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Geography, Human Capital and Urbanization: A Regional Analysis

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Abstract

Using subnationally representative data for 679 regions from 64 low- and middle-income countries, we document that geography, especially trade-related features, explains a substantial share of within-country variation in the level of human capital. We further show that a large part of the explanatory power can be attributed to the close interrelation between urbanization and concentration of human capital. These results bring together studies that identify human capital as an important determinant of regional development and the literature on economic geography that emphasizes geography's role in shaping the spatial distribution of urbanization and economic activity.

JEL CLASSIFICATION: O11, R11, E24

KEYWORDS: Geography, Human Capital, Urbanization

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1 Introduction

Geography's influence on national and subnational economic development is well documented. The relative importance of individual channels through which geography affects development, however, is much less clear. In this study we investigate if and how geography influences the regional level of human capital, a potentially important proximate determinant of economic prosperity (e.g., Acemoglu and Dell (2010); Gennaioli et al. (2013)).

For our empirical analysis, we draw on subnationally representative survey data from the Demographic and Health Surveys (DHS) Program and construct various proxies for human capital at the regional level, such as average years of schooling or prevalence of stunting. In total, our dataset encompasses 679 regions from 64 low-and middle income countries. To assess the explanatory power of geographic variables regarding within-country differences in the level of human capital, we employ a cross-sectional OLS regression approach in which we control for country-fixed effects. For parsimony, we include only nine geographic characteristics in our main regression setup. Following Henderson et al. (Forthcoming), this set of features encompasses five agriculture-related variables (caloric suitability, temperature, precipitation, elevation and absolute latitude) and four trade-associated characteristics (accessibility by inland waterways, share of a region's area that lies within 25 kilometers of an ocean, distance to coastline and terrain ruggedness). Overall, the nine geographic covariates explain a meaningful 23% of the within-country sample variation in average years of schooling. These findings suggest that an important channel through which geography influences regional development is via its effect on human capital.

A potentially important intermediary factor linking geography to human capital is urbanization. The location of these urban areas is strongly influenced by geography-related productivity advantages (Henderson et al., Forthcoming). Additionally, human capital is highly concentrated in the urban, skill-intensive, sectors in developing countries. As these sectors are primarily located in urban areas, this generates a rural-urban gap in the level of human capital (Young, 2013) and suggests that a large part of the explanatory power of geography may be accounted for by its effect on urbanization. Our regression results provide empirical support for this notion. Overall, 65% of the R-squared produced by the nine geographic characteristics is accounted for by the partial correlation of these covariates with urbanization.

Our study contributes to the large body of research that analyzes the effects of geography on economic development at the national (Mellinger et al., 2000; Nunn and Puga, 2012)

and subnational level (Rappaport and Sachs, 2003; Dell et al., 2009; Mitton, 2016). Closely related to our paper is further the literature that documents the existence of a large rural-urban gap in productivity in developing countries (Gollin et al., 2014; Young, 2013). By documenting that geography’s effects on the regional level of human capital is to a large part accounted for by its interrelation with urbanization, our paper provides a link between these studies and the branch of economic geography that identifies geography as a centrally important determinant of the spatial distribution of urbanization and economic activity (e.g., Davis and Weinstein (2002); Rappaport and Sachs (2003)). Within the latter field, the study of Henderson et al. (Forthcoming), showing that geography is highly predictive of economic activity at the local level, is most closely related to ours.

The remainder of this paper is organized as follows: In the next section, we present our empirical strategy; the data are in described in Section 3 and the results discussed in Section 4. A conclusion is offered in Section 5.

2 Empirical Strategy

We assess the explanatory power of geographic characteristics using the following cross-sectional OLS regression:

$$y_{i,c} = \psi' \mathbf{G}_{i,c} + \mu_i + \varepsilon_{i,c}, \quad (1)$$

where the dependent variable $y_{i,c}$ represents different proxies for the level of human capital in region i and country c . Vector \mathbf{G} encompasses geographic characteristics. Following Henderson et al. (Forthcoming), we classify these into two groups: Characteristics that primarily influence agricultural productivity and characteristics particularly relevant for trade. In our main analysis, the former category includes precipitation, temperature, agricultural suitability, elevation and absolute latitude; the latter encompasses distance to coastline, the percentage of a region’s surface area that lies within 25 kilometer of the coastline, accessibility by inland waterways (captured by the Strahler index), as well as terrain ruggedness.

All regressions control for the landmass encompassed by the individual regions as well as country fixed effects (μ_i). Our estimates therefore capture reduced-form effects of geography within countries. The idiosyncratic error term is symbolized by $\varepsilon_{i,c}$; the standard errors are clustered at the country level.

3 Data and Descriptive Analysis

Human Capital

The proxies for human capital are drawn from the Demographic and Health Surveys (DHS) Program. The surveys are representative at sub-national reporting areas which typically correspond to first-level administrative country subdivisions or groups thereof. To construct regional-level variables we average survey responses over regions using the sample weights provided by the DHS. Specifically, we compute average years of schooling, the percentage of households with access to improved drinking water sources and toilet facilities as well as prevalence of stunting in children under five. The representativeness of the DHS data further allows us to determine the urbanization rate for each region, i.e., the share of total population residing in areas that are defined as urban. In total, our dataset encompasses 679 regions from 64 low-and middle-income countries.¹ The spatial distribution of these regions is depicted in Figure 1.

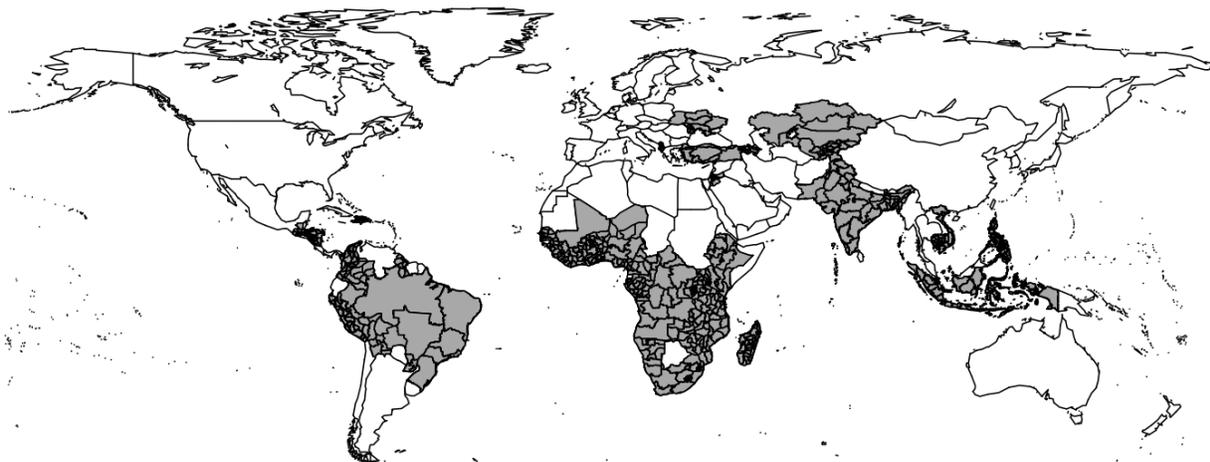


Figure 1: Sample coverage. Regions shaded gray are included in our dataset.

Geography

Long-run (1960–1990) temperature, precipitation and elevation are drawn from Hijmans et al. (2005), terrain ruggedness is taken from Nunn and Puga (2012), and the caloric suitability index from Galor and Ozak (2015). Based on information provided by www.naturalearthdata.com, we compute a region’s average distance to the coastline and its

¹When multiple survey waves exist, we use the most recent one. Tables A.1–A.2 report details on countries and survey waves included in our analysis as well as data construction processes.

share of total area that lies within 25 kilometers of an ocean. Accessibility by inland waterways is measured by the Strahler order of streams. For the empirical analysis, we standardize all variables to a mean of zero and a standard deviation of one. The summary statistics are depicted in Appendix A, Table A.3.

4 Results

We evaluate the explanatory power of geography by looking at the within-country R-squared, i.e., the share of variation our model can account for once country fixed effects and surface areas of regions have been partialled out. Below, we first look at reduced-form relationships and then investigate the importance of urbanization as an intermediary factor.

4.1 Geography and Human Capital

The results of Table 1 column (1) show that the five agricultural covariates explain about 9% of the within-country variation in average years of schooling. Of the individual variables, temperature exerts the strongest and statistically most significant effect. This is consistent with studies that document the negative influence of temperature on regional income per capita (e.g., Dell et al. (2009)). The point estimate of -0.881 implies that a one-standard deviation increase in temperature reduces average years of schooling by 0.881 standard deviations. In column (2) we replace the agriculture variables with trade-associated covariates. Compared to the former, the explanatory power of the latter is substantially higher. More than 15% of subnational variation in average years of schooling is explained by the four trade-related geographic characteristics. Among these, access to coast, measured as the percentage of a region's area that lies within 25 kilometers of an ocean, is economically the most significant. A one-standard deviation increase in this covariate raises average years of schooling by 0.2 standard deviations. This results ties in with the literature that identifies coastal access as an important determinant of the spatial distribution of income per capita (e.g., Mellinger et al. (2000); Rappaport and Sachs (2003)). In column (3) we include both agriculture- and trade-related variables in the regression setup. Combined, the nine characteristics explain a remarkable 23.2% of the relevant sample variation.

We next investigate geography's influence on proxies for health, a second important dimen-

Table 1: Geography and Human Capital

Dependent Variable	Average Years of Schooling		Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted	
	(1)	(2)	(3)	(4)	(5)	(6)
Caloric suitability	0.068 (0.051)		0.176*** (0.058)	0.065 (0.106)	0.101 (0.071)	-0.130 (0.080)
Temperature	-0.881*** (0.180)		-0.783*** (0.177)	-0.097 (0.209)	-0.440** (0.186)	0.855*** (0.275)
Precipitation	0.012 (0.055)		0.012 (0.045)	-0.135* (0.082)	-0.006 (0.069)	0.108** (0.049)
Elevation	-0.571*** (0.111)		-0.333*** (0.118)	0.070 (0.132)	-0.187 (0.117)	0.557*** (0.175)
Absolute latitude	-0.356* (0.183)		-0.299** (0.145)	0.134 (0.181)	-0.029 (0.144)	0.375* (0.196)
Strahler index		0.056*** (0.018)	0.063*** (0.019)	0.086** (0.043)	0.036 (0.031)	-0.089*** (0.019)
% Area within 25km of coastline		0.200*** (0.033)	0.200*** (0.028)	0.221*** (0.058)	0.214*** (0.042)	-0.274*** (0.042)
Average distance to coastline		-0.099 (0.062)	-0.127** (0.056)	-0.061 (0.083)	-0.155*** (0.048)	-0.017 (0.078)
Ruggedness		0.014 (0.036)	-0.113** (0.050)	-0.116 (0.074)	-0.023 (0.051)	0.058 (0.061)
Observations	679	679	679	671	662	600
Within-country R ²	0.090	0.154	0.232	0.081	0.143	0.183

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. The table reports standardized beta coefficients. Standard errors clustered at the country level. All regressions control for country fixed effects and surface area of regions.

sion of human capital. For brevity, we only report full specifications in which we control for both agriculture- and trade-related covariates. As depicted in columns (4)–(5), geography explains a meaningful fraction of the relevant sample variation in access to drinking water (8.1%) and sanitation (14%), both of which are important determinants of health (Duflo et al., 2015; Alsan and Goldin, Forthcoming). The results are similar when we look at the share of stunted children in population aged five and below (column (6)). Stunting is an indicator for health during childhood which, in turn, strongly influences the level of human capital accumulated as an adult (e.g., Bleakley (2007); Shah and Steinberg (2017)).

Taken together, the results of Table 1 document that variation in geography—particularly trade-associated geography—is of great importance in accounting for regional differences in the level of human capital.² This finding consolidates studies that link human capital to regional income differences (Gennaioli et al., 2013) with papers that identify geography

²The explanatory power increases slightly if we include further control variables (Appendix C.1).

as central for regional development (e.g., Dell et al. (2009)). Next, we investigate the importance of urbanization as an intermediary factor underlying the reduced-form effects of geography.

4.2 Geography, Urbanization and Human Capital

In Table 2, we re-run the regressions of Table 1 but additionally include the urbanization rate as an explanatory variable. We then analyze how much of geography’s explanatory power can be accounted for by urbanization. That is, we compute the share of geography’s total explanatory power regarding variation in the outcome variables that is absorbed by the partial correlation with urbanization.³ The results document the close relationship between urbanization and proximate determinants of development in low- and middle-income countries. With urbanization included, the predictive power of the regression model grows substantially compared to Table 1. Averaged across all columns, the within-country R-squared increases from 16% to 48%. A likely driver underlying this result is the structural urban-rural gap in human capital (e.g., Young (2013)).⁴

The last row in Table 2 reports the share of explanatory power of geographic variables accounted for by urbanization. With average years of schooling as dependent variable, 38% of the explanatory power produced by agriculture-related covariates is absorbed by the partial correlation with urbanization rates (column (1)). For the set of trade-associated variables, the share accounted for by urbanization increases to a staggering 81%. This implies that the bulk of explanatory power of the trade-related geographic characteristics can be explained by the close relationship between urbanization and human capital: Trade geography strongly influences the extent to which a region is urbanized. This, in turn, influences its average level of human capital.⁵ When incorporating both sets of geographic characteristics into the regression setup, 64.9% of their joint explanatory power is absorbed

³Formally, this share is given by:

$$\text{Within } R^2 \text{ accounted for by urbanization} = \frac{R_g^2 - (R_{u+g}^2 - R_u^2)}{R_g^2} = 1 - \frac{(R_{u+g}^2 - R_u^2)}{R_g^2},$$

where R_g^2 represents the within-country R-squared produced by the geographic characteristics, R_u^2 the within-country R-squared produced by urbanization and R_{u+g}^2 the joint explanatory power of urbanization and geographic features.

⁴Table B.1 documents that this gap is also observable at the individual level in our survey data.

⁵Urbanization and human capital are two reinforcing processes. Disentangling directions of causality between the two variables is not possible within the framework of our analysis. The point estimates for urbanization should therefore not be interpreted as causal effects.

Table 2: Geography, Urbanization and Human Capital

Dependent Variable	Average Years of Schooling			Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted
	(1)	(2)	(3)	(4)	(5)	(6)
Share urban population	0.374*** (0.021)	0.353*** (0.022)	0.345*** (0.022)	0.446*** (0.041)	0.387*** (0.035)	-0.312*** (0.032)
Caloric suitability	0.070 (0.043)		0.119** (0.047)	-0.009 (0.087)	0.045 (0.058)	-0.067 (0.072)
Temperature	-0.558*** (0.129)		-0.525*** (0.123)	0.242 (0.214)	-0.166 (0.184)	0.611** (0.252)
Precipitation	0.059 (0.041)		0.056 (0.036)	-0.079 (0.068)	0.044 (0.069)	0.053 (0.040)
Elevation	-0.277*** (0.078)		-0.189*** (0.073)	0.258* (0.145)	-0.042 (0.140)	0.419*** (0.150)
Absolute latitude	-0.304** (0.134)		-0.277** (0.118)	0.162 (0.149)	-0.013 (0.121)	0.312* (0.189)
Strahler index		0.020** (0.009)	0.029*** (0.008)	0.047* (0.028)	0.002 (0.022)	-0.060*** (0.016)
% Area within 25km of coastline		0.074*** (0.020)	0.079*** (0.019)	0.065 (0.058)	0.077* (0.041)	-0.149*** (0.047)
Average distance to coastline		-0.060 (0.048)	-0.084** (0.042)	-0.004 (0.073)	-0.102** (0.047)	-0.050 (0.072)
Ruggedness		0.062** (0.027)	-0.039 (0.035)	-0.020 (0.066)	0.062 (0.041)	0.006 (0.052)
Observations	679	679	679	671	662	600
Within-country R ²	0.560	0.534	0.586	0.323	0.445	0.337
% Within R ² accounted for by urbanization	0.380	0.811	0.649	0.704	0.716	0.674

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. The table reports standardized beta coefficients. Standard errors clustered at the country level. All regressions control for country fixed effects and surface area of regions. % *Within R² accounted for by urbanization* represents the share of the total explanatory power of the geographic variables that is absorbed by the partial correlation with the urban population share (see the main text for more details).

by the partial correlation with urbanization. The pattern of results is very similar when we look at health-related outcomes. Averaged across columns (4)–(6), the share of explanatory power accounted for by the correlation with urbanization is 70%.

We conduct a number of robustness checks in order to document the stability of our results. Appendix C shows that the results remain qualitatively unaltered if we drop regions that are home to capital cities or drop the five most and five least urbanized countries.

5 Conclusion

In this study, we document that both agriculture- and trade-associated geographic characteristics are important in accounting for regional differences in the level of human capital. A large part of the explanatory power, particularly for the trade-related variables, is absorbed by the correlation with urbanization. This implies that the effects of geography on the concentration of human capital and urbanization are closely interlinked and is consistent with the existence of a urban-rural gap in human capital.

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Web Appendix

A Data Construction

Table A.1: DHS data used to construct dataset

Country	Phase	Year	Country	Phase	Year
Albania	2008-09	DHS-V	Lesotho	2014	DHS-VII
Angola	2011	DHS-VI	Liberia	2013	DHS-VI
Armenia	2010	DHS-VI	Madagascar	2008-09	DHS-V
Azerbaijan	2006	DHS-V	Malawi	2010	DHS-VI
Bangladesh	2014	DHS-VII	Mali	2012-13	DHS-VI
Benin	2011-12	DHS-VI	Moldova	2005	DHS-V
Bolivia	2008	DHS-V	Mozambique	2011	DHS-VI
Brazil	1996	DHS-III	Namibia	2013	DHS-VI
Burkina Faso	2010	DHS-VI	Nicaragua	2001	DHS-IV
Burundi	2010	DHS-VI	Niger	2012	DHS-VI
Cambodia	2014	DHS-VII	Nigeria	2013	DHS-VI
Cameroon	2011	DHS-VI	Pakistan	2012-13	DHS-VI
Central African Republic	1994-95	DHS-III	Paraguay	1990	DHS-II
Colombia	2010	DHS-VI	Peru	2009	DHS-V
Congo	2011-12	DHS-VI	Philippines	2008	DHS-V
Congo Democratic Republic	2013-14	DHS-VI	Rwanda	2014-15	DHS-VII
Cote d'Ivoire	2011-12	DHS-VI	Senegal	2010-11	DHS-VI
Dominican Republic	2013	DHS-VI	Sierra Leone	2013	DHS-VI
Ethiopia	2011	DHS-VI	South Africa	1998	DHS-III
Gabon	2012	DHS-VI	Swaziland	2006-07	DHS-V
Ghana	2014	DHS-VII	Tajikistan	2012	DHS-VI
Guatemala	1998-99	DHS-IV	Tanzania	2010	DHS-VI
Guinea	2012	DHS-VI	Gambia	2013	DHS-VI
Guyana	2009	DHS-V	Timor-Leste	2009-10	DHS-VI
Haiti	2012	DHS-VI	Togo	2013-14	DHS-VI
Honduras	2011-12	DHS-VI	Turkey	2003	DHS-IV
India	2005-06	DHS-V	Uganda	2011	DHS-VI
Indonesia	2012	DHS-VI	Ukraine	2007	DHS-V
Jordan	2012	DHS-VI	Uzbekistan	1996	DHS-III
Kazakhstan	1999	DHS-IV	Vietnam	2002	DHS-IV
Kenya	2014	DHS-VII	Zambia	2013-14	DHS-VI
Kyrgyz Republic	2012	DHS-VI	Zimbabwe	2010-11	DHS-VI

Table A.2: DHS variable construction

Variable Name	DHS variable(s)	Construction	Explanation
Average years of schooling	v133		
Share Secondary Education or Higher	v106	$v106 \geq 2$	
Modern Sector Occupation Share	v717	$v717 \in \{1,2,3,7,8,9\}$	Following occupations do <i>not</i> belong to the modern (industry and service) sector: (4) Agricultural - self employed; (5) Agricultural - employee. Agricultural categories also include fishermen, foresters and hunters. For this variable we exclude all individuals without employment as well as occupation in household and domestic.
Urban Share	v102	V102=1	
Share Improved Drinking Water	v133	Includes the categories: piped into dwelling, piped to yard/plot, protected spring, protected well, HJH=public tap/standpipe, rainwater, tube well or borehole	See WHO and UNICEF (2006)
Share Improved Sanitation Facilities	v116	Includes (non-shared) categories: flush - to piped sewer system, flush - to septic tank, flush - to pit latrine, flush - to somewhere else, flush - don't know where, pit latrine - ventilated improved pit, pit latrine - with slab, composting toilet	See WHO and UNICEF (2006)
Share Children Stunted	hw70 < -2	height-for-age z-score is below -2	Living children between 0-59 month (Rutstein and Rojas, 2006, p.143)

Table A.3: Descriptive Statistics Key Variables

Variable	Raw Mean (1)	Mean (2)	Std. Dev. (3)	Within Std. Dev. (4)	Min-Max Gap (5)	Within Min-Max Gap (6)	Obs. (7)
Human Capital							
Average years of schooling	6.622	0	1	0.433	4.582	2.828	679
Share improved drinking water	0.761	0	1	0.672	4.083	3.232	671
Share improved sanitation facilities	0.475	0	1	0.529	3.232	2.883	662
Share children stunted	0.311	0	1	0.600	5.621	4.205	600
Share urban population	0.394	0	1	0.811	3.631	3.631	679
Share modern occupation	0.670	0	1	0.621	3.433	3.222	657
Geography							
Caloric suitability	1,485	0	1	0.441	5.055	4.176	679
Temperature	21.94	0	1	0.459	5.462	4.655	679
Precipitation	112.4	0	1	0.527	6.963	5.807	679
Elevation	702.3	0	1	0.668	5.700	5.473	679
Absolute latitude	15.21	0	1	0.215	4.453	2.002	679
Strahler index	1.196	0	1	0.926	6.883	6.883	679
Average distance to coastline	354.645	0	1	0.459	5.379	4.472	679
% Area within 25km of coastline	0.182	0	1	0.709	3.237	3.237	679
Ruggedness	12.49	0	1	0.631	6.407	6.220	679

Notes: All summary statistics except for 'Raw Mean' generated using standardized variables (mean zero and standard deviation one). 'Within' refers to within-country values. 'Min-Max Gap' represents the absolute difference between the lowest and the highest value.

B Supporting Results

Table B.1: Rural-Urban Education Gap – Individual Level

Dependent Variable	Years of Schooling	
	(1)	(2)
Urban status	3.168*** (0.088)	2.703*** (0.077)
Country FE	yes	yes
Region FE	no	yes
Observations	992,322	992,322
Within-country R ²	0.363	0.406

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. Standard errors clustered at the DHS cluster level. Variables are in raw form (i.e., not standardized). *Urban* represents a dummy variable that indicates whether a respondent resides in an urban area or not.

C Robustness Tests

C.1 Extensive Set of Geographic Variables

Table C.1: Geography and Human Capital

Dependent Variable	Average Years of Schooling			Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted
	(1)	(2)	(3)	(4)	(5)	(6)
Within R ²	0.104	0.156	0.262	0.100	0.161	0.200
Agriculture related geography variables	yes	no	yes	yes	yes	yes
Trade related geographic variables	no	yes	yes	yes	yes	yes
Observations	679	679	679	671	662	600

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. The table reports standardized beta coefficients. Standard errors clustered at the country level. All regressions control for country fixed effects and surface area of regions. Agriculture-related geography variables include precipitation, temperature, agricultural suitability, elevation, absolute latitude, length of growing season and biomes fixed effects. Trade related geography variables include distance to coastline, the percentage of a region that lies within 25 kilometers of the coastline, the Strahler index, terrain ruggedness, a dummy variable that indicates the existence of a natural harbor and a dummy variable that indicates the intersection with a big lake.

Table C.2: Geography, Urbanization and Human Capital

Dependent Variable	Average Years of Schooling			Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted
	(1)	(2)	(3)	(4)	(5)	(6)
Within R ²	0.571	0.534	0.606	0.341	0.459	0.351
% Within R ² accounted for by urbanization	0.358	0.811	0.611	0.585	0.662	0.634
Agriculture related geography variables	yes	no	yes	yes	yes	yes
Trade related geographic variables	no	yes	yes	yes	yes	yes
Observations	679	679	679	671	662	600

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. The table reports standardized beta coefficients. Standard errors clustered at the country level. All regressions control for country fixed effects and surface area of regions. Agriculture-related geography variables include precipitation, temperature, agricultural suitability, elevation, absolute latitude, length of growing season and biomes fixed effects. Trade related geography variables include distance to coastline, the percentage of a region that lies within 25 kilometers of the coastline, the Strahler index, terrain ruggedness, a dummy variable that indicates the existence of a natural harbor and a dummy variable that indicates the intersection with a big lake. % Within R² accounted for by urbanization represents the share of the total explanatory power of the geographic variables that is absorbed by the partial correlation with the urban population share (see the main text for more details).

C.2 Dropping Capital Regions

Table C.3: Geography and Human Capital

Dependent Variable	Average Years of Schooling			Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted
	(1)	(2)	(3)	(4)	(5)	(6)
Within R ²	0.107	0.138	0.220	0.046	0.108	0.153
Agriculture related geography variables	yes	no	yes	yes	yes	yes
Trade related geographic variables	no	yes	yes	yes	yes	yes
Observations	610	610	610	604	597	540

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. Standard errors clustered at the country level. All variables are in standardized z-scores. Agriculture related geography variables include precipitation, temperature, agricultural suitability, elevation and absolute latitude. Trade related geography variables include distance to coastline, the percentage of a region that lies within 25 kilometer of the coastline, the Strahler index and terrain ruggedness.

Table C.4: Geography, Urbanization and Human Capital

Dependent Variable	Average Years of Schooling			Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted
	(1)	(2)	(3)	(4)	(5)	(6)
Within R ²	0.394	0.366	0.442	0.170	0.286	0.228
% Within R ² accounted for by urbanization	0.220	0.601	0.400	0.368	0.543	0.469
Agriculture related geography variables	yes	no	yes	yes	yes	yes
Trade related geographic variables	no	yes	yes	yes	yes	yes
Observations	610	610	610	604	597	540

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. The table reports standardized beta coefficients. Standard errors clustered at the country level. All regressions control for country fixed effects and surface area of regions. Agriculture-related geography variables include precipitation, temperature, agricultural suitability, elevation, and absolute latitude. Trade-related geography variables include distance to coastline, the percentage of a region that lies within 25 kilometers of the coastline, the Strahler index and terrain ruggedness. % *Within R² accounted for by urbanization* represents the share of the total explanatory power of the geographic variables that is absorbed by the partial correlation with the urban population share (see the main text for more details).

C.3 Dropping the 5 Least and the 5 Most Urbanized Countries from the Sample

Table C.5: Geography and Human Capital

Dependent Variable	Average Years of Schooling			Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted
	(1)	(2)	(3)	(4)	(5)	(6)
Within R ²	0.118	0.162	0.253	0.109	0.153	0.243
Agriculture related geography variables	yes	no	yes	yes	yes	yes
Trade related geographic variables	no	yes	yes	yes	yes	yes
Observations	552	552	552	544	535	473

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. The table reports standardized beta coefficients. Standard errors clustered at the country level. All regressions control for country fixed effects and surface area of regions. Agriculture-related geography variables include precipitation, temperature, agricultural suitability, elevation, and absolute latitude. Trade-related geography variables include distance to coastline, the percentage of a region that lies within 25 kilometers of the coastline, the Strahler index and terrain ruggedness.

Table C.6: Geography, Urbanization and Human Capital

Dependent Variable	Average Years of Schooling			Share Improved Drinking Water	Share Improved Sanitation Facilities	Share Children Stunted
	(1)	(2)	(3)	(4)	(5)	(6)
Within R ²	0.550	0.520	0.578	0.334	0.439	0.371
% Within R ² accounted for by urbanization	0.474	0.804	0.647	0.657	0.635	0.607
Agriculture related geography variables	yes	no	yes	yes	yes	yes
Trade related geographic variables	no	yes	yes	yes	yes	yes
Observations	552	552	552	544	535	473

Notes: * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. The table reports standardized beta coefficients. Standard errors clustered at the country level. All regressions control for country fixed effects and surface area of regions. Agriculture-related geography variables include precipitation, temperature, agricultural suitability, elevation, and absolute latitude. Trade-related geography variables include distance to coastline, the percentage of a region that lies within 25 kilometers of the coastline, the Strahler index and terrain ruggedness. % *Within R² accounted for by urbanization* represents the share of the total explanatory power of the geographic variables that is absorbed by the partial correlation with the urban population share (see the main text for more details).