GIS for Economists 3

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Overview The plan for today

Replication: Michalopoulos (2012)

- Introduction to the paper
- Empirical strategy and how it relates to GIS data
 - Cross-Country analysis
 - Cross-Virtual-Country analysis
 - Dyadic analysis
- Replication with geopandas

Research question

Michalopoulos, Stelios. (2012). "The Origins of Ethnolinguistic Diversity," American Economic Review, 102(4): 1508-1539

What determines ethnolinguistic diversity within and across countries?

Main idea: Diversity in land endowments across regions \Rightarrow formation and persistence of ethnic diversity.

- 1. Variation in regional land quality ⇒ region specific human capital
- 2. Differences in region specific human capital ⇒ barrier to population mixing
- 3. Limited population mixing between regions \Rightarrow emergence of differential ethnic traits

This was a significant contribution to a large literature that had focused on ethnic diversity as a RHS variable.

This section is based on presentation slides available on the author's website



Research design: Cross-country analysis

The empirical analysis establishes that geographic variability, captured by variation in regional land quality and elevation, is a fundamental determinant of contemporary linguistic diversity.

How to do this in practice?

Three main approaches:

- Country level analysis
- Virtual-countries
- Dyadic-analysis

Research design: Cross-country analysis

At the **country**-level, what is the effect of land quality variation on the number of languages?

 $log(Number of languages_i) = \beta_0 + \beta_1 Variation in Land Quality_i + \gamma X_i + \eta_i, (1)$

- \rightarrow where *i* indexes countries.
 - What is a concern?

Research design: Cross-country analysis

At the **country**-level, what is the effect of land quality variation on the number of languages?

 $log(Number of languages_i) = \beta_0 + \beta_1 Variation in Land Quality_i + \gamma X_i + \eta_i, (1)$

- \rightarrow where *i* indexes countries.
 - What is a concern?
 - Modern centralized states (which often formed along geographic boundaries) affected the distribution of languages (education, language policies, conquest, genocide).
 - Have to account for state-specific histories.

Research design: Cross-virtual-country analysis

Idea: Virtual countries

- Divide earth into cells of equal size ("virtual countries")
- Then run, as before (note X_i can include country fixed effects):

$$log(Number of languages_i) = \beta_0 + \beta_1 Variation in Land Quality_i + \gamma X_i + \eta_i, (2)$$

- \rightarrow where i now indexes virtual countries.
 - Could we still be concerned?

Idea: Virtual countries

- Divide earth into cells of equal size ("virtual countries")
- Then run, as before (note X_i can include country fixed effects):

 $log(Number of languages_i) = \beta_0 + \beta_1 Variation in Land Quality_i + \gamma X_i + \eta_i, (2)$

- \rightarrow where i now indexes virtual countries.
 - Could we still be concerned?
 - Standard concern: omitted variable bias, η_i could be correlated with *Variation in Land Quality*_i
 - Can we focus on similar regions that differ only in land quality?

Research design: Pairwise analysis of adjacent regions

Idea: Dyadic analysis of adjacent regions

- Divide earth into cells of equal size (1/25 the size of the previous virtual countries)
- Then run dyad pairs with cell fixed effects:

Percentage of common languages
$$_{ij} = \alpha_i + \alpha_j + \beta_1 Ab$$
solute difference in Land Quality $_{ij} + \gamma X_{ij} + \xi_{ij}$, (3)

- \rightarrow where now i and j index adjacent cells.
 - Advantage of dyadic structure
 - Minimize concerns that differences in unobservables drive differences in number of languages since focus on adjacent cells
 - See related modern methods paper by Druckenmiller and Hsiang (2019) on Spatial First-Differences.



Replication with geopandas

We will cover GIS methods to create data for:

- Countries
- Virtual-countries
- Dyads

For each of these analyses we need data on:

- Languages
 - Linguistic groups' homelands from WLMS, accurate between 1990 and 1995.
- Land quality
 - Agricultural suitability, elevation, climate, proximity to coast, etc.

Replication with geopandas: full list of inputs

- Languages: Michalopoulos uses WLMS
 http://www.worldgeodatasets.com/language/. We have an old version of this called langa.shp
- Agricultural suitability: https://nelson.wisc.edu/sage/data-and-models/atlas/data.php? incdataset=Suitability%20for%20Agriculture
- Population density for different years http://themasites.pbl.nl/ tridion/en/themasites/hyde/download/index-2.html
- Country boundaries http://www.naturalearthdata.com/downloads/ 10m-cultural-vectors/10m-admin-0-countries/
- Coastline http://www.naturalearthdata.com/downloads/ 10m-physical-vectors/10m-coastline/
- Lakes http://www.naturalearthdata.com/downloads/ 10m-physical-vectors/10m-lakes/
- Elevation https://cgiarcsi.community/data/ srtm-90m-digital-elevation-database-v4-1/
- Temperature and rainfall https://www.worldclim.com/current

Replication with geopandas: Settings

Each stage of the python script is listed on the following slides. Check the table of contents in the Jupyter notebook to find the corresponding code location.

- 1. Main settings
 - Import python packages
 - Set paths for file locations on your computer
 - Set file locations for inputs, temporary, and output files
 - Define other settings (e.g. coordinate systems)

Replication with geopandas: Cleaning

- 2.A. Clean language data
 - add unique ID, rename columns, and drop unnecessary fields
- 2.B. Clean agricultural suitability data
 - add coordinate system
- 2.C. Clean population data
 - convert from ASCII raster type to GeoTiff
- 2.D. Clean elevation data
 - convert from AIG raster type to GeoTiff
- 2.E. Clean climate data
 - take averages across months

Replication with geopandas: Cross-Country analysis

3.A. Aggregate data by country

- We have a bunch of raster data (agricultural suitability, elevation, temperature, rainfall, population density in different years)
- We want to compute zonal statistics (such as mean and standard deviation of agricultural suitability and elevation) in a country.
- Pre-assign all the variables
- Write a loop where each iteration computes Zonal Statistics of a different raster
- Output the results to .csv

Replication with geopandas: Cross-Country analysis

- 3.B. Count languages for countries
 - Intersect WLMS and countries
 - groupby to count number of languages that intersect each country
- 3.C. Calculate distances between countries and coast
 - Re-set the country coordinate system to an equal area projection
 - Find country centroids
 - Calculate distance to coast from each centroid.
- 3.D. Calculate country areas
 - Re-set the country coordinate system to an equal area projection
 - Calculate area in square kilometers
 - Output all the results to .csv



Replication with geopandas: Cross-Virtual-Country analysis

4.A. Build Grid for virtual countries

- Create a grid of 2.5×2.5 degree cells covering the world
- Add unique ID for each virtual country
- Intersect the cells with the actual countries to remove oceans
- Dissolve the intersections to get single units for each virtual country
- Clean up holes created by unaligned country borders
- Save grid as .shp

4.B. Count languages for virtual countries

- Intersect virtual countries with WLMS
- groupby to count number of languages that intersect each country
- Output the language counts to .csv
- Get virtual countries without languages with Difference
 - This creates a few 'broken' geometries that we need to fix.
- Save virtual countries with and without languages as .shp



Replication with geopandas: Cross-Virtual-Country analysis

- 4.C. Calculate land and water areas
 - Calculate virtual country area
 - Intersect virtual countries with lakes, calculate water area for each virtual country
- 4.D. Calculate distances between virtual countries and coast
 - Calculate coordinates of each virtual country centroid
 - Calculate distance between centroids and coast
 - Get coordinates of point on coast nearest to virtual country
- 4.E. Aggregate data by virtual countries
 - As for the countries, we loop over the different rasters and each iteration uses Zonal Statistics
 - Output all the results to .csv

Replication with geopandas: Dyadic analysis

- 5.A. Create dyad cells
 - As for virtual countries above, just change the resolution to 0.5×0.5 decimal degrees.
- 5.B. Languages spoken in dyad cells
 - Before we only cared about the number. Now we want the percentage common to the dyad. ⇒ need the actual languages. ⇒ Spatial Join the cells to WLMS
 - Output languages per virtual country to .csv
- 5.C. Calculate land and water area for dyad cells
 - Same as above for virtual countries
- 5.D. Calculate distances between dyad cells and coast
 - Same as above for virtual countries

Replication with geopandas: Cross-Country analysis

5.E. Aggregate data by dyad cells

- Similar to countries and virtual countries, but now use point query to extract values from points
- We do this partly because the suitability raster has resolution 0.5×0.5 degrees, and partly because we want to show a new tool, but we could also do this with Zonal Statistics.
- Output the results to .csv

5.F. Create polygon neighbours

- Loop through every dyad cell
- Identify each cell's neighbours (neighours are all cells that are NOT disjoint)
- Remove self as neighbour
- Output the neighbour pairs to .csv