More inequality, less social mobility

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We investigate the relationship between inequality and intergenerational mobility. Proxying fathers’ earnings with using detailed occupational data, we find that sons who grew up in countries that were more unequal in the 1970s were less likely to have experienced social mobility by the late-1990s.

I. Introduction

A common view among citizens of large industrialized countries is that economic inequality is fair, provided there are equal opportunities.\textsuperscript{1} At the same time, there tends to be a belief that equal opportunity norms are violated when the degree of intergenerational mobility is low and family background exerts a strong influence on children’s income in adulthood. It is therefore reasonable to think that inequality may be more acceptable in a society with a high level of social mobility.

Despite this important conceptual nexus between social mobility and inequality, the literatures on inequality and intergenerational mobility have largely developed in isolation from one another. That very little is known about the association between inequality and intergenerational mobility stands in contrast to the burgeoning literature on the consequences of inequality for variables such as economic growth, health and political behaviour. To the extent that other studies have looked at the relationship between inequality and social mobility, the analysis has been descriptive, or focused around the question of ‘American exceptionalism’. To our knowledge, there is no previous study that has formally tested the hypothesis that there is a relationship between a country’s level of inequality and the degree of intergenerational mobility.

From a theoretical perspective, the relationship between inequality and intergenerational mobility is unclear. One possibility is that when inequality between parents increases, intergenerational mobility will fall because it is easier for rich parents to buy their children educational advantages that less well-off parents cannot afford (Burtless and Jencks, 2003). But as Solon (2004) argues, this effect will be undermined to the extent that children from less advantaged backgrounds disproportionately benefit from public programs.

Another channel through which inequality might affect intergenerational mobility is via the demand for redistribution. In more unequal societies, the median voter will tend to lie further below the mean income and may have a stronger preference for redistribution (Alesina and Glaeser, 2004). Conversely, if higher inequality increases the political influence of the wealthy – perhaps through campaign finance contributions – then the scope for government to institute progressive policies may narrow (Burtless and Jencks, 2003). Finally, higher inequality might reduce intergenerational mobility to the extent that it leads to segregation along income lines, resulting in adverse peer effects for children from low-income families (Durlauf, 1996).

II. Estimating Intergenerational Mobility

A major obstacle to systematic empirical research into the link between inequality and intergenerational mobility, proxied fathers’ earnings with using detailed occupational data, we find that sons who grew up in countries that were more unequal in the 1970s were less likely to have experienced social mobility by the late-1990s.
mobility is the lack of suitable data. The ideal dataset to address this question would have two main features. First, it would be comparable such that cross-country differences in estimated mobility are meaningful and do not derive from differences in data construction across countries. Second, it would contain panel data on the incomes of fathers and sons at economically active ages. The 1999 Social Inequality III module of the International Social Survey Program (ISSP) – which we utilize in this article – scores highly on the first criteria to the extent that it contains information, collected on a consistent basis, for individuals from a large number of countries. However, this is partly offset by the fact that the ISSP does not explicitly contain data on parental earnings. We therefore follow a spate of previous studies (e.g. Björklund and Jäntti, 1997; Grawe, 2001; Leigh, 2007) in using predicted parental earnings as a proxy for actual parental earnings.

Our empirical strategy involves a three-step estimation procedure, using data on men aged 25–54 in the 1999 ISSP. First, for each of the 16 countries where data on fathers’ occupation is available, we regress the relationship between log hourly wages $y_{ij}$ of individual $i$ in occupation $j$ on a vector of dummies for each occupation $X_{ij}$, and a quadratic in age $A_i$,

$$y_{ij} = \theta_j X_{ij} + A_i + A_i^2 + \varepsilon_{ij}$$

(1)

Second, the earnings of fathers in occupation $j$ are then predicted to be the same as those of a 40 year old in occupation $j$. Algebraically, where $A = 40$, $\hat{y}_{40j} = \hat{y}_j$.

Third, we estimate the relationship between sons’ actual log hourly wages and fathers’ predicted log hourly wages,

$$y_{si} = \alpha + \beta \hat{y}_{40j} + A_i + A_i^2 + \varepsilon_i$$

(2)

These three steps are performed separately for each country in the sample.

The coefficient $\beta$ in Equation 2 denotes the intergenerational elasticity (IGE), being the percentage change in the son’s earnings for doubling of the father’s earnings. Another common measure of intergenerational mobility is $\rho$, the intergenerational correlation (IGC) which is based on estimating the same regression, but with the variance in earnings held constant between the two periods. The relationship between the two measures is a function of the ratio of the standard deviation of earnings in the two generations,

$$\rho = \beta \frac{\sigma_f}{\sigma_s}$$

(3)

In this article, we focus primarily on the IGC because the above approach, which imputes fathers’ earnings with fathers’ occupations, compresses the variance of fathers’ earnings, which in turn inflates our estimates of the IGE. However, we also test the robustness of our results to using the IGE. Estimates of the IGE and IGC for each country are listed in Andrews and Leigh (2008).

Among employed fathers and sons, three factors drive intergenerational mobility: (i) sons working in different occupations from their fathers (inter-occupational mobility); (ii) sons working in the same occupation but with lower or higher earnings than their fathers (intra-occupational mobility) and (iii) changes in the average earnings of occupations over time. The method employed here will capture inter-occupational mobility (factor i), but will only capture part of intra-occupational mobility (factor ii), since fathers in the same occupation are assigned the same wage. Moreover, this approach will not take account of changes in the average earnings of occupations over time (factor iii). To gauge the likely importance of this issue, we use microdata from US Censuses to calculate the mean age-adjusted log earnings of men aged 25–54 in 192 occupations. The correlation between an occupation’s mean earnings in 1970 and 2000 was 0.71. While this correlation is reassuringly high, there is still a possibility that our approach will mis-estimate the true level of intergenerational mobility.

To help validate our estimates, we calculated the correlation between our estimated levels of intergenerational mobility and those published elsewhere. With three common countries, the correlation between our IGCs and those of Jäntti et al. (2006) is 0.70. With four common countries, the correlation between our IGEs and those of Solon (2002) is 0.77.

### III. Inequality and Social Mobility

To measure income inequality, we use the Gini coefficient, a measure of the income gap between two randomly selected individuals in the population. Where possible, we utilize the Luxembourg Income Study (LIS) – probably the most reliable source for cross-country estimates of income inequality (Atkinson, 2004). For the four countries in our sample that do not participate in the LIS, we use the highest quality estimate from the Deininger and Squire (1996) database. We use the closest estimate to the year 1975 – the likely period when the parents in the sample were making decisions about investments in their children’s human capital.
(the sons in our sample were aged 25–54 in 1999, so they were aged 1–30 in 1975). For Australia, the Czech Republic, Hungary, Poland, Slovakia and Spain, LIS estimates of income inequality are only available for the 1980s and 1990s, though estimates for the early 1970s are available in the Deininger–Squire database. For these countries, a tradeoff exists between using the most appropriate time period and the highest-quality inequality estimates. As such, we also conduct our analysis using data sourced solely from each dataset.

The relationship between inequality and intergenerational mobility for all 16 countries in our sample is depicted in Fig. 1. While a positive relationship between inequality and intergenerational mobility is discernible, this is not statistically significant at conventional levels. However, for the six former Warsaw Pact countries in our sample, which were not market economies in the 1970s, it may be unreasonable to draw a link between inequality in the 1970s and intergenerational mobility between the 1970s and late 1990s. (Recall that the theoretical explanations suggesting a relationship between inequality and social mobility include private expenditure on education, political donations and median voter models. These are more likely to apply in capitalist democracies than in Communist countries.)

As such, Fig. 2 excludes the six former Warsaw Pact countries. The coefficient on inequality becomes statistically significant at the 1% level and the magnitude of the coefficient almost doubles. In addition, the $R^2$ rises to 0.71. To account for the possibility that these statistical results are gaining leverage from the inclusion of Chile, we excluded this country from the sample and re-estimated the model. Even when we dropped Chile, however, the Gini coefficient was significant at the 6% level.

We performed a series of robustness checks to confirm these results. First, we re-estimated our results using Gini coefficients sourced exclusively from the Deininger–Squire database, which contains inequality measures of lower quality but of the appropriate vintage (the database has pre-1980 inequality measures for all but two countries, Russia and Latvia). Second, we used only Gini coefficients from the LIS database, whose estimates are generally regarded as higher quality, but are not necessarily derived from surveys in the 1970s. Third, we re-specified the dependent variable to the IGE instead of the IGC. Fourth, we added controls for the rate of return to education in the 1970s, and the log of GDP per-capita in 1975 (sources for these variables are described in Andrews and Leigh, 2008). In each of these robustness checks, and in combinations of them, the results are very close to those above, both in magnitude and statistical significance. The only exception is that when we use solely LIS inequality estimates, the coefficient on the IGC/IGE is not statistically significant at conventional levels. By contrast, when we use Gini coefficients sourced exclusively from the Deininger–Squire database, the relationship between inequality and the IGC/IGE is statistically significant at the 5% level, even including the former Warsaw Pact countries.

IV. Conclusion

Using cross-country data, we find that sons who grew up in more unequal countries in the 1970s were less
likely to have experienced social mobility by 1999. Across countries, our estimates suggest that a 10-point rise in the Gini coefficient is associated with a 0.07–0.13 increase in the intergenerational earnings correlation. Moving from rags to riches is harder in more unequal countries.

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References

Solon, G. (2002) Cross-country differences in intergenera-