The smaller is the difference between the two, the lower is the amount, or average degree, of downward mobility.

4 Results: Intergenerational Mobility in Latin America

4.1 Cross-Country Patterns

Before reporting the intergenerational mobility trends through the summary measures described in Section 3, we describe the cross-country differences in mobility patterns for the entire sample. First, [Figure 3](#) illustrates absolute (or structural) mobility patterns, and, then, [Figure 4](#) illustrates relative (or exchange) mobility; both using Latinobarómetro as data source. [Tables 1, 2 and 3](#) show

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descriptive statistics of the summary measures described in Section 3 for each country and the Latin American average using both data sources.

Figure 3 ranks countries in Latin America according to the percentage of people who have higher education than their parents, measured in completed years of schooling. We see that more than 50 % of people born between 1940 and 1990 in all countries in the region have higher educational attainment than their parents. Venezuela and Paraguay lead the group of countries with high absolute mobility, while Guatemala, Nicaragua and Honduras are at the bottom end of the ranking. Although this evidence is illustrative of the differences between countries in terms of mobility, it is far from complete because it takes no account of the position of individuals in the distribution and the size of the change between generations.

Figure 4 is more informative about the movement of families within the distribution. In the upper part, a transition matrix for Latin America is displayed. Here, individuals and their parents are ranked regarding to their relative educational position, measured in standard deviations from the country average years of education, and ordered in three different classes: high, middle and low level of education. The cells of the transition matrix contain the percentage of individuals in the children's generation associated with the respective parental educational class. Complete intergenerational mobility is displayed by equal entries in each cell of a transition matrix. As has been argued in past, under certain circumstances complete mobility can be understood as equality of opportunity.¹²

We see that the Latin American reality is far away from complete mobility. Focusing on three most meaningful cells of the transition matrix – the ones that display persistence at the top and at the bottom of the distribution, as well as the degree of bottom-up mobility – Latin America shows up to be a region with low intergenerational mobility on average. Almost 60 % of children with high and low education, respectively, have parents in the same educational class. Moreover, only 14 % of the individuals in the high education class descend from low-education families. The lower

¹²For an exhaustive discussion of conceptual differences between intergenerational mobility and equality of opportunity, see [Roemer \(2004\)](#).

part of Figure 4 ranks the countries by this last indicator for bottom-up mobility. We see that the share ranges from less than 10 % in Chile to about 20 % in Nicaragua and Dominican Republic.¹³

It is noteworthy that depending on the adopted concept of mobility (relative or absolute) the country ranking is very different. Particularly striking is, for instance, Nicaragua that is one of the countries with the highest relative mobility and the lowest absolute mobility at the same time. What explains this seemingly controversial finding is that Nicaragua is one of the countries with the lowest and most unequally distributed educational attainments on average. Hence, while the opportunities of children from low educated families to improve their educational level are high, the chances that this improvement translates into a considerable jump within the distribution are quite modest. This finding confirms the importance of i) evaluate intergenerational mobility adopting multiple measures and ii) to measure the mobility of people born in different year spans separately.

4.2 Trends

Figures 5, 7 and 9 show the trends and geography of intergenerational mobility in Latin America measured by the seven indexes explained in Section 3 with the Latinobarómetro survey. Figures 6, 8 and 10 show the corresponding averages for the nine countries where we have National Household Surveys available to perform the analysis. Since the trends and levels obtained with the National Household Surveys basically mirror the results obtained with Latinobarómetro for all the estimated indexes, we will restrict the descriptive analysis in this section mainly to the results obtained with Latinobarómetro. Charts for each country with both surveys are included in Appendix C (Supplemental Material).

Figure 5 and 6 show intergenerational mobility measured by the regression coefficient (β), the standardized coefficient (r) and the Spearman's rank correlation coefficient (ρ). Aggregate results for Latin America are constructed as the unweighted average of the 18 or 9 countries analyzed,

¹³To give a benchmark for these estimates, we compute transition matrices for the US and Germany using the same sample restriction criteria and comparable household surveys (PSID and SOEP, respectively). It turns out, that in these two countries persistence at the bottom is higher than the Latin American average (USA 61.5 %, Germany 56.5 %). In contrast, persistence at the top is lower (USA 51.2 %, Germany 55.8 %) and bottom-up mobility higher (USA 21.5 %, Germany 17.8 %) than in most Latin American countries.

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depending on whether we use Latinobarómetro or National Household Surveys. β changes substantially and significantly over the observed period. For people born in the forties, an additional year of parental education is associated with an average increase of about 0.6 years of education, while for people born in the eighties the same measure is around 0.4.¹⁴ The map shows that this increase was recorded in almost all Latin American countries. In contrast, r and ρ are relatively stable around 0.5 over the entire period. This shows that the type of mobility experienced in Latin America has mainly been structural. However, in the two countries where the raise in intergenerational mobility has been the strongest, Dominican Republic and Venezuela, both, structural as well as exchange mobility increased significantly. Guatemala and Honduras are the only countries where structural as well as exchange mobility did not rise over the observation period.

Figure 7 and 8 illustrate the level and differences across cohorts of the probability of upward mobility for people at the bottom of the distribution, as well as the probability of class persistence at the top. On average the predicted probability of upper class persistence is high and oscillates around 0.7. In contrast, the predicted probability of individuals born in the eighties from low-educated parents to attain a secondary school degree is more than twice as high as the same probability for individuals born in the forties. However, not all countries show the same pattern. Although in most of the countries bottom-up mobility increased – up to a 300 % increase in Brazil and Mexico – it is on low levels and almost unchanged over time in the Central American countries, like Guatemala, Honduras and Nicaragua.¹⁵ Very high bottom-up mobility rates in the youngest cohorts (higher than 0.5) are observed in Argentina, Mexico, Peru and Venezuela.

Figure 9 and 10 show absolute and directional mobility trends. These measures show the magnitude and pattern of the change between the educational attainment of parents and children on average. As is evident, since the outcome measure – completed years of education – is bounded, rising parental education also reduces the margins and possibilities for the children to experience an improvement. This fact explains the inverted U-shape pattern of the time series for these two

¹⁴The results for the older cohorts are consistent with past estimates, e.g. by [Hertz et al. \(2007\)](#).

¹⁵ The spatial dimension of this phenomenon is a striking finding that might deserve special attention in future studies.

indexes. In the sixties, the distance between parents' and children's education reaches a top and later decreases as the education of parents rises. Interestingly, the gap between M1 and M2 does not change significantly across cohorts, showing that downward mobility is almost stable around one year of schooling on average.

4.3 Heterogeneity by Gender and Assortative Mating

In this part of the analysis, we first disentangle our estimates by father-son and mother-daughter lineages. These estimates give an overview of how social, cultural or institutional factors may influence the educational mobility of men and women differently. For instance, families might dedicate more resources to the education of male offspring, either because the returns to sons' education are expected to be higher, or because of traditional gender roles. For this last reason, imitation might cause the educational attainments of children to be related more strongly to the education of the parent with the same sex (see e.g. [Schneebaum et al., 2015](#)).

Then, we relate our intergenerational mobility estimates to the degree of assortative mating, i.e. the likelihood of people with similar socioeconomic status to marry each other. This analysis is particularly interesting since there seems to be a fundamental interrelation between the two concepts which only few studies could empirically prove so far (e.g. [Chadwick and Solon, 2002](#); [Ermisch et al., 2006](#); [Guell et al., 2015](#)). We can perform this evaluation for nine countries where we have information on both, father's and mother's educational attainment.

As shown in [Figure 11](#), the estimates for father-son and mother-daughter pairs show the same trend and are rather similar in younger cohorts. Coinciding with the expansion of educational attainment among women, also the mobility of daughters rises considerably and approaches the mobility levels experienced by sons on average. Generally, the patterns confirm the picture of rising intergenerational mobility in Latin America driven by high upward mobility from the bottom and with substantial immobility at the top of the distribution.

Taking into account the high degree of assortative mating in Latin American countries, these findings are not particularly surprising: when the education of both parents is similar, the education

of only one of the two is a valid proxy for the education of the other. Our findings show that assortative mating in Latin America, measured by the correlation of father's and mother's educational attainment, is constantly high (around 0.7, with countries ranging between 0.6 and 0.8; see Figure 12). Interestingly, most countries show a slight but decreasing trend. Indeed, past research found an inverse relationship between assortative mating and intergenerational mobility (Guell et al., 2015).

We test the relationship between assortative mating and intergenerational mobility with our database, regressing the seven estimated mobility indexes on the estimated degree of spouse correlation controlling for cross country heterogeneity by fixed effects. As shown in Table 4, the degree of spouse correlation is positively and significantly associated with educational persistence (measured by the regression coefficient, the correlation coefficient and the rank correlation) and negatively with the index of bottom upward mobility. The relationship with the index for upper class persistence and the measures of directional and absolute mobility point at the same picture – higher spouse correlation associated with lower intergenerational mobility – but are not statistically significant. Hence, our findings confirm a clear association between assortative mating and intergenerational mobility.

4.4 Intergenerational Mobility, Institutions and Economic Performance

The aim of this part of the analysis is to show the association of intergenerational mobility with macroeconomic and institutional characteristics. We focus the first descriptive part of this analysis on the regression coefficient as indicator of intergenerational persistence because this indicator comprises both, structural as well as exchange mobility. In order to make use of all the available data, we take the average of our educational persistence estimates forming three broader cohorts (people born 1940-54, 1955-69 and 1970-84) and associate them with data at the country level averaged over three time periods (1990-99, 2000-09 and 2010-14), respectively. The criteria on how to match the two data sets is hereby completely determined by the time periods where information is available. Hence, these correlations surely cannot be interpreted as causal effects. Yet, they might be seen as a first step to understand potential underlying mechanisms.

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Figure 13 and Figure 14 show scatter-plots, linear fits and the related correlation coefficients. We find that higher degrees of intergenerational mobility are associated with: i) High levels of household income per capita and GDP per capita. ii) Lower levels of income inequality and poverty. iii) Lower returns to education, as measured by the ratio of hourly wages of people with high and low education. iv) Higher amounts of public expenditure in education and, in particular, the share of expenditure devoted to primary education. These findings confirm the predictions of influential theoretical models (Becker and Tomes, 1979) and the patterns uncovered in empirical findings in past.¹⁶

Finally, we regress a series of macroeconomic outcomes separately on our intergenerational mobility estimates controlling for country fixed effects. The association of our estimates and the macroeconomic outcomes is performed as explained above. Figure 15 shows the estimated standardized coefficients and their respective confidence interval. We see that the two mobility indicators that capture the structural mobility component, educational mobility (β) and bottom upward mobility (BUM), are positively associated with economic growth and well-being, and negatively with poverty and inequality. The standardized persistence estimates (r) have a qualitatively similar association with the above mentioned macroeconomic outcomes, which is however not statistically significant. A possible interpretation of these findings is that what positively influences economic performance is not the amount of exchange mobility – the raise of families which is necessarily accompanied by the fall of other families – but the opportunities of children from the lower bottom of the distribution to improve their human capital in comparison to their parents. Furthermore, since absolute ($M1$) and directional mobility ($M2$) – i.e. the magnitude of the change from one generation to the next – show no meaningful association, the strength of the structural mobility component seems to be less influential than the marginal improvement of human capital in itself. Last, the probability of upper class persistence (UCP) shows the same pattern of conditional correlation with economic performance as the BUM . This is mainly due to the fact that the two

¹⁶For instance, the negative relationship between inequality and intergenerational mobility has been shown to hold within the US (Chetty et al., 2014a) and China (Fan et al., 2015), as well as across and within Latin American countries (Neidhöfer, 2016). Güell et al. (2015) find that within Italy intergenerational mobility is positively correlated with economic performance.

measures are highly correlated: changes in the probability to attain a secondary education degree, like reforms increasing the length of compulsory education, are likely to somehow affect all individuals regardless of their parental background. Including both as independent variables in the regressions, the coefficients of *BUM* are significantly different from zero, while the coefficients of *UCP* are not. These preliminary analyses with our database open interesting avenues for future research.

5 Conclusions

In this paper we introduced a new panel data set of intergenerational mobility estimates for Latin America and provided a comprehensive descriptive analysis of the observed trends and patterns. We found that intergenerational mobility of educational attainment has been rising in Latin America, especially driven by the educational expansions of the last decades that particularly benefited children from the bottom of the distribution. In contrast, the educational persistence at the top of the distribution stayed constantly on high levels and did not change substantially. Furthermore, we found intergenerational mobility to be positively associated with economic growth and progressive public expenditure in education, and negatively associated with income inequality, poverty, returns to education and the degree of assortative mating. The positive relationship between intergenerational mobility and economic performance was also found in estimations controlling for cross-country heterogeneity by fixed effects.

The strength of our analysis is to provide highly comparable estimates of educational mobility for people born over a span of over 50 years in multiple countries. These estimates can be used in future to analyze the characteristics that influence or are influenced by the degree of intergenerational mobility of socioeconomic status. For instance, in the context of developing countries, key aspects are the intergenerational transmission of poverty, the impact of educational expansions and social programs on equality of opportunity, and the role played by institutions.

In our view, the availability of such a data set is useful for at least one important reason: equality of opportunity and social mobility seem to be common goals for policy makers, as well as among

egalitarians and utilitarians. Hence, our panel provides an essential tool for discussions and future research on the topic at the cross country and within country level.

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6 Tables

Table 1: Descriptive Statistics: Regression and Correlation Coefficients.

Panel A – Source: Latinobarometro, own estimates.								
	Regression coeff.				Correlation coeff.			
	Mean	C.V.	Min.	Max.	Mean	C.V.	Min.	Max.
Argentina	0.44	0.16	0.32	0.54	0.51	0.06	0.46	0.56
Bolivia	0.54	0.14	0.40	0.64	0.55	0.04	0.51	0.60
Brazil	0.56	0.21	0.38	0.74	0.50	0.08	0.44	0.59
Chile	0.49	0.11	0.42	0.56	0.62	0.10	0.54	0.79
Colombia	0.54	0.16	0.38	0.72	0.54	0.07	0.50	0.63
Costa Rica	0.41	0.12	0.34	0.49	0.42	0.07	0.36	0.47
Dominican Rep.	0.44	0.27	0.33	0.65	0.42	0.17	0.34	0.57
Ecuador	0.54	0.10	0.47	0.63	0.53	0.06	0.48	0.58
El Salvador	0.62	0.19	0.43	0.81	0.56	0.09	0.48	0.63
Guatemala	0.58	0.08	0.49	0.65	0.51	0.07	0.45	0.56
Honduras	0.54	0.09	0.44	0.63	0.47	0.10	0.40	0.54
Mexico	0.38	0.21	0.29	0.53	0.40	0.12	0.35	0.48
Nicaragua	0.43	0.14	0.32	0.56	0.42	0.11	0.36	0.50
Panama	0.49	0.12	0.42	0.59	0.51	0.06	0.43	0.56
Paraguay	0.55	0.14	0.40	0.70	0.52	0.08	0.43	0.60
Peru	0.51	0.20	0.39	0.70	0.56	0.05	0.51	0.64
Uruguay	0.48	0.12	0.41	0.58	0.49	0.06	0.42	0.53
Venezuela	0.39	0.21	0.31	0.60	0.42	0.11	0.36	0.52
Latin America	0.50	0.15	0.39	0.63	0.50	0.08	0.44	0.57

Panel B – Source: National Household Surveys, own estimates.								
	Regression coeff.				Correlation coeff.			
	Mean	C.V.	Min.	Max.	Mean	C.V.	Min.	Max.
Brazil	0.54	0.26	0.33	0.75	0.51	0.10	0.43	0.59
Chile	0.39	0.29	0.24	0.56	0.50	0.09	0.41	0.58
Colombia	0.59	0.20	0.38	0.77	0.51	0.09	0.44	0.62
Ecuador	0.59	0.13	0.49	0.69	0.58	0.05	0.54	0.63
Guatemala	0.80	0.11	0.64	0.91	0.62	0.05	0.58	0.66
Mexico	0.49	0.22	0.33	0.64	0.51	0.08	0.44	0.56
Nicaragua	0.62	0.19	0.47	0.76	0.51	0.12	0.42	0.58
Panama	0.54	0.18	0.42	0.72	0.58	0.07	0.54	0.66
Peru	0.56	0.30	0.32	0.80	0.54	0.11	0.45	0.64
Latin America	0.57	0.21	0.40	0.74	0.54	0.08	0.47	0.61

Table 2: Descriptive Statistics: Upper Class Persistence and Bottom Upward Mobility.

Panel A – Source: Latinobarometro, own estimates.								
	Upper class persistence				Bottom-Up Mobility			
	Mean	C.V.	Min.	Max.	Mean	C.V.	Min.	Max.
Argentina	0.84	0.06	0.71	0.91	0.38	0.25	0.21	0.58
Bolivia	0.81	0.09	0.69	0.90	0.26	0.43	0.12	0.46
Brazil	0.76	0.11	0.55	0.84	0.27	0.44	0.11	0.48
Chile	0.85	0.05	0.79	0.94	0.37	0.17	0.28	0.49
Colombia	0.78	0.09	0.65	0.88	0.28	0.36	0.11	0.42
Costa Rica	0.65	0.12	0.50	0.74	0.22	0.23	0.13	0.30
Dominican Rep.	0.52	0.24	0.32	0.71	0.25	0.34	0.10	0.37
Ecuador	0.78	0.15	0.54	0.88	0.31	0.36	0.12	0.43
El Salvador	0.81	0.11	0.61	0.90	0.19	0.35	0.08	0.28
Guatemala	0.67	0.11	0.57	0.77	0.14	0.26	0.09	0.20
Honduras	0.71	0.12	0.58	0.86	0.14	0.18	0.11	0.18
Mexico	0.63	0.20	0.45	0.91	0.36	0.42	0.15	0.66
Nicaragua	0.62	0.16	0.45	0.79	0.16	0.29	0.06	0.21
Panama	0.78	0.06	0.70	0.89	0.36	0.20	0.23	0.42
Paraguay	0.80	0.07	0.69	0.91	0.25	0.32	0.16	0.40
Peru	0.86	0.07	0.73	0.93	0.42	0.24	0.24	0.56
Uruguay	0.70	0.07	0.62	0.79	0.23	0.12	0.17	0.28
Venezuela	0.61	0.34	0.25	0.84	0.35	0.34	0.15	0.54
Latin America	0.73	0.12	0.58	0.85	0.27	0.29	0.15	0.40

Panel B – Source: National Household Surveys, own estimates.								
	Upper class persistence				Bottom-Up Mobility			
	Mean	C.V.	Min.	Max.	Mean	C.V.	Min.	Max.
Brazil	0.85	0.07	0.71	0.92	0.36	0.39	0.15	0.56
Chile	0.81	0.11	0.65	0.91	0.44	0.40	0.16	0.69
Colombia	0.82	0.08	0.71	0.90	0.34	0.42	0.12	0.55
Ecuador	0.74	0.12	0.50	0.82	0.21	0.40	0.06	0.30
Guatemala	0.78	0.09	0.62	0.84	0.11	0.37	0.04	0.17
Mexico	0.78	0.10	0.63	0.92	0.23	0.37	0.09	0.35
Nicaragua	0.57	0.29	0.29	0.79	0.12	0.36	0.05	0.17
Panama	0.77	0.05	0.70	0.83	0.29	0.26	0.15	0.38
Peru	0.88	0.03	0.82	0.92	0.41	0.27	0.19	0.57
Latin America	0.78	0.11	0.62	0.87	0.28	0.36	0.11	0.42

Table 3: Descriptive Statistics: Absolute and Directional Mobility.

Panel A – Source: Latinobarometro, own estimates.

	Absolute mobility (M1)				Directional mobility (M2)			
	Mean	C.V.	Min.	Max.	Mean	C.V.	Min.	Max.
Argentina	3.4	0.1	2.7	3.6	2.8	0.1	1.9	3.2
Bolivia	4.3	0.1	3.3	4.8	3.6	0.2	2.5	4.3
Brazil	4.0	0.1	2.9	4.5	3.3	0.2	2.2	3.9
Chile	3.4	0.1	2.7	3.9	2.8	0.2	1.8	3.2
Colombia	4.0	0.1	2.9	4.5	3.1	0.1	2.2	3.7
Costa Rica	3.9	0.1	3.5	4.5	2.8	0.3	1.6	3.8
Dominican Rep.	4.4	0.1	3.3	5.0	3.3	0.2	2.4	4.1
Ecuador	3.8	0.1	3.2	4.4	3.1	0.2	2.2	3.9
El Salvador	4.0	0.1	3.0	4.6	3.4	0.1	2.5	3.9
Guatemala	3.2	0.1	2.6	3.6	2.0	0.2	1.5	2.5
Honduras	3.5	0.1	3.2	3.9	2.7	0.2	2.0	3.3
Mexico	4.3	0.1	3.6	4.8	3.1	0.1	2.5	3.6
Nicaragua	3.9	0.1	2.8	4.7	2.7	0.2	1.7	3.6
Panama	4.2	0.1	3.4	4.8	3.5	0.2	2.1	4.3
Paraguay	3.8	0.1	3.4	4.3	3.2	0.1	2.8	4.0
Peru	4.1	0.1	3.5	4.6	3.3	0.2	2.5	4.0
Uruguay	3.2	0.1	2.6	3.6	2.3	0.3	1.3	2.9
Venezuela	4.4	0.1	3.7	5.2	3.8	0.2	2.7	4.5
Latin America	3.9	0.11	3.1	4.4	3.0	0.17	2.1	3.7

Panel B – Source: National Household Surveys, own estimates.

	Absolute mobility (M1)				Directional mobility (M2)			
	Mean	C.V.	Min.	Max.	Mean	C.V.	Min.	Max.
Brazil	4.2	0.2	2.9	5.1	3.3	0.2	1.9	4.2
Chile	4.0	0.1	3.1	4.6	3.3	0.2	2.2	4.0
Colombia	4.1	0.2	2.7	4.6	3.2	0.3	1.5	3.9
Ecuador	3.2	0.1	2.5	3.6	2.1	0.4	0.5	2.8
Guatemala	2.6	0.2	1.6	3.7	1.8	0.4	0.7	3.0
Mexico	4.5	0.1	3.4	5.2	3.8	0.2	2.7	4.8
Nicaragua	3.2	0.2	2.2	4.0	2.2	0.3	0.9	3.0
Panama	3.6	0.1	3.0	4.1	2.6	0.2	1.6	3.5
Peru	4.5	0.1	3.1	5.1	3.8	0.2	2.1	4.5
Latin America	3.77	0.15	2.74	4.43	2.91	0.26	1.55	3.73

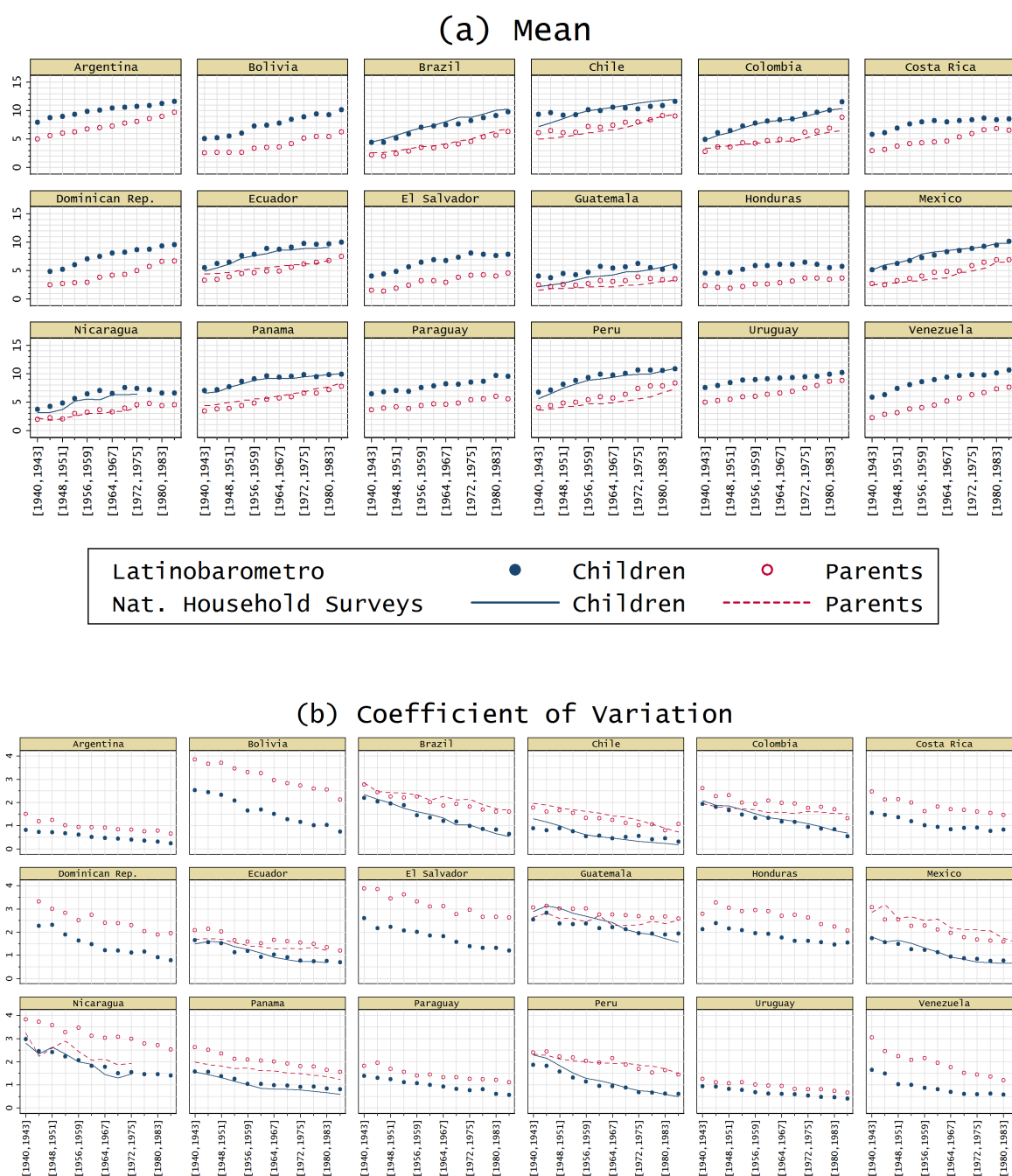
Table 4: Assortative mating and intergenerational mobility – Linear Regressions.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	β	ρ	r	BUM	UCP	$M1$	$M2$
Spouse correlation (parents)	1.516*** (0.3420)	0.638*** (0.1405)	0.295*** (0.1006)	-1.223*** (0.4169)	0.405 (0.3621)	-2.355 (1.9412)	-2.448 (2.5856)
Country F.E.	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	104	104	104	104	104	104	104

Notes: Table shows the coefficients of the computed spouse correlation index in linear regressions using the mobility indexes as dependent variable. All regressions include country dummies. Robust standard errors in parentheses. Statistical significance level * 0.1 ** 0.05 *** 0.01. *Source:* National Household Surveys 1982-2015, own estimates.

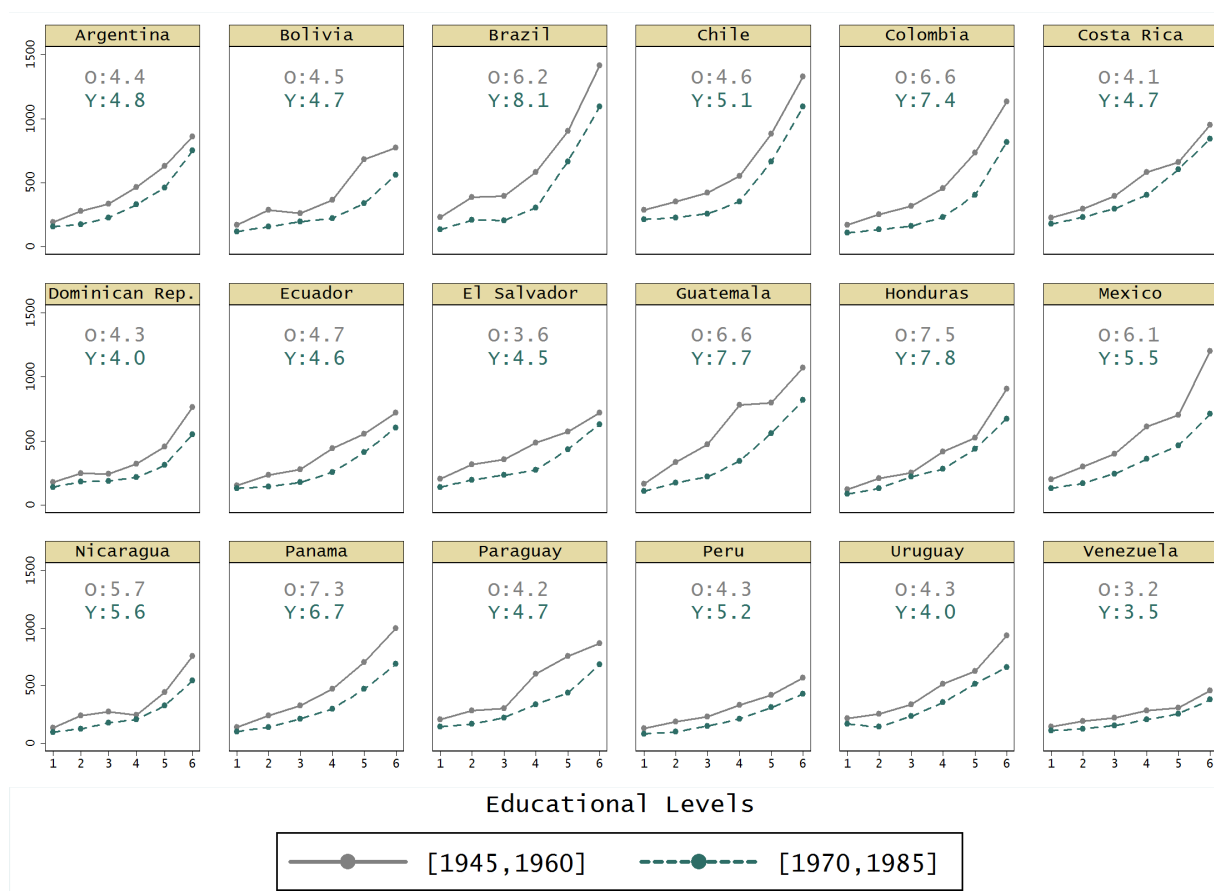
7 Figures

Figure 1: Completed years of education. Sample means and coefficients of variation by cohorts.



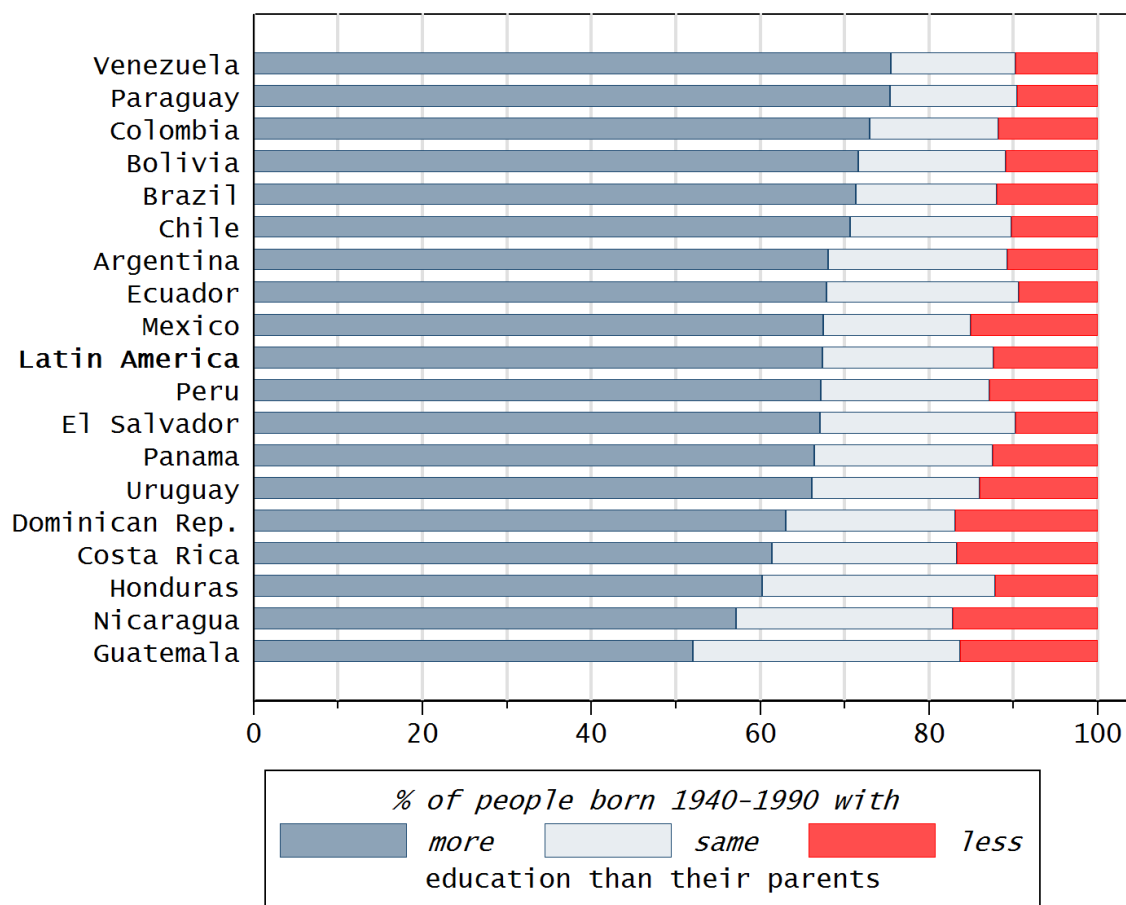
Source: Latinobarometro 1998-2015, National Household Surveys 1982-2015.

Figure 2: Education as indicator for well-being: average income by educational level.



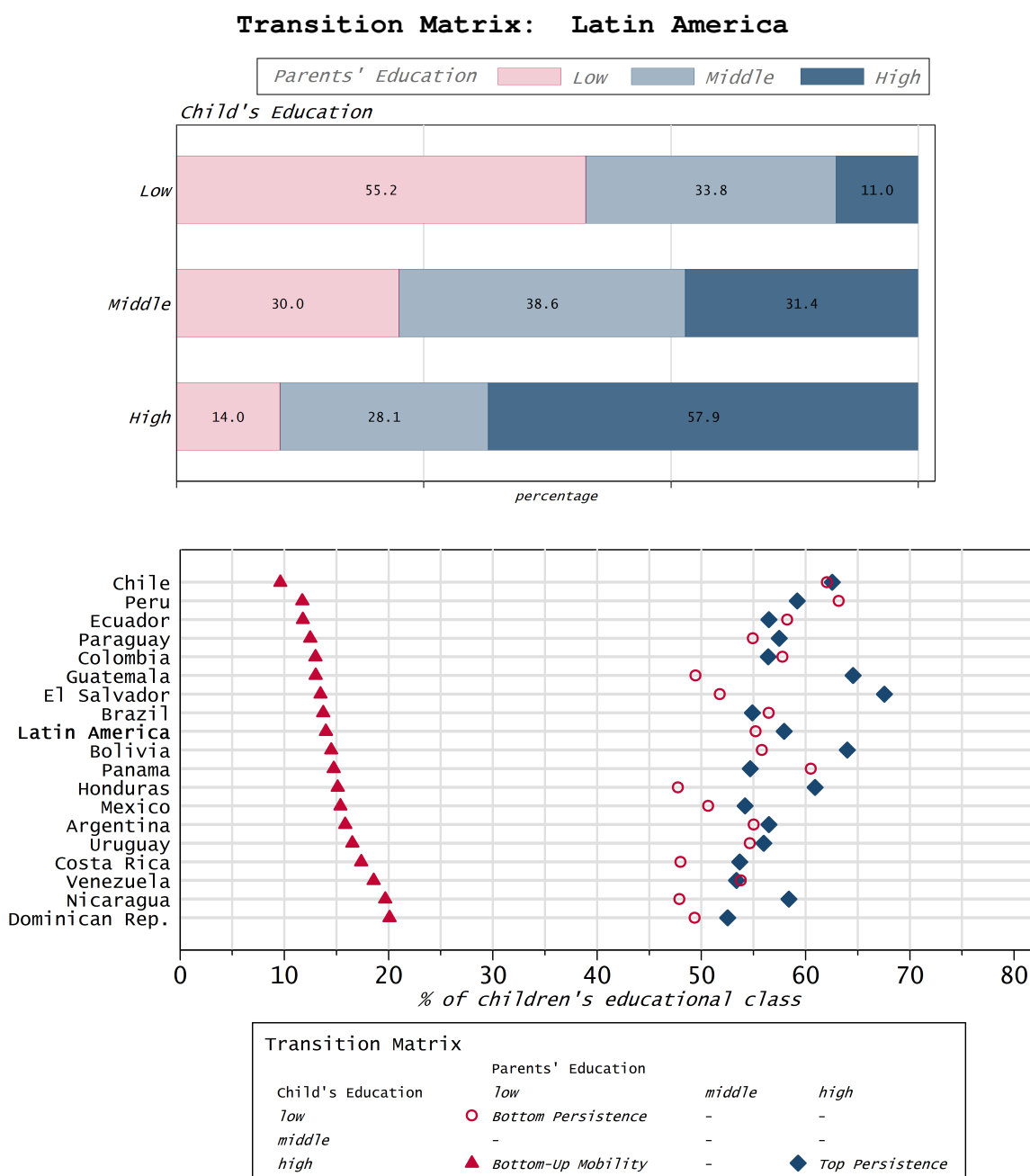
Notes: Average household per capita income (constant 2005 PPP international USD). Educational levels: 1 without education or primary incomplete; 2 primary complete; 3 secondary incomplete; 4 secondary complete; 5 tertiary incomplete; 6 tertiary complete. Numbers show the ratio of the monetary returns to education for people with a completed tertiary degree (category 6) and without education or with incomplete primary education (category 1). O: Older Cohort. Y: Younger Cohort. Example on how to read this numbers: In Argentina, individuals with completed tertiary degree born between 1945 and 1960 have a 4.4 times higher average household per capita income than their peers without education or with incomplete primary education. *Source:* SEDLAC circa 2005, own estimates.

Figure 3: Absolute educational mobility in Latin America.



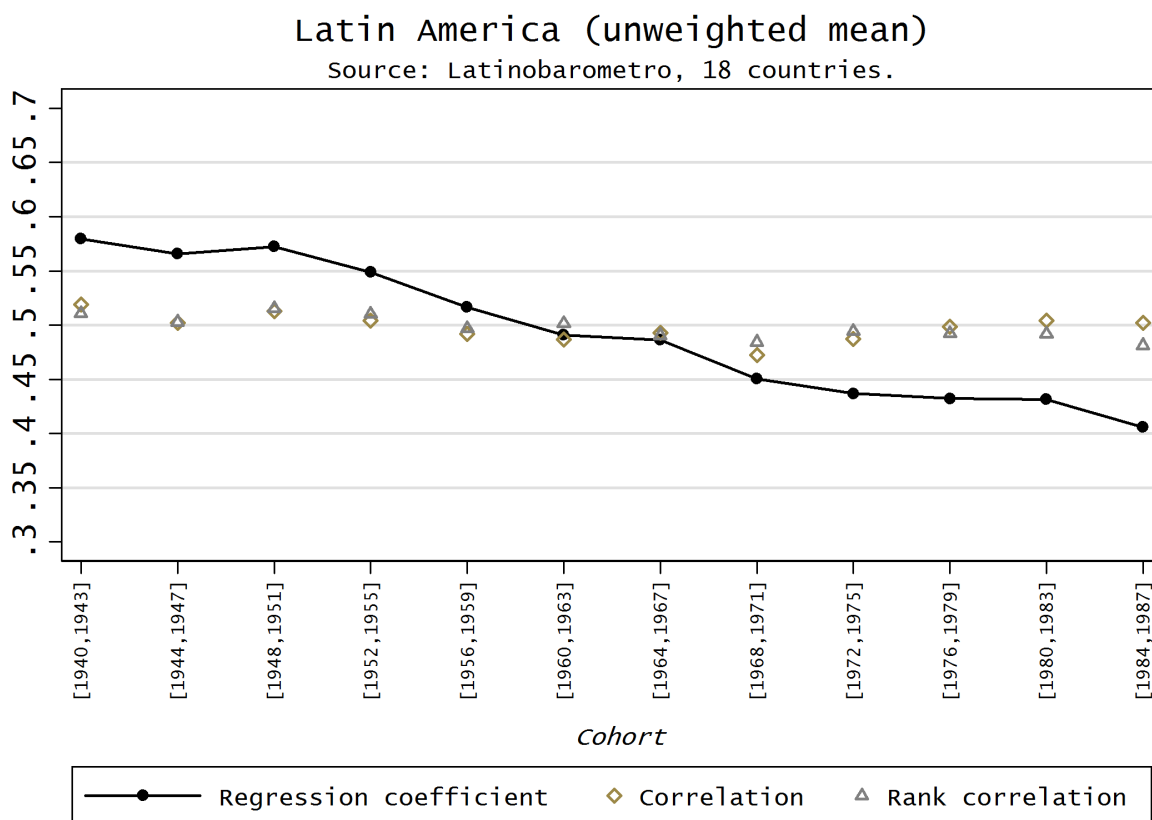
Source: Latinobarometro 1998-2015, own estimates.

Figure 4: Educational persistence in Latin America: Insights from transition matrices (People born 1940-1990).

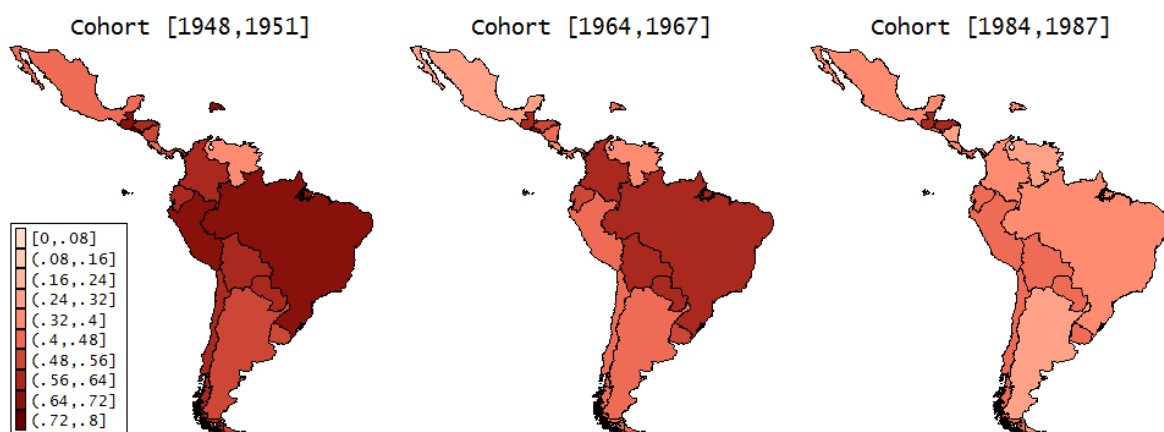


Notes: The points show the percentage of individuals in three different cells of the transition matrix. *Bottom persistence:* Individuals with low education and low parental education. *Bottom-Up Mobility:* Individuals with high education and low parental education. *Top persistence:* Individuals with high education and high parental education. Educational classes (low, middle, high) refer to three quantiles of the within-country and within-cohort distributions. Benchmarks USA (PSID, own estimates) / Germany (SOEP, own estimates): *Bottom persistence* 61.5 % / 56.5 %, *Top persistence* 51.2 % / 55.8 %, *Bottom-up mobility* 21.5 % / 17.8 %. *Source:* Latinobarometro 1998-2015, own estimates.

Figure 5: Educational persistence in Latin America: Regression and correlation coefficients.

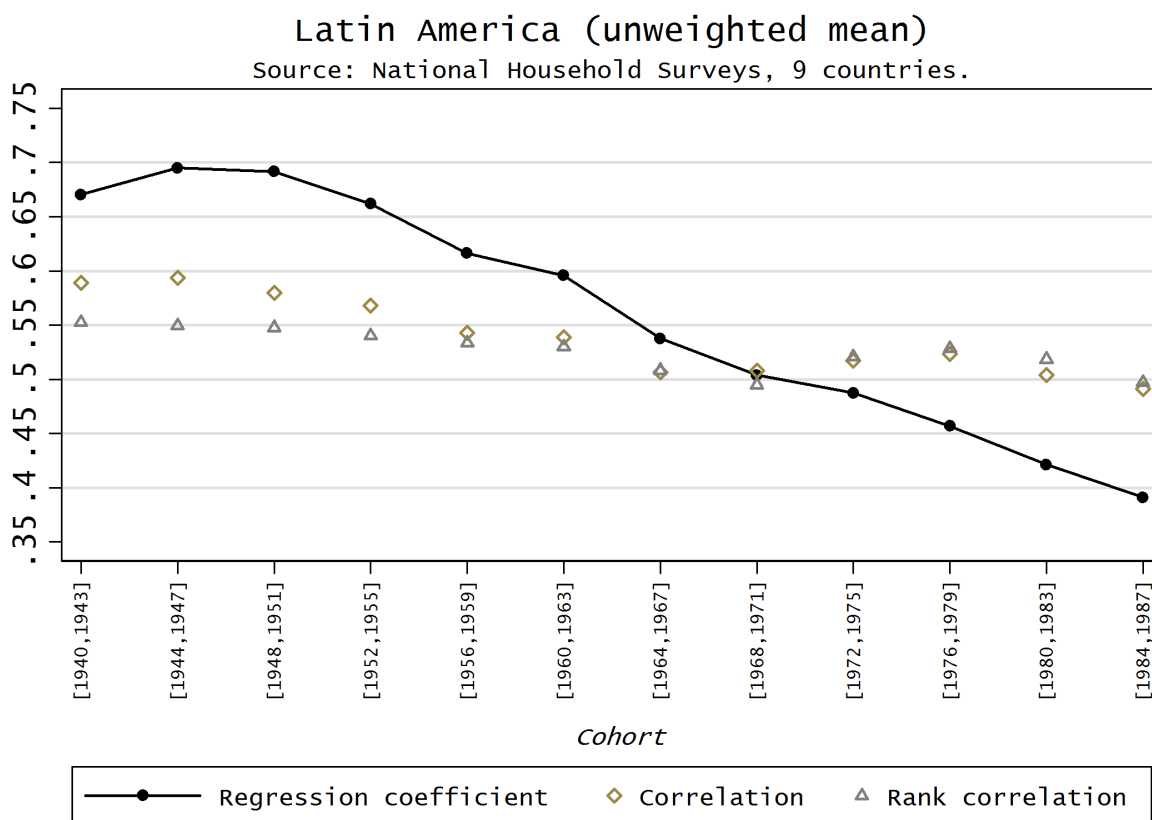


Regression coefficient: Geography and Trends for Latin America



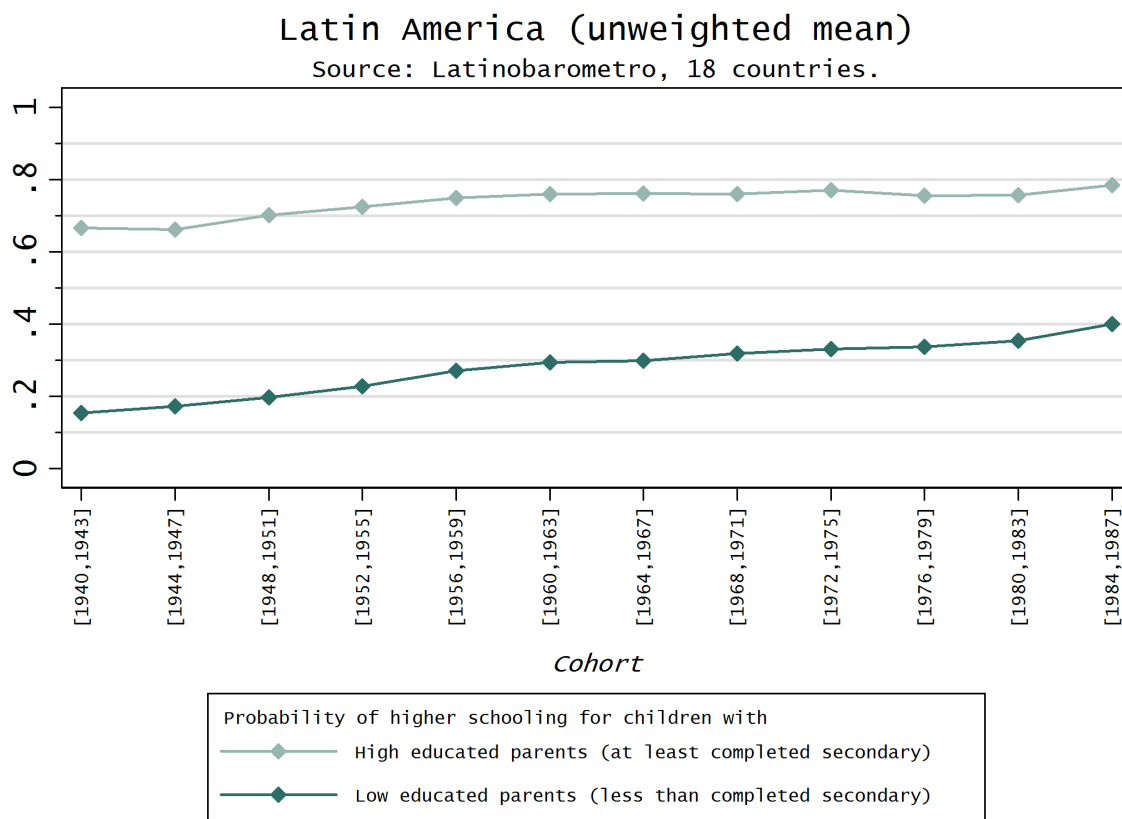
Notes: Points show the unweighted mean over all countries of the estimates for each cohort. Samples for each cohort and country restricted to individuals older than 22. Bootstrapped confidence interval. *Source:* Latinobarometro 1998-2015, own estimates.

Figure 6: Educational persistence in Latin America: Regression and correlation coefficients.

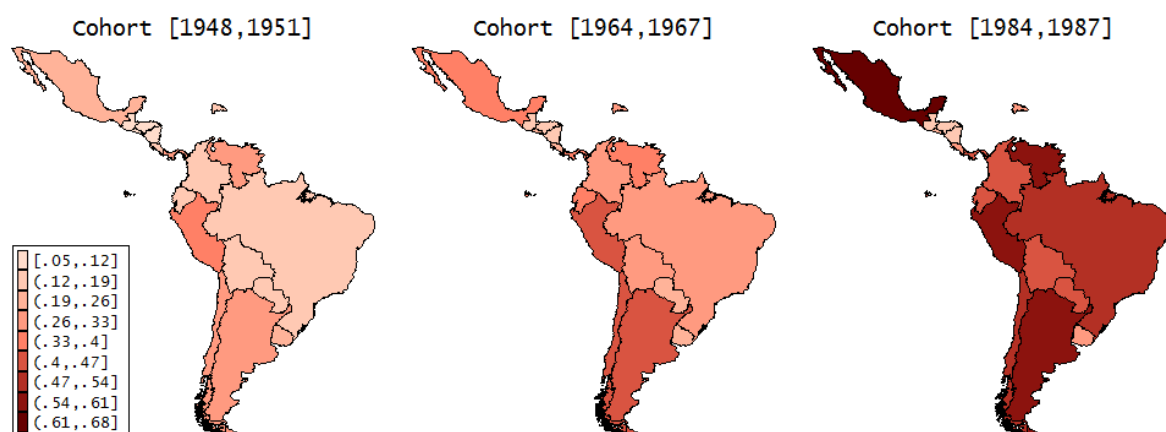


Notes: Points show the unweighted mean over all countries of the estimates for each cohort. Samples for each cohort and country restricted to individuals older than 22. Bootstrapped confidence interval. *Source:* National Household Surveys 1982-2015, own estimates.

Figure 7: Educational inequality in Latin America: bottom-upward Mobility (*BUM*) and upper class persistence (*UCP*).

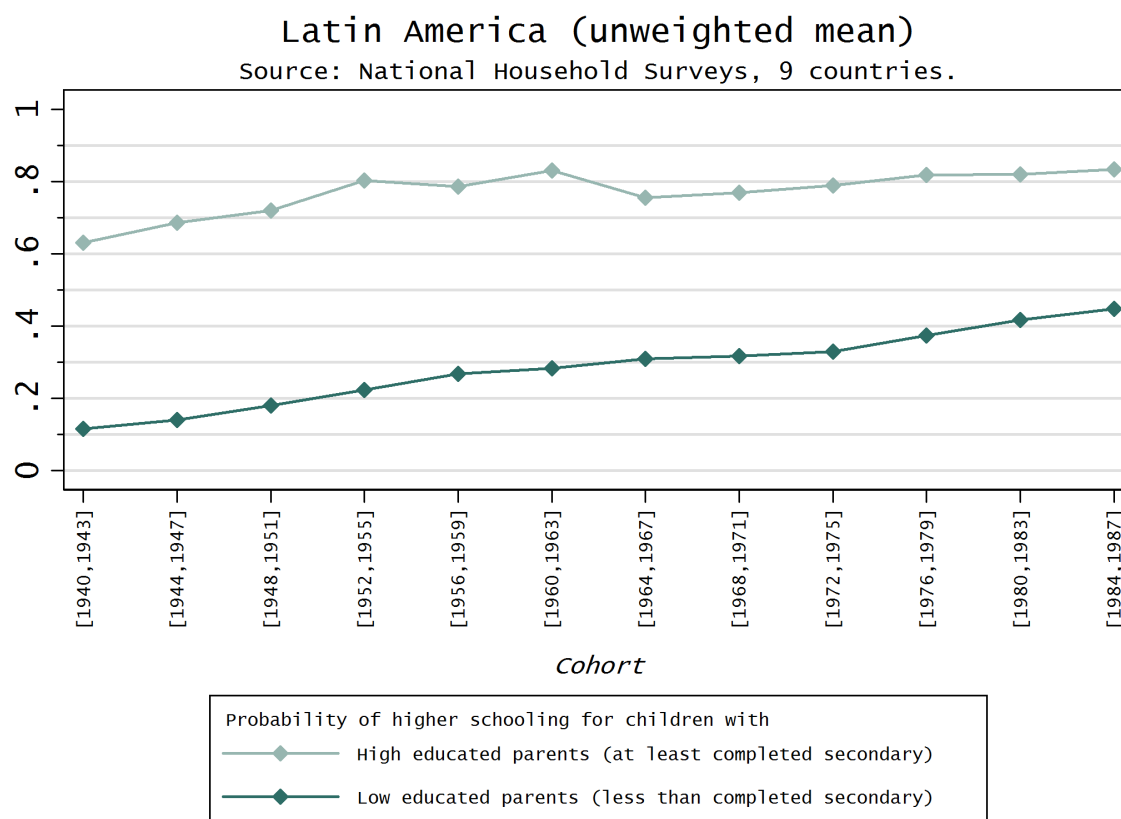


Bottom upward Mobility: Geography and Trends for Latin America



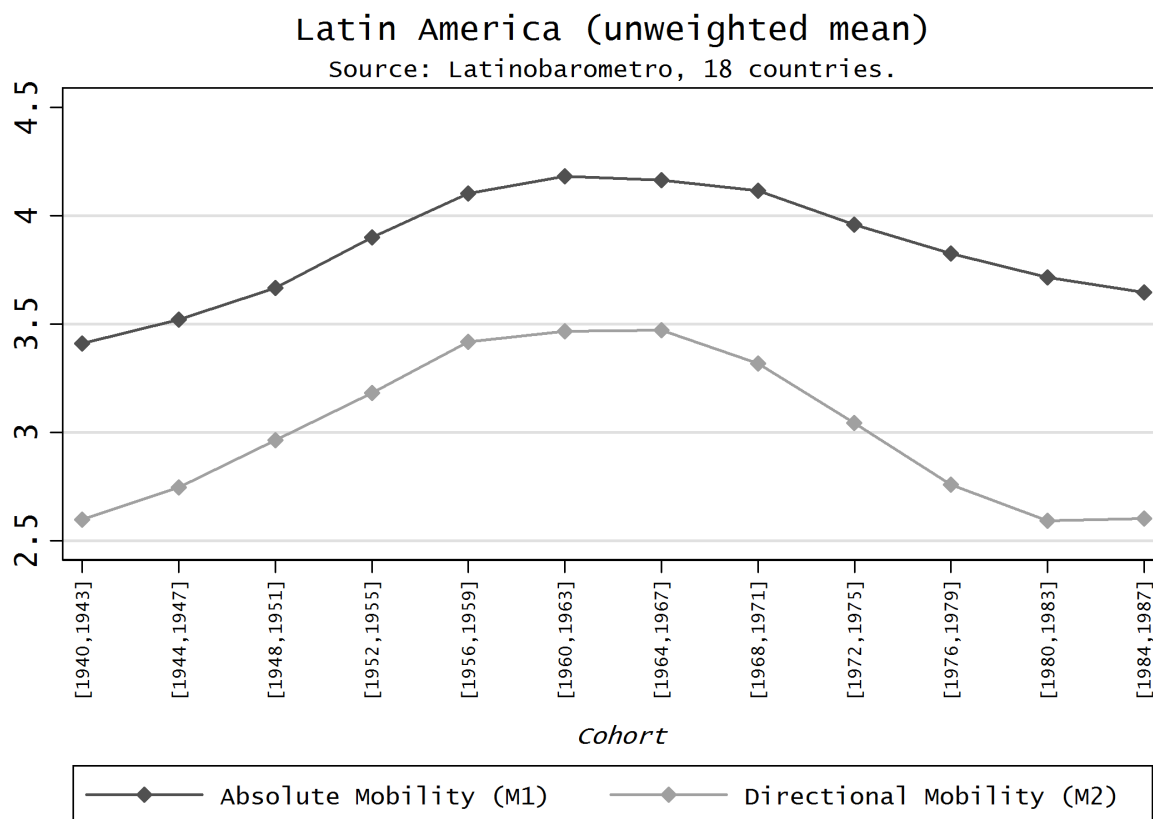
Notes: Estimated probability of higher education (at least completed secondary) of children with different parental educational background. Points show the unweighted mean over all countries of the estimates for each cohort. Samples for each cohort and country restricted to individuals older than 22. Bootstrapped confidence interval. *Source:* Latinobarometro 1998-2015, own estimates.

Figure 8: Educational inequality in Latin America: bottom-upward Mobility (*BUM*) and upper class persistence (*UCP*).

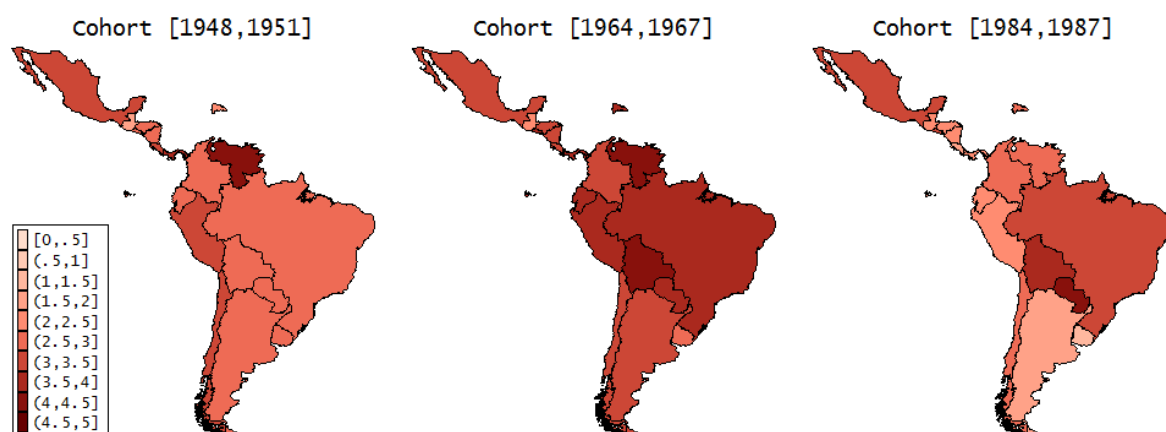


Notes: Estimated probability of higher education (at least completed secondary) of children with different parental educational background. Points show the unweighted mean over all countries of the estimates for each cohort. Samples for each cohort and country restricted to individuals older than 22. Bootstrapped confidence interval. *Source:* National Household Surveys 1982-2015, own estimates.

Figure 9: Educational mobility in Latin America: absolute ($M1$) and directional ($M2$) mobility in years of education.

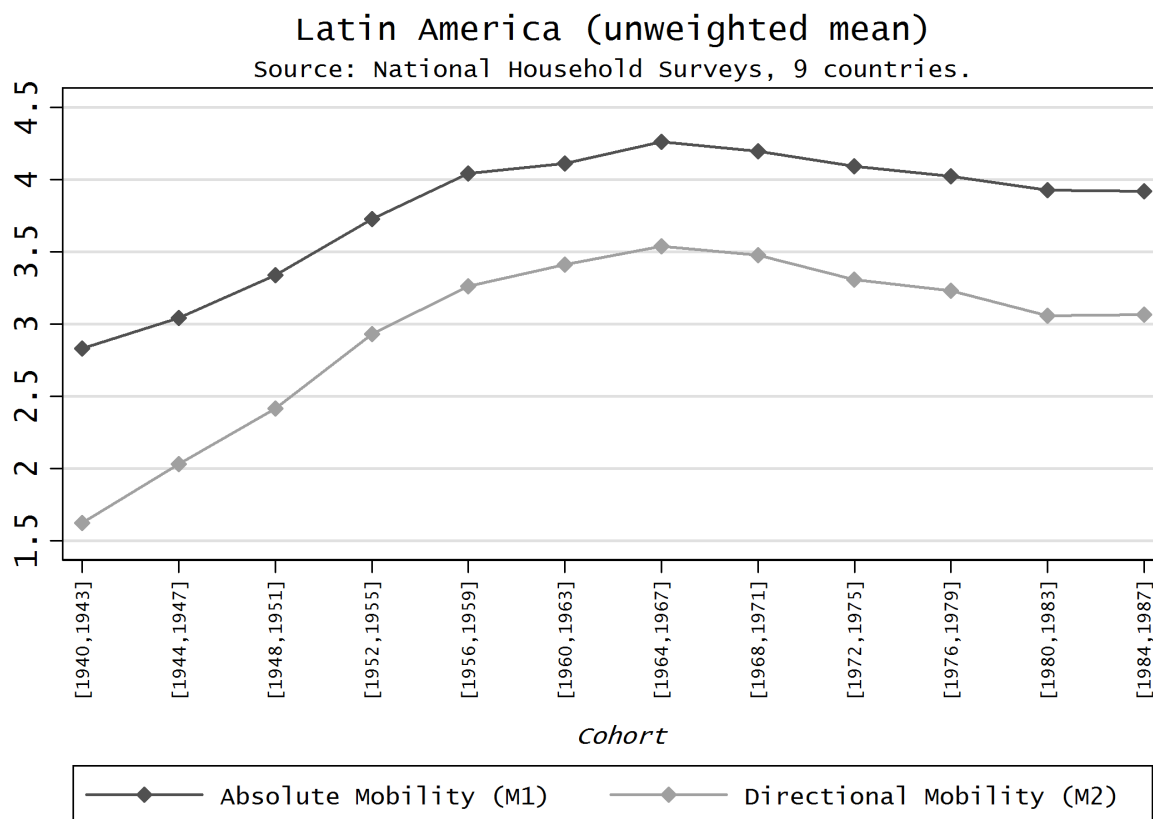


Directional Mobility: Geography and Trends for Latin America



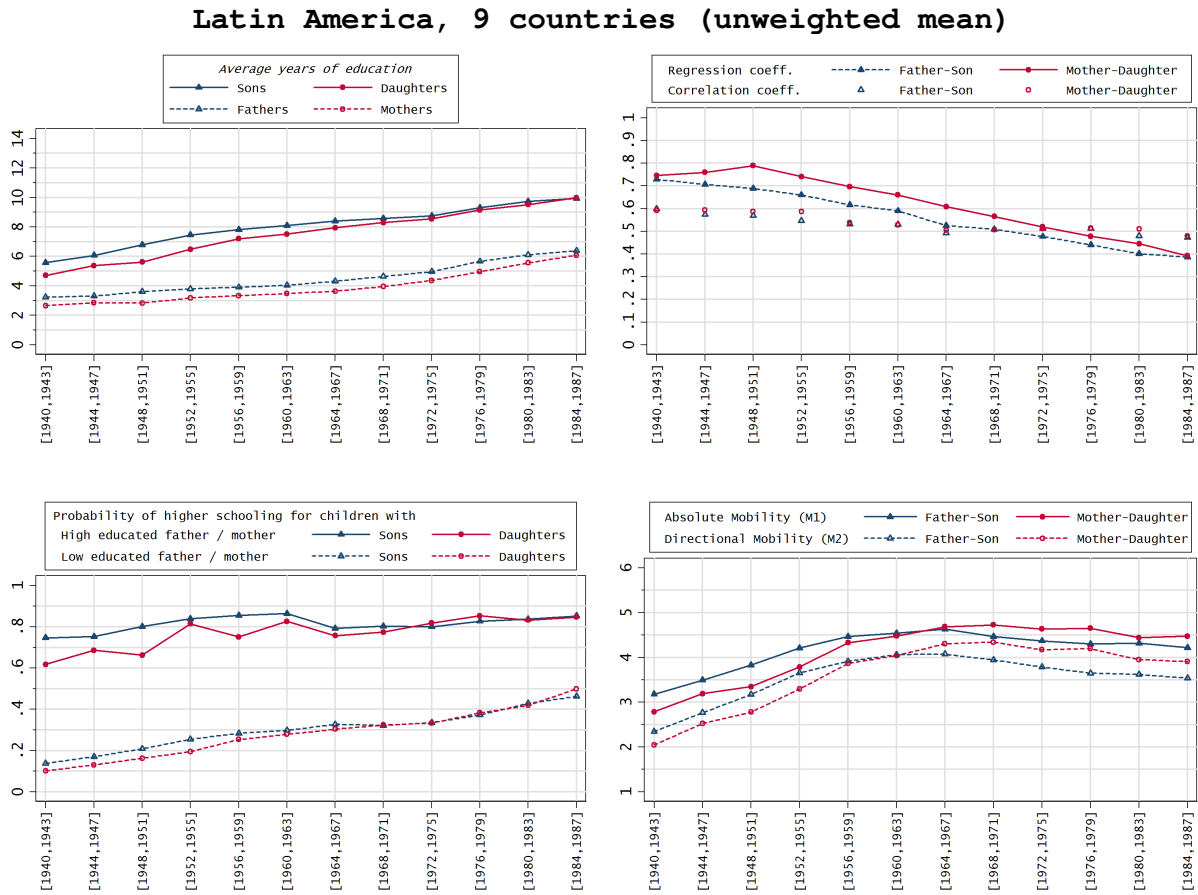
Notes: Points show the unweighted mean over all countries of the estimates for each cohort. Samples for each cohort and country restricted to individuals older than 22. Source: Latinobarometro 1998-2015, own estimates.

Figure 10: Educational mobility in Latin America: absolute ($M1$) and directional ($M2$) mobility in years of education.



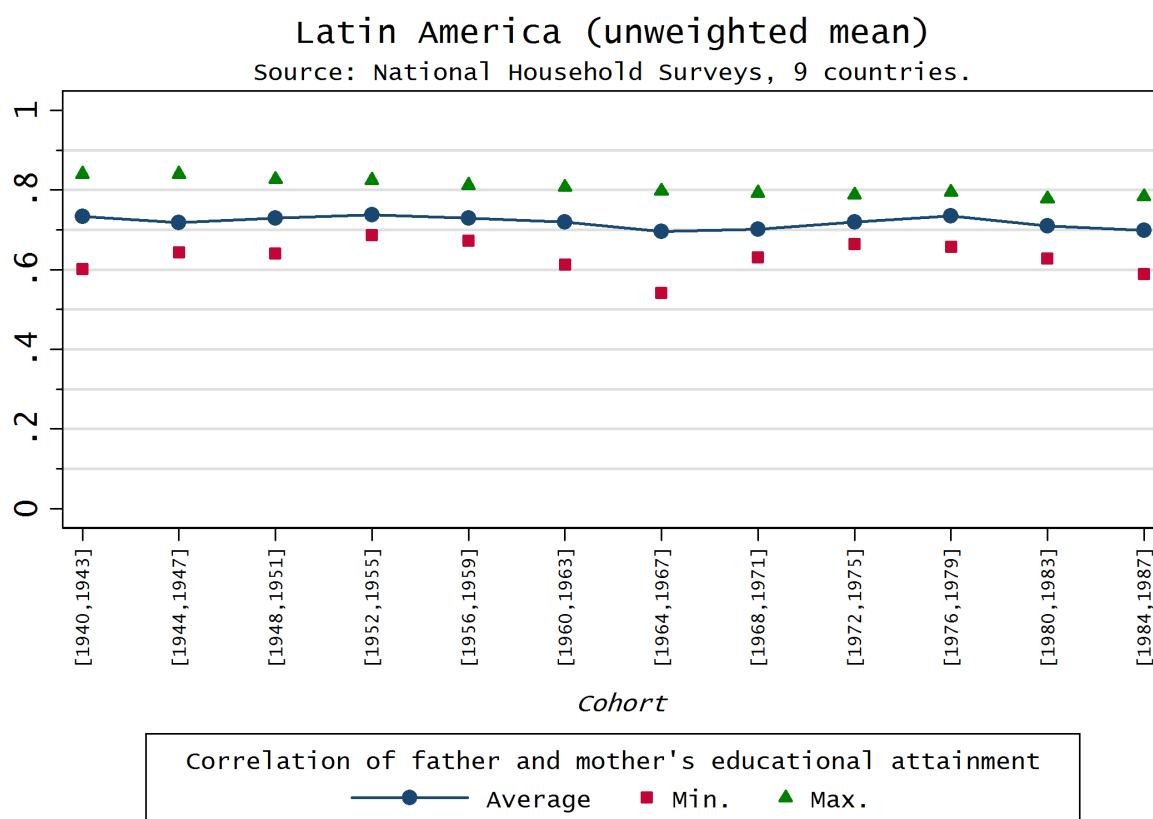
Notes: Points show the unweighted mean over all countries of the estimates for each cohort. Samples for each cohort and country restricted to individuals older than 22. *Source:* National Household Surveys 1982-2015, own estimates.

Figure 11: Average educational attainment by gender and intergenerational mobility for father-son and mother-daughter pairs.



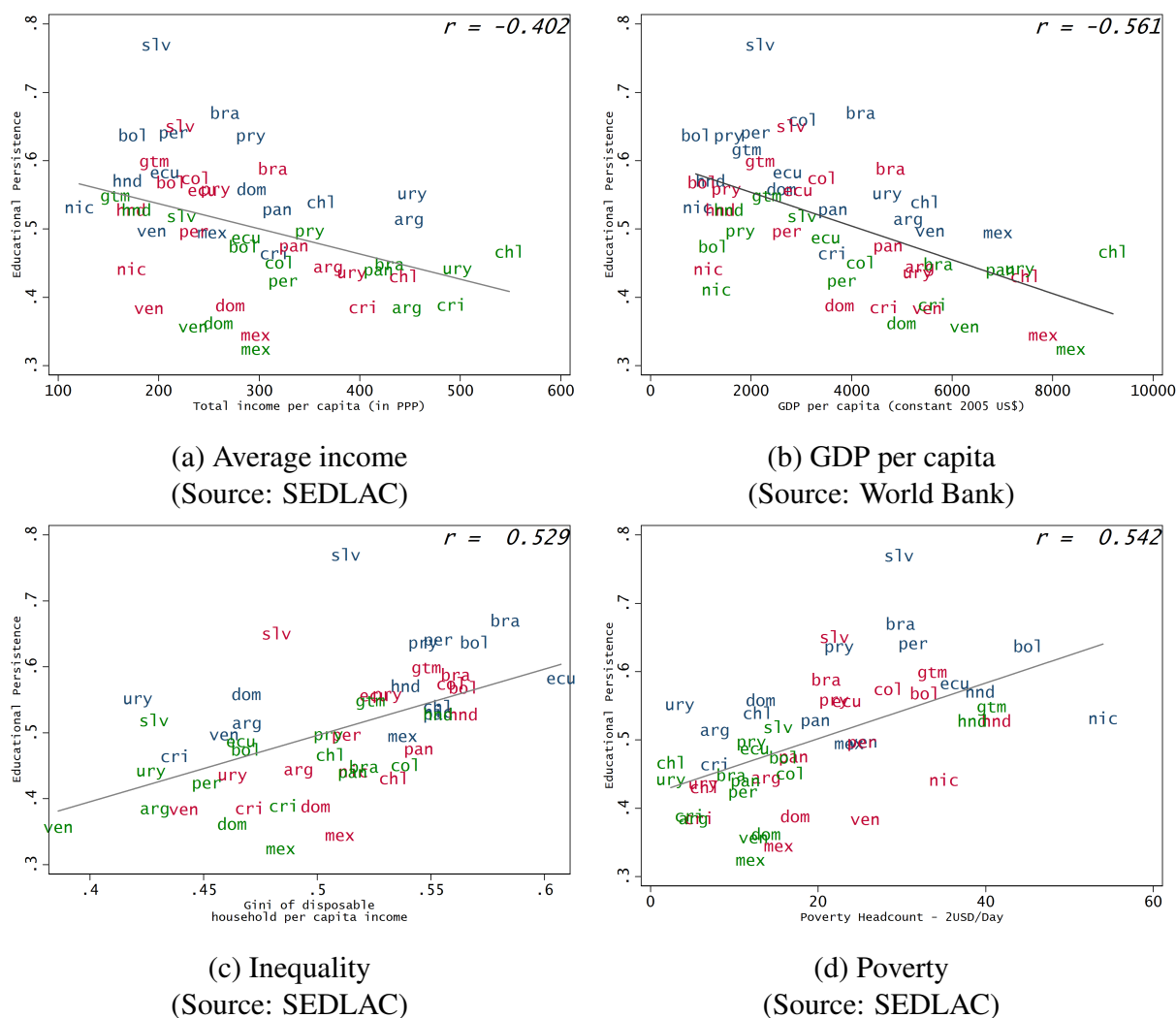
Source: National Household Surveys 1982-2015, own estimates.

Figure 12: Assortative mating – spouse correlation in educational attainments (parental generation).



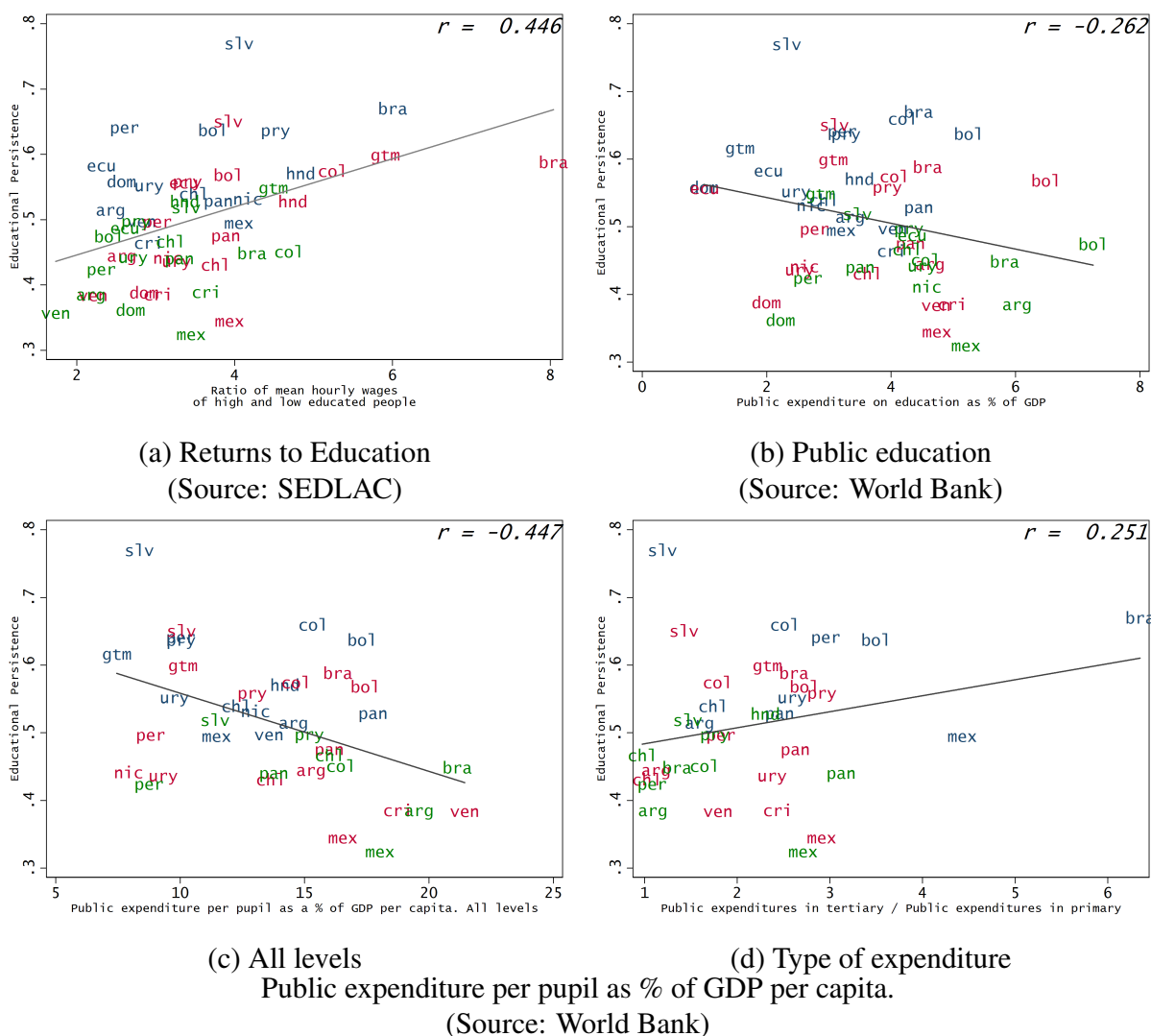
Notes: Points show the unweighted mean over all countries of the estimates for each cohort. Samples for each cohort and country restricted to individuals older than 22. *Source:* National Household Surveys 1982-2015, own estimates.

Figure 13: Educational persistence and economic performance.



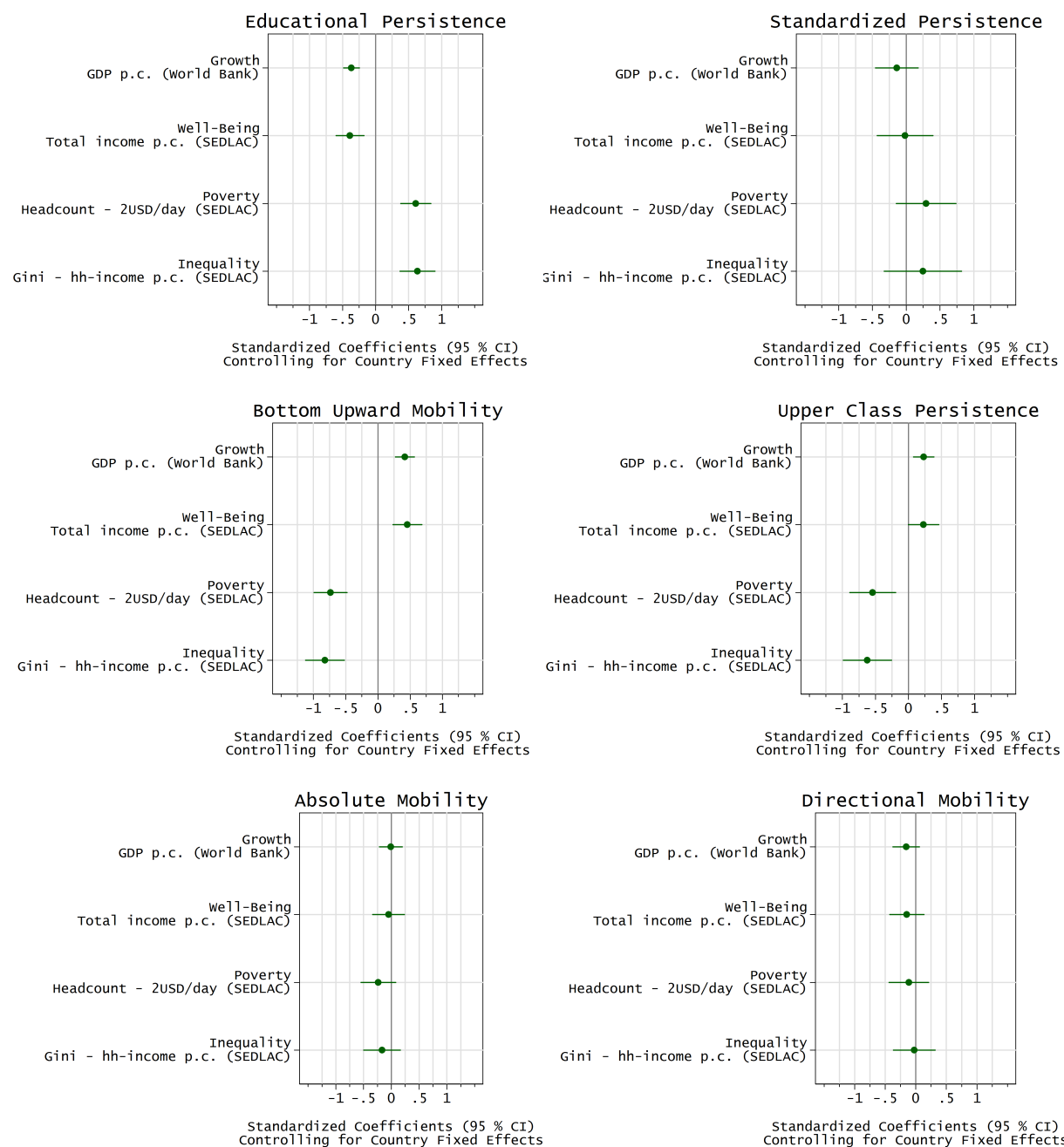
Notes: Intergenerational mobility of the cohorts '40-'54, '55-'69, '70-'84 is associated with the corresponding macroeconomic or institutional characteristic in the years 1990-99, 2000-09, 2010-14. *Sources:* Latinobarometro 1998-2015, own estimates of educational persistence; SEDLAC; World Bank Data.

Figure 14: Educational persistence and institutional characteristics of the education system.



Notes: Intergenerational mobility of the cohorts '40-'54, '55-'69, '70-'84 is associated with the corresponding macroeconomic or institutional characteristic in the years 1990-99, 2000-09, 2010-14. *Sources:* Latinobarometro 1998-2015, own estimates of educational persistence; SEDLAC; World Bank Data.

Figure 15: Intergenerational mobility and economic performance. Regression analysis controlling for country fixed effects.



Notes: Intergenerational mobility of the cohorts '40-'54, '55-'69, '70-'84 is associated with the corresponding macroeconomic or institutional characteristic in the years 1990-99, 2000-09, 2010-14. Points show the standardized linear regression coefficients and their corresponding confidence interval (95 %) of separate regressions for each of the indicated variables. All regressions control for country fixed effects. In cases where the confidence interval line crosses the zero-line, point estimates are not statistically significant at the 0.05 level. *Sources:* Latinobarometro 1998-2015, own estimates of intergenerational mobility; SEDLAC; World Bank Data.

SUPPLEMENTAL MATERIAL

For Online Publication

APPENDIX A Summary of Data Sources: National Household Surveys	I
APPENDIX B Description of the Database	VII
APPENDIX C Country-Wise Estimates	VIII

A Summary of Data Sources

A.1 Household Surveys

Our main source of information for all 18 Latin American countries in our analysis is the Latino-barometro survey. Using the survey waves 1998 to 2015 our overall sample comprises 211,401 observations. We complement this with National Household Surveys that include information on parental educational achievements collected through retrospective questions. This second data set comprises 1,078,445 observation in total that derive from different data sources.

Data from Brazil comes from the *Pesquisa Nacional por Amostra de Domicilios* (PNAD), which is carried out by the *Instituto Brasileiro de Geografia y Estadísticas* (IBGE) on a yearly basis. This survey included mobility modules in 1982, 1988, 1996 and 2014. Since the coding of the educational variable is not comparable between 2014 and the other three survey waves, we opt to use only the most recent one in our analysis. The survey is nationally and regionally representative, rural and urban, except for the rural areas of the Northern Region, which roughly corresponds to the Amazon rainforest and accounted for 2.3% of Brazil's population in the 2000 Census.

For Chile, we use the *Encuesta de Caracterización Socioeconómica Nacional* (CASEN), which is a nationally and regionally representative household survey carried out by the Ministry of Social Development (in collaboration with the National Institute of Statistics, INE) through the Department of Economics at the *Universidad de Chile*, which is responsible for the data collection, digitalization and consistency checking of the database.¹⁷ The survey has been regularly implemented every two years since 1985 during November and in some cases, up to mid-December. We use surveys for 2006 to 2015, since previous surveys don't provide information about parents.

The same is true for Peru, using the *Encuesta Nacional de Hogares* (ENAHO), which is carried out in four waves since 1997, and continues until today. The fourth wave of the survey is nationally representative, and it is officially used to estimate poverty rates. After year 2000 the survey was enlarged and a new sample frame was used, including questions about parents. We use surveys

¹⁷Before 2011 the survey was carried out by the Ministry of Planning (MIDEPLAN).

for 2001 to 2015. However, from 2002 on the survey asked only the household head about the education of parents. Since most household heads are male the sex composition of our sample is therefore unbalanced.

For the other countries we use different versions of Living Standards Measurement Surveys, originally developed and promoted by the World Bank, which are all nationally representative. Data from Ecuador comes from the *Encuesta de Condiciones de Vida* (ECV) for years 1994, 1995, 1998 and 2006. In the case of Colombia we use the *Encuesta Nacional de Condiciones de Vida* (ECV), which was carried out by the *Departamento Administrativo Nacional de Estadística* (DANE). We use surveys for six years between 2003 and 2013. Although Guatemala is a country with relatively few household surveys, the *Encuesta Nacional sobre Condiciones de Vida* (ENCOVI) have information about individuals' parents (2000, 2006 and 2011). Panama carried out Living Standards Measurement Surveys in 1997, 2003 and 2008, which are called *Encuesta Nacional sobre Condiciones de Vida* (ENV).

The source of information for our estimations of Mexico's statistics is the Mexican Family Life Survey (MxFLS), which is a longitudinal and multi-thematic survey, representative of the Mexican population at the national, urban, rural and regional level. The MxFLS has been developed and managed by researchers from the Iberoamerican University (UIA, per its name in Spanish) and the Center for Economic Research and Teaching (CIDE, per its name in Spanish) in collaboration with researchers from Duke University. Currently, the MxFLS contains information for a 10-year period, collected in three rounds: 2002, 2005-2006 and 2009-2012.

Finally, for Nicaragua the only useful source for our analysis we could find besides Latino-barometro is the 1998 wave of the *Encuesta Nacional de Hogares sobre Medición de Nivel de Vida* (EMNV).

A. SUMMARY OF DATA SOURCES

Table A1: *Household surveys used to construct the intergenerational mobility estimates*

Country	Name of survey	Acronym	Coverage	Survey waves
Argentina	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Bolivia	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Brazil	<i>Pesquisa Nacional por Amostra de Domicilios</i>	PNAD	National	2014
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Chile	<i>Encuesta de Caracterización Socioeconómica Nacional</i>	CASEN	National	2006, 2009, 2011, 2013, 2015
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Colombia	<i>Encuesta Nacional de Condiciones de Vida</i>	ECV	National	2003, 2008, 2010, 2011, 2012, 2013
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Costa Rica	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Dominican Rep.	<i>Latinobarometro</i>		National	2004-2011, 2013, 2015

A. SUMMARY OF DATA SOURCES

Table A1: *Household surveys used to construct the intergenerational mobility estimates*

Country	Name of survey	Acronym	Coverage	Survey waves
Ecuador	<i>Encuesta de Condiciones de Vida</i>	ECV	National	1994, 1995, 1998, 2006
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
El Salvador	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Guatemala	<i>Encuesta Nacional sobre Condiciones de Vida</i>	ENCOVI	National	2000, 2006, 2011
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Honduras	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Mexico	<i>Encuesta Nacional sobre Niveles de Vida de los Hogares</i>	MXFLS	National	2002, 2005-2006, 2009-2012
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Nicaragua	<i>Encuesta Nacional de Hogares sobre Medición de Nivel de Vida</i>	EMNV	National	1998
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Panama	<i>Encuesta de Niveles de Vida</i>	ENV	National	1997, 2003, 2008
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015

A. SUMMARY OF DATA SOURCES

Table A1: *Household surveys used to construct the intergenerational mobility estimates*

Country	Name of survey	Acronym	Coverage	Survey waves
Paraguay	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Peru	<i>Encuesta Nacional de Hogares</i>	ENAHO	National	2001-2015
	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Uruguay	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015
Venezuela	<i>Latinobarometro</i>		National	1998, 2000-2011, 2013, 2015

A.2 Codification of Educational Attainment

	0	Illiterate
	1	Incomplete primary
	2	'
	3	'
	4	'
Completed Years of Education	5	'
	6	Complete primary
	7	'
	8	Incomplete secondary
	9	'
	10	'
	11	Complete secondary
	12	'
	13	Incomplete university or technical training
	14	Complete technical training
	15	Complete university

B Description of the Database

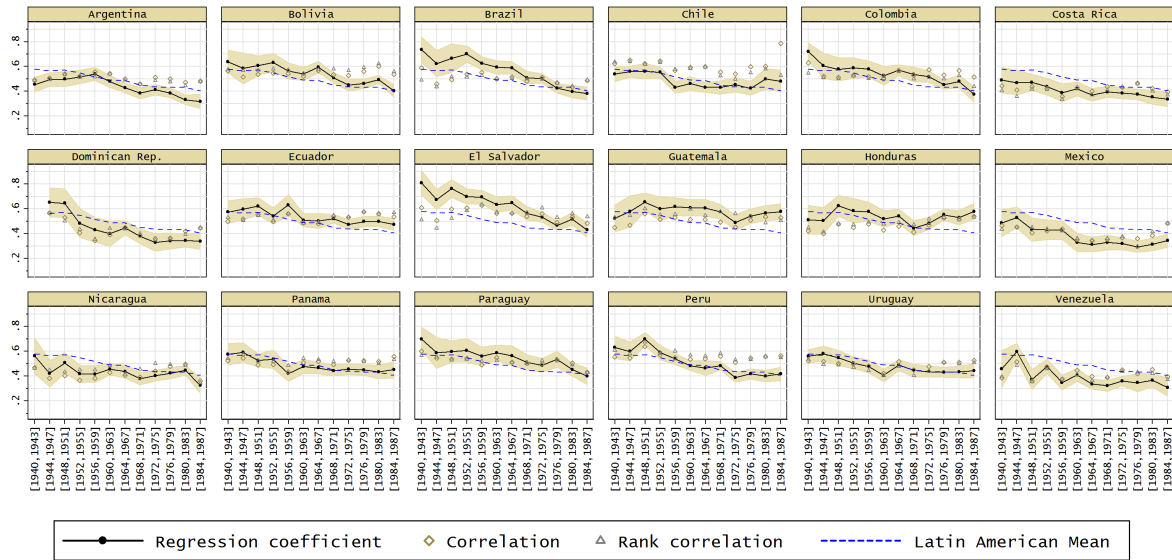
We provide databases containing all mobility indicators described in this project. The variables contained in each database are described in Table B1. The data is divided in four different sets of different periodization of the birth cohorts, separated at intervals of one to four years, respectively. In addition to the main statistics and the identification variables of each country, survey and cohort, we also include complementary variables that may be useful, such as mean and variance of the years of education of individuals and their parents, the average age of individuals and the share of males in the sample for each cohort. Finally, we add a variable that contains the number of observations used for the estimation of mobility statistics to make it possible to evaluate the quality of the estimates.

Table B1: Summary table of the database.

Variable	Label	Definition	Mean	Std. Dev.	Min	Max
country	Country name	Name of country				
idenpa	Country code	World Bank country code				
cohort	Cohort	Cohort indicator	7.28	3.95	1.00	14.00
survey	Survey name	Name of the survey				
N	Number of observations	Number of observations used to estimate indicators	3421.34	7508.60	19.00	45046.00
b	Intergenerational persistence parameter	Conditional correlation between years of education of children and parents (beta)	0.49	0.14	0.02	0.91
bstd	Intergenerational correlation (b standardized)	Parameter b weighted by the ratio of standard deviations of years of schooling of children and parents	0.50	0.09	0.06	0.79
corr_spearman	Spearman's correlation	Spearman's rank correlation coefficient (rho)	0.49	0.08	-0.05	0.67
blog	Intergenerational elasticity	Parameter b estimated using the logarithm of the outcome of interest (years of schooling)	0.34	0.12	0.00	0.70
prob_high	Prob(high education) High parental education	Predicted probability of upper class persistence (UCP)	0.75	0.13	0.16	0.97
prob_low	Prob(high education) Low parental education	Predicted probability of bottom upward mobility (BUM)	0.27	0.15	0.03	0.81
M1	Absolute mobility	Absolute mobility (M1)	3.79	0.68	1.60	5.23
M2	Directional mobility	Directional mobility (M2)	2.90	0.83	0.50	4.78
educ	Years of schooling	Average of own years of schooling	8.14	2.19	2.22	14.26
educ_parents	Parental Years of schooling	Average of parents' years of schooling (the highest level of educational attainment among the two)	5.25	2.12	1.39	12.58
var	Variance of years of schooling	Variance of own years of schooling	16.66	5.43	0.84	33.08
var_parents	Variance of parental years of schooling	Variance of parents' years of schooling	17.56	4.31	6.79	32.96
age	Age	Average age of individuals in sample	40.93	13.59	23.00	72.54
male	Share of males	Share of males in sample	0.49	0.06	0.33	0.81

C Country-Wise Estimates

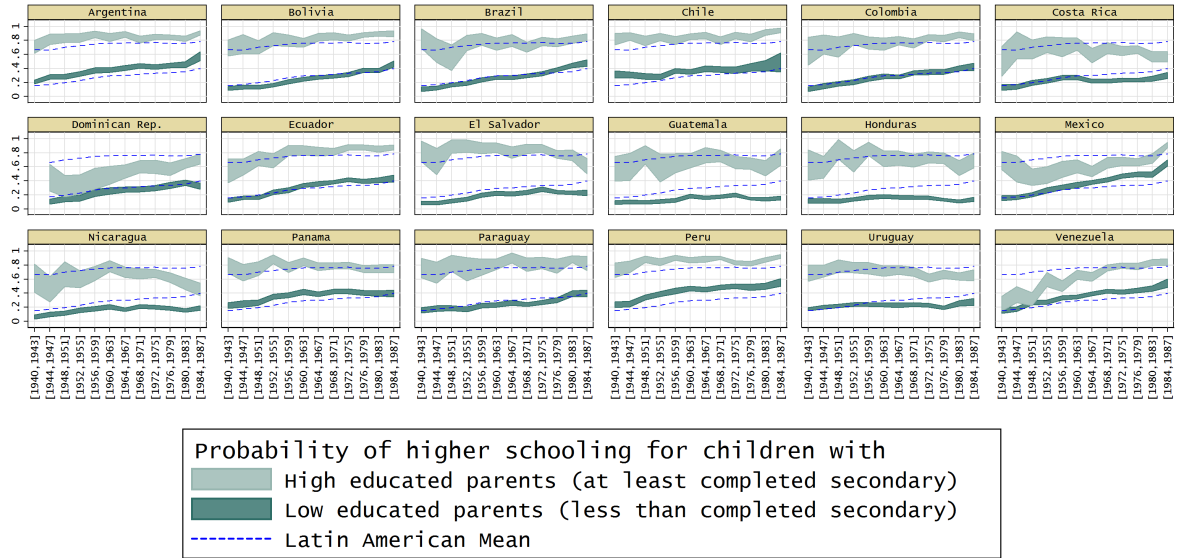
Figure C1: Educational persistence in Latin America: Regression and correlation coefficients by country. *Source:* Latinobarometro 1998-2015, own estimates.



Notes: Samples for each cohort and country restricted to individuals older than 22. Only point estimates displayed relying on at least 200 observations. Bootstrapped confidence interval.

C. COUNTRY-WISE ESTIMATES

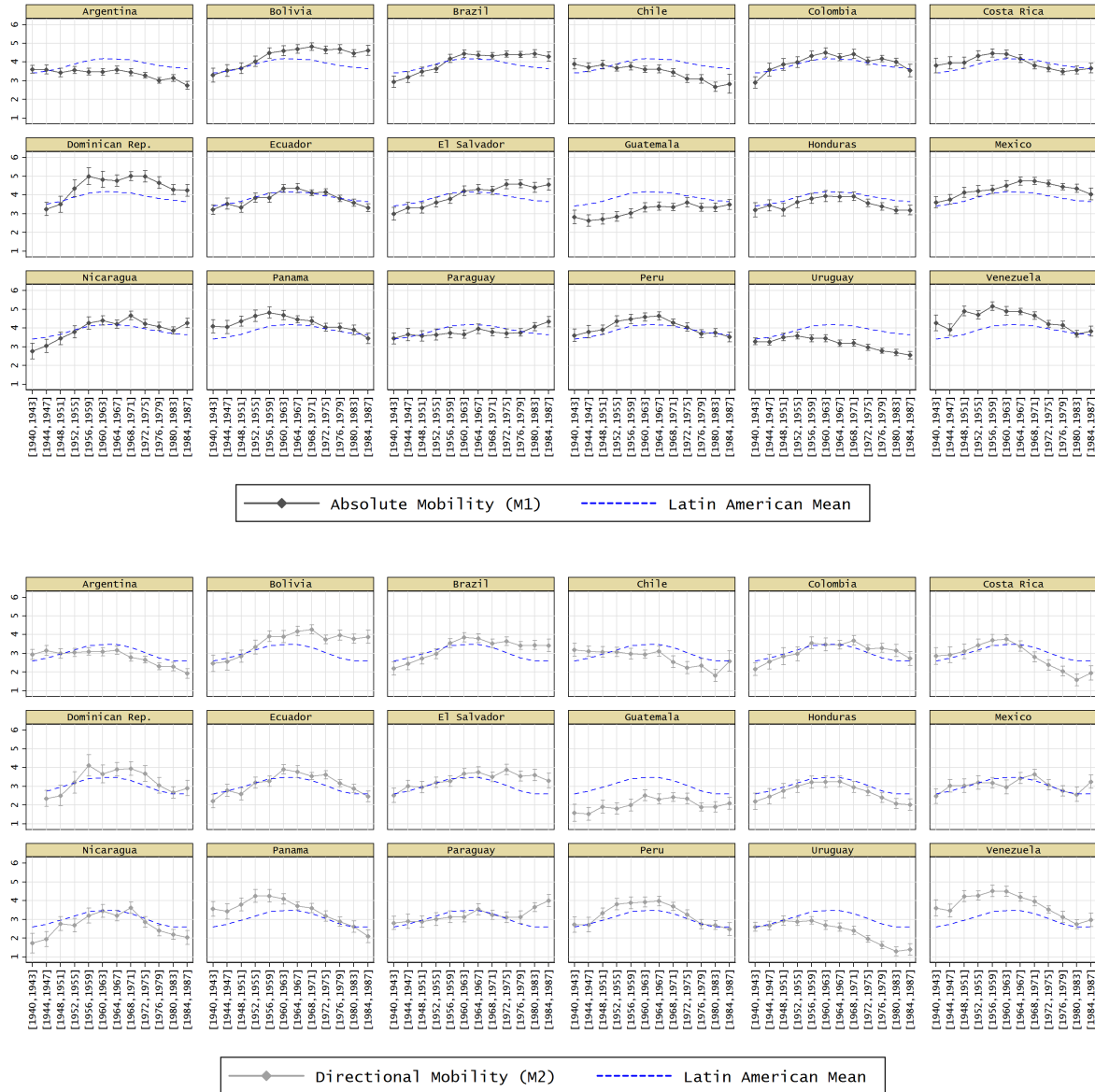
Figure C2: Educational inequality in Latin America: bottom-upward Mobility (*BUM*) and upper class persistence (*UCP*). *Source*: Latinobarometro 1998-2015, own estimates.



Notes: Samples for each cohort and country restricted to individuals older than 22. Only point estimates displayed relying on at least 200 observations. Bootstrapped confidence interval.

C. COUNTRY-WISE ESTIMATES

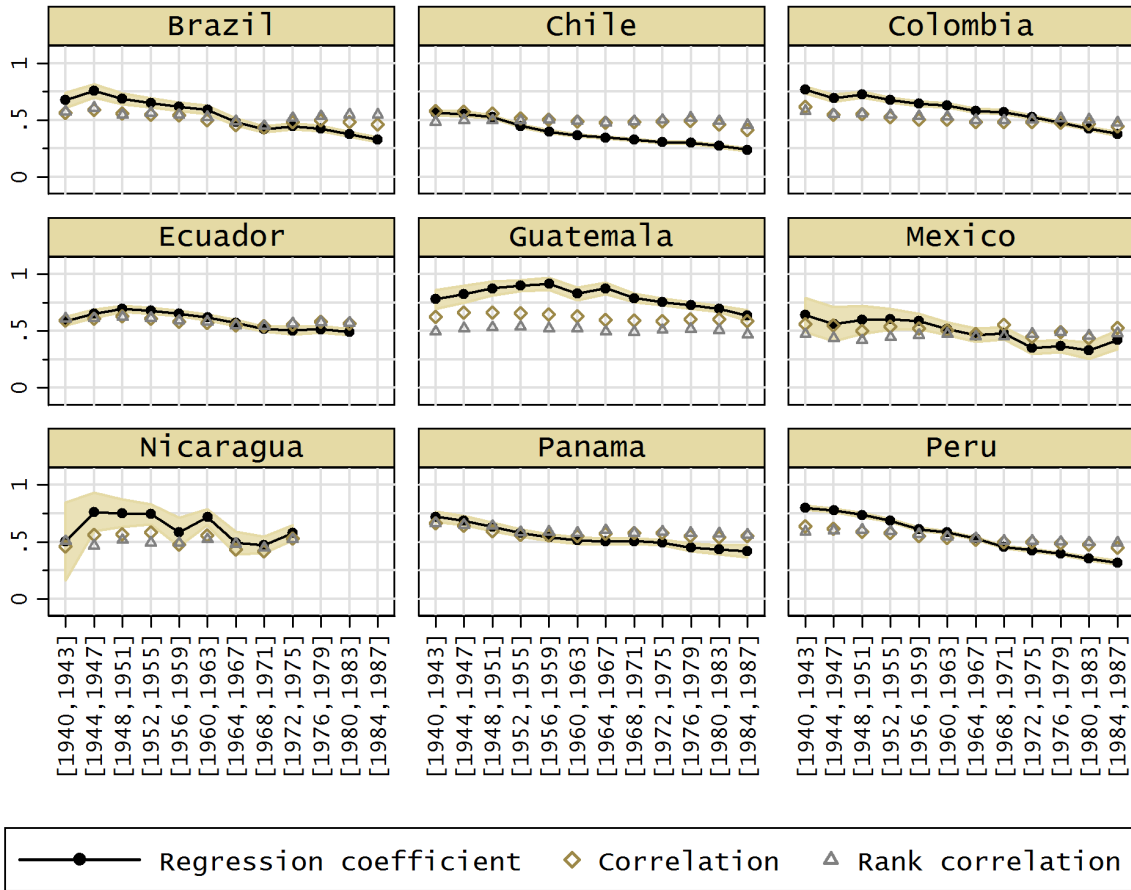
Figure C3: Educational mobility in Latin America: absolute ($M1$) and directional ($M2$) mobility in years of education. *Source:* Latinobarometro 1998-2015, own estimates.



Notes: Samples for each cohort and country restricted to individuals older than 22. Only point estimates displayed relying on at least 200 observations. Bootstrapped confidence interval.

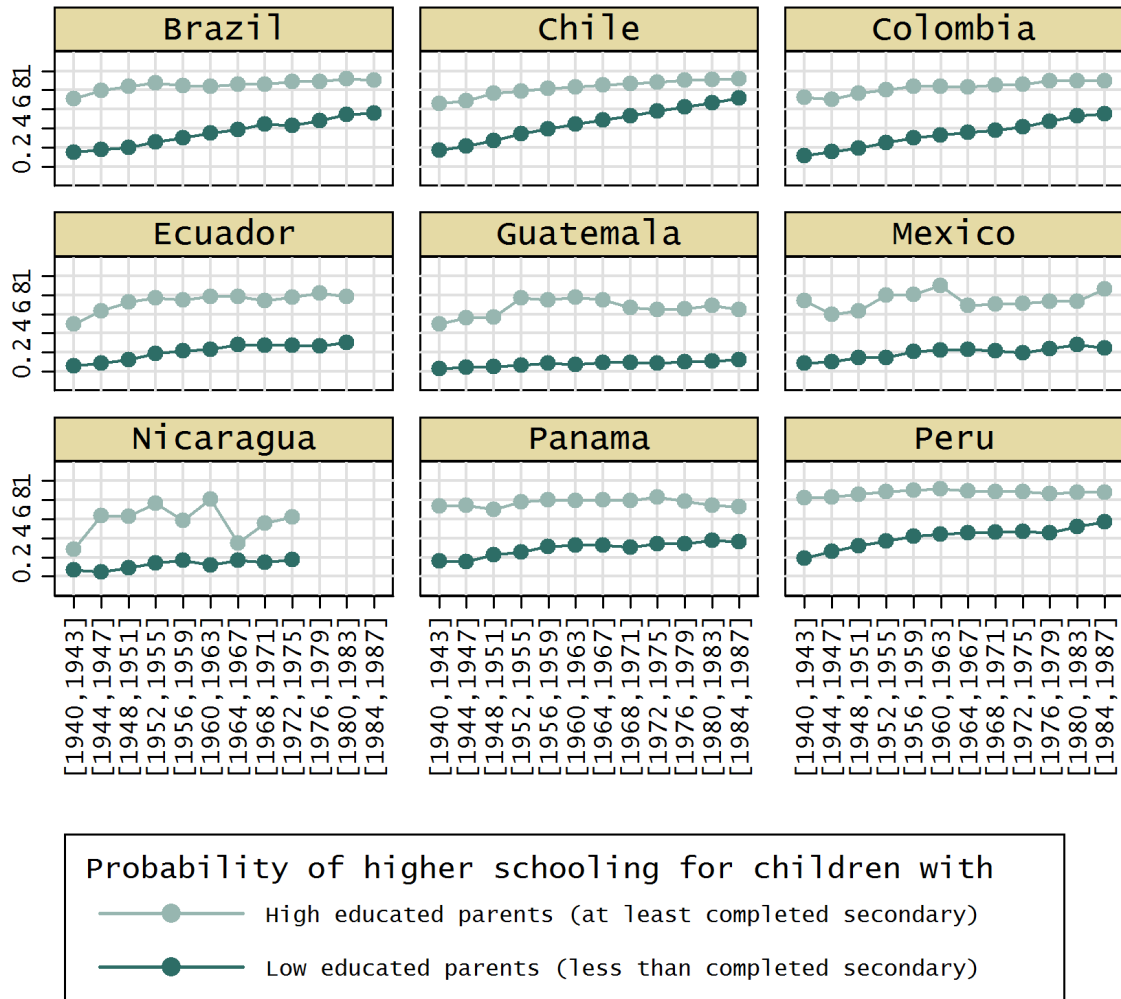
C. COUNTRY-WISE ESTIMATES

Figure C4: Educational persistence in Latin America: Regression and correlation coefficients by country. *Source:* National Household Surveys 1982-2015, own estimates.



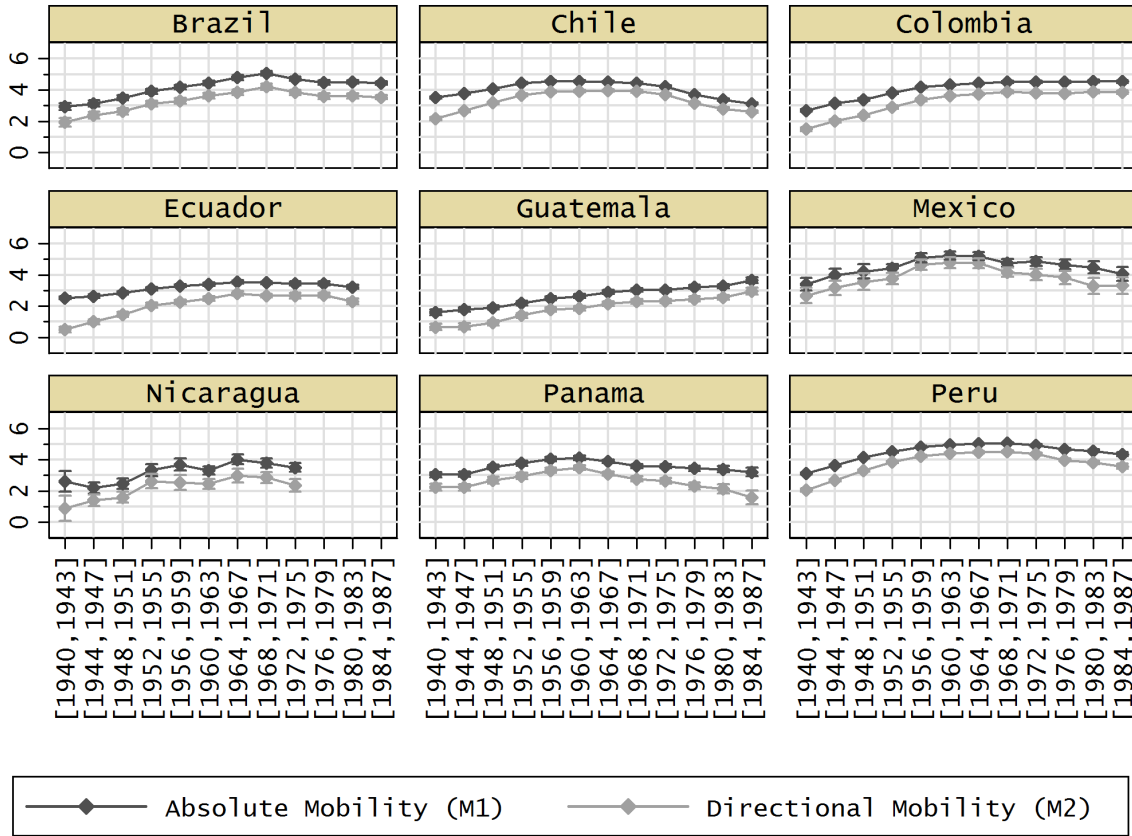
Notes: Samples for each cohort and country restricted to individuals older than 22. Only point estimates displayed relying on at least 200 observations. Bootstrapped confidence interval.

Figure C5: Educational inequality in Latin America: bottom-upward Mobility (*BUM*) and upper class persistence (*UCP*). *Source:* National Household Surveys 1982-2015, own estimates.



Notes: Samples for each cohort and country restricted to individuals older than 22. Only point estimates displayed relying on at least 200 observations. Bootstrapped confidence interval.

Figure C6: Educational mobility in Latin America: absolute ($M1$) and directional ($M2$) mobility in years of education. *Source:* National Household Surveys 1982-2015, own estimates.



Notes: Samples for each cohort and country restricted to individuals older than 22. Only point estimates displayed relying on at least 200 observations. Bootstrapped confidence interval.

C. COUNTRY-WISE ESTIMATES

Figure C7: Educational persistence in Latin America for father-son and mother-daughter pairs.
 Source: National Household Surveys 1982-2015, own estimates.

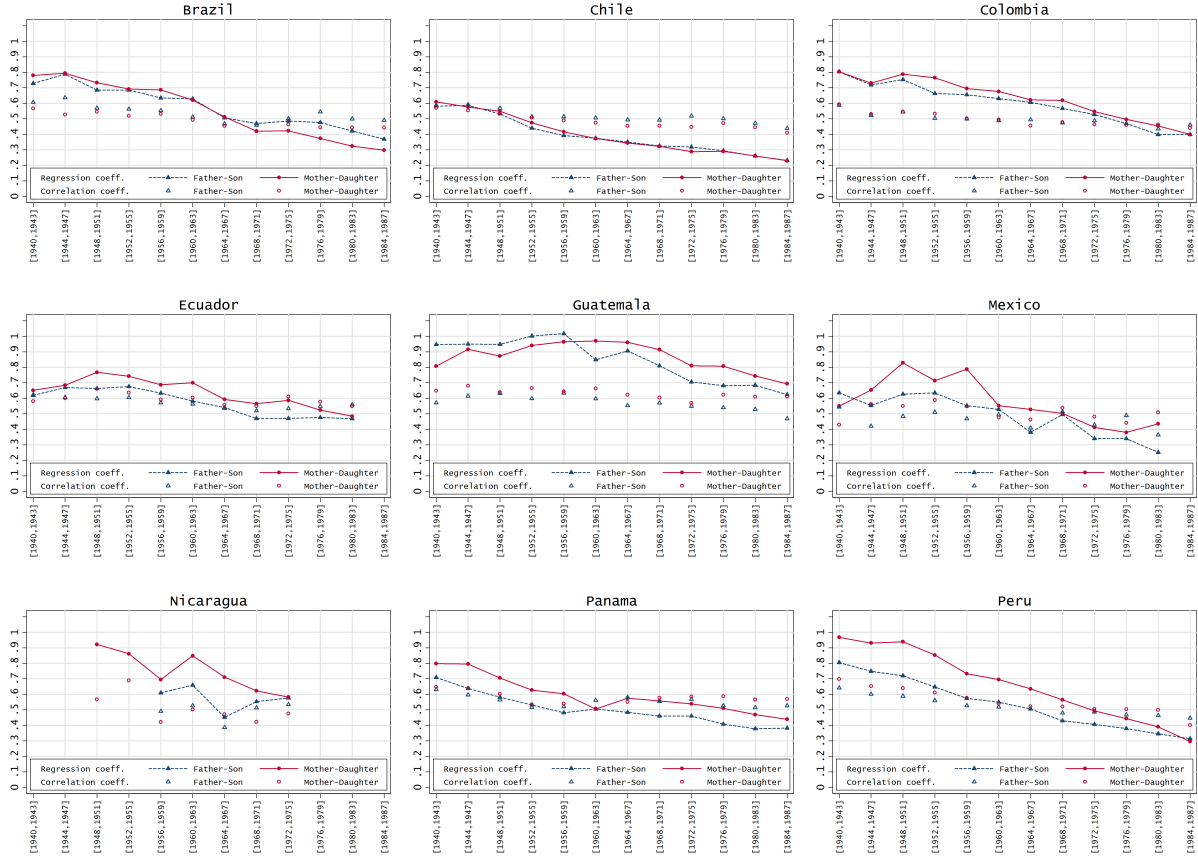
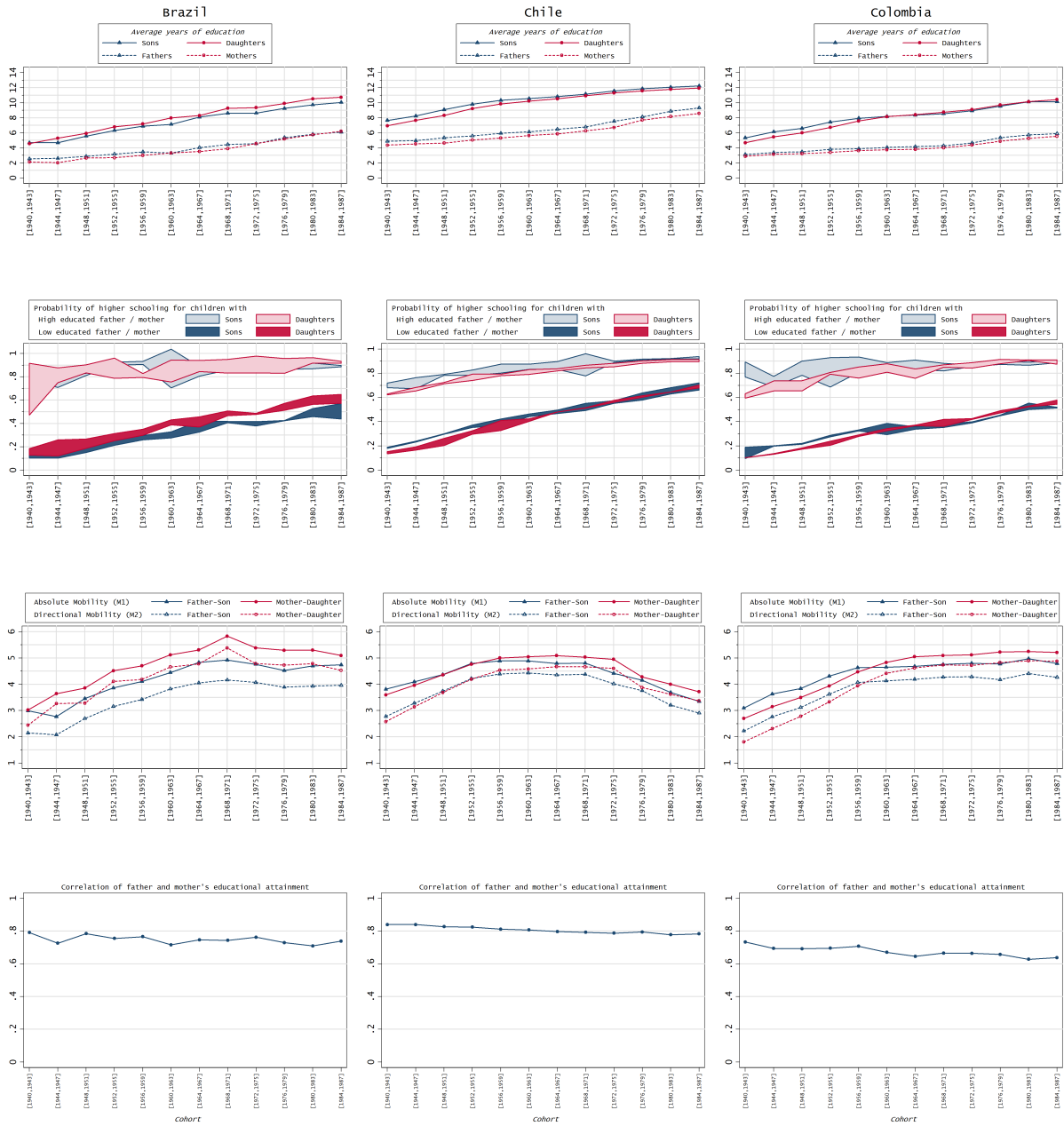


Figure C8: Average educational attainment, intergenerational mobility for father-son and mother-daughter pairs, and assortative mating. *Source:* National Household Surveys 1982-2015, own estimates.



C. COUNTRY-WISE ESTIMATES

Figure C9: Average educational attainment, intergenerational mobility for father-son and mother-daughter pairs, and assortative mating. *Source:* National Household Surveys 1982-2015, own estimates.

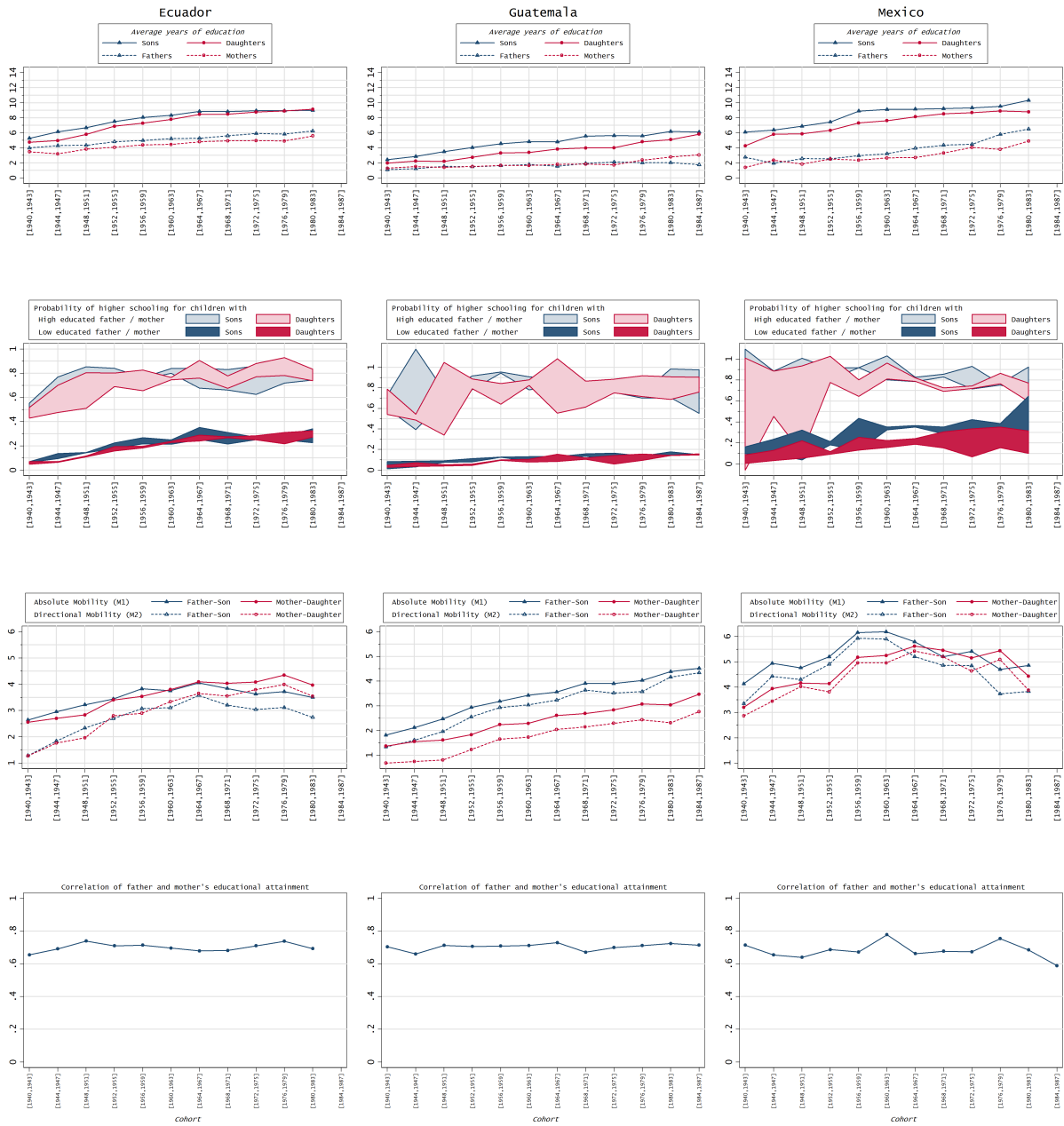


Figure C10: Average educational attainment, intergenerational mobility for father-son and mother-daughter pairs, and assortative mating. *Source:* National Household Surveys 1982-2015, own estimates.

