Online Appendix Strangers in the Homeland? The Academic Performance of Children of Return Migrants in Mexico

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Full Regression Tables

Table 1: Full table of OLS coefficients for Mexican sample

	Dependent variable:							
	Reading	Math	Science	Reading	Math	Science		
	(1)	(2)	(3)	(4)	(5)	(6)		
treat	9.100	21.000***	11.000^{\dagger}	10.000^\dagger	21.000***	12.000*		
	(6.300)	(5.500)	(5.500)	(5.900)	(5.800)	(5.500)		
mom_ed	5.000***	4.700***	4.400***	1.100^{*}	1.600**	1.500**		
	(0.410)	(0.400)	(0.400)	(0.560)	(0.550)	(0.500)		
dad_ed	4.600***	3.400***	4.000***	0.470	0.230	0.860		
	(0.490)	(0.420)	(0.380)	(0.630)	(0.580)	(0.550)		
female	19.000***	-10.000***	-7.200***	18.000***	-11.000***	-7.700***		
	(1.300)	(1.400)	(1.300)	(1.200)	(1.300)	(1.400)		
$early_ed0-1$	25.000***	20.000***	19.000***	20.000***	15.000***	15.000***		
	(3.200)	(3.100)	(2.900)	(3.200)	(3.100)	(2.900)		
$early_ed1+$	26.000***	22.000***	20.000***	21.000***	18.000***	16.000***		
	(3.000)	(2.800)	(2.700)	(3.000)	(2.800)	(2.700)		
$cultural_pos$	4.700***	3.400***	5.000***	-0.420	-1.500	-0.075		
	(0.780)	(0.700)	(0.740)	(1.100)	(1.100)	(1.100)		
$home_ed$	11.000***	9.100***	8.900***	-1.200	-1.500	-1.500		
	(0.750)	(0.990)	(0.730)	(1.400)	(1.700)	(1.400)		
age	20.000***	17.000***	17.000***	19.000***	16.000***	17.000***		
	(2.600)	(2.300)	(2.500)	(2.600)	(2.400)	(2.600)		
year2015	-4.200	-9.300***	-2.900	-3.600	-8.500***	-2.400		
	(2.900)	(2.600)	(2.500)	(2.900)	(2.500)	(2.500)		
year2018				-12.000***	-13.000***	-4.400		
	(3.100)	(2.900)	(3.000)	(3.000)	(2.900)	(3.000)		
non_urban				-19.000***	-12.000***	-14.000***		
				(2.500)	(2.400)	(2.300)		
wealth				-9.000**	-12.000***	-13.000***		
				(3.400)	(3.200)	(3.300)		
$home_pos$				17.000***	18.000***	19.000***		
				(4.600)	(4.700)	(4.600)		
ict_res				2.400^*	2.700**	2.400**		
				(1.100)	(1.000)	(0.910)		
escs				7.300*	5.700	5.500^{\dagger}		
				(3.400)	(3.700)	(2.900)		
parent_isei				()	()	()		
	,			(0.086)	(0.081)	(0.068)		
Constant	71.000^{\dagger}	127.000***	115.000**	120.000**	162.000***	148.000***		
	(40.000)	(37.000)	(40.000)	(41.000)	(38.000)	(41.000)		
Observations	$41,\!175$	$41,\!175$	41,175	39,302	39,302	39,302		

Note: $\dagger p < 0.1$; *p < 0.05; **p < 0.01; ***p < 0.001. OLS estimates comparing children of return migrants in Mexico to children in Mexico. All models cluster standard errors at the school level and incorporate sampling weights. Source: PISA data from 2012, 2015, and 2018

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Table 2: Full table of OLS coefficients for U.S. sample

	Dependent variable:							
	Reading	Math	Science	Reading	Math	Science		
	(1)	(2)	(3)	(4)	(5)	(6)		
treat	-45.000***	-15.000^*	-38.000***	-69.000***	-31.000**	-65.000***		
	(8.700)	(7.000)	(7.700)	(12.000)	(12.000)	(12.000)		
$\operatorname{mom_ed}$	-1.700	-0.059	-1.200	-9.500**	-4.800	-8.900^*		
	(2.300)	(2.200)	(1.900)	(3.500)	(3.200)	(3.700)		
dad_ed	1.900	0.870	1.700	-1.600	-2.400	-2.700		
	(2.300)	(2.100)	(2.200)	(4.000)	(3.600)	(3.900)		
female	24.000***	-9.900	-4.100	23.000*	-5.000	-2.300		
	(6.400)	(6.300)	(6.700)	(9.000)	(8.000)	(9.400)		
$early_ed0-1$	-2.200	-0.280	-5.900	-55.000*	-37.000	-40.000		
	(17.000)	(15.000)	(17.000)	(26.000)	(25.000)	(29.000)		
$early_ed1+$	8.300	6.600	3.100	14.000	11.000	8.800		
	(11.000)	(8.900)	(11.000)	(12.000)	(10.000)	(13.000)		
cultural_pos	8.000*	6.300*	8.100*	-20.000*	-15.000*	-15.000^{\dagger}		
_	(3.600)	(3.200)	(3.500)	(8.300)	(6.800)	(8.000)		
home ed	9.700**	8.100**	6.400°	-12.000	-12.000	-11.000		
_	(3.700)	(3.000)	(3.700)	(9.700)	(7.700)	(8.300)		
age	22.000^{\dagger}	24.000*	16.000	36.000*	31.000*	27.000^{\dagger}		
G	(12.000)	(11.000)	(11.000)	(16.000)	(13.000)	(14.000)		
year2015	-3.300	2.400	17.000	-52.000****	-26.000^{\dagger}	-38.000**		
	(12.000)	(11.000)	(12.000)	(14.000)	(15.000)	(14.000)		
year2018	,	,	,	-13.000	$-2.200^{'}$	-9.100		
	(12.000)	(11.000)	(15.000)	(15.000)	(15.000)	(15.000)		
non urban	,	,	,	-6.600	-2.600	-1.200		
_				(9.500)	(9.600)	(9.400)		
wealth				-69.000**	-49.000**	-55.000*		
				(24.000)	(19.000)	(21.000)		
home pos				83.000**	66.000**	70.000*		
_				(31.000)	(25.000)	(29.000)		
ict_res				-3.300	-5.000	-12.000		
				(12.000)	(9.600)	(9.600)		
escs				36.000°	26.000	38.000*		
				(19.000)	(16.000)	(19.000)		
parent_isei				(0.4=0)	(0.600)	(0.400)		
a	110.000	5 0.000	015 000	(0.470)	(0.390)	(0.480)		
Constant	119.000	70.000	217.000	9.400	32.000	160.000		
	(195.000)	(169.000)	(173.000)	(265.000)	(202.000)	(217.000)		
Observations	1,391	1,391	1,391	896	896	896		

Note: $\dagger p < 0.1$; *p < 0.05; **p < 0.01; ***p < 0.001. OLS estimates comparing children of return migrants in Mexico to children of Spanish-speaking immigrants in the U.S. All models cluster standard errors at the school level and incorporate sampling weights. Source: PISA data from 2012, 2015, and 2018

.2 Sensitivy to Unobserved Confounders

Important variables might be correlated with both immigration (and return migration) as well as test scores. For the Mexican comparison, results are already close to 0, so this section focuses on

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the U.S. comparison. How strong would an unobserved confounder need to be in order to create a null effect for migration to Mexico for the 0.5 generation? I use the omitted variable bias (OVB) analysis tools of the sensemakr package (Cinelli & Hazlett, 2020) to help answer this question.

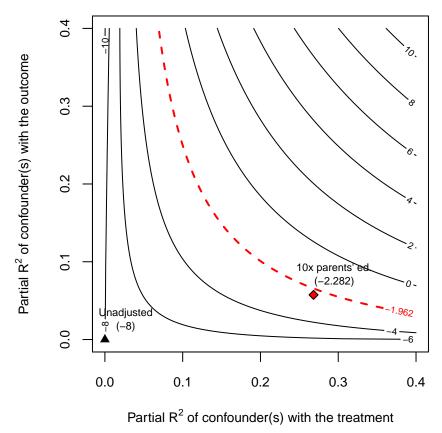


Figure 1: Contour plot of possible confounders of reading scores for the U.S. comparison. Contour lines represent t-values for the return migration coefficient in an OLS model for reading scores with eight additional covariates and hypothetical levels of confounding. "Unadjusted" shows the t-value of the immigration coefficient with no confounding. "10x parents' ed." shows the t-value of the immigration coefficient after accounting for a hypothetical confounder ten times as strong as the join effects of parents' education.

I find that an unobserved confounder would need to explain more than 19 percent of the residual variance of both return migration and reading scores in order to bring the estimate of immigration down to 0. Figure 1 shows that a confounder even ten times as strong as the joint effect of mother's and father's education would still not produce an insignificant coefficient. The t-value in the original model is -8. This plot shows that a confounder even five times as strong as mother's education (mom_ed) would only reduce this t-value to -2.3, still significant at the $\alpha=5$ percent level. Results are similar for science and math scores.

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References

Cinelli, C., & Hazlett, C. (2020). Making sense of sensitivity: Extending omitted variable bias. $Journal\ of\ the\ Royal\ Statistical\ Society:\ Series\ B\ (Statistical\ Methodology),\ 82(1),\ 39-67.$ https://doi.org/10.1111/rssb.12348