

# A Stata 17 implementation of the local autonomy ratio: Calling Python

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# SUMMARY

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- 1 Introduction: The Local Autonomy Ratio
- 2 How Python Can Help
- 3 The Code: Python and Stata
- 4 Applications
- 5 Do-file execution

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# Introduction: The Local Autonomy Ratio

- Countries whose public sector is decentralized: +Efficiency in public services
- Level of decentralization approximated through the **Local Income Ratio**
  - **but** it has been proven that this covariate is **endogenous**
  - **and** due to the **unobservable heterogeneity**



# Introduction: The Local Autonomy Ratio

- In **Chile**, the participation of the **Municipal Compensation Fund (FCM)**, in the total revenues of the municipality, is the best **indicator of decentralization** (Letelier and Sáez-Lozano, 2020)
- In 2011, Martínez-Vazquez, Vulovic, and Liu paper:
  - an **instrument variable** ( $V_{st}$ ) shall be used to test the robustness of the empirical model
  - to control for **possible measurement errors** of the FCM
  - and detect **endogeneity biases**



# Introduction: The Local Autonomy Ratio

- Martínez-Vázquez et al. (2011) and Sanogo (2019) define the **instrumental variable** of municipality  $s$  in year  $t$ ,  $V_{st}$ :

- $$V_{st} = \frac{1}{\sum_{m=1}^M \frac{1}{d_m}} \sum_{m=1}^M \frac{1}{d_m} FCM_{mt}, s \neq m \mid d_m \leq \bar{d} \quad (1)$$

- where,  $d_m$  is distance between municipality  $s$  and  $m$ ;  $FCM_{mt}$  is FCM of the municipality  $m$  in year  $t$ ;  $M$  is the total number of municipalities;  $d_m$  is the distance between municipality  $m$  and the other municipalities in the country;  $\bar{d}$  is the threshold distance.



# Introduction: The Local Autonomy Ratio

- Letelier and Sáez (2020) **modified**  $V_{st}$  to apply it to Chile, including, also to the "distance" restriction, the population size limit:

$$V_{st} = \frac{1}{\sum_{m=1}^M \frac{1}{d_m}} \sum_{m=1}^M \frac{1}{d_m} FCM_{mt}, s \neq m \mid d_m \leq \bar{d} \text{ and } Pob_{st} \geq \overline{Pob} \quad (2)$$

- where,  $Pob_{st}$  is the population of municipality  $s$  in period  $t$ ; and  $\overline{Pob}$  is the threshold population.



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# Why python?

- To compute the instrumental variable “v” for the **Improved Local Autonomy**

**Ratio** we face:

1. Large dataset (**a matrix with distances between every single pair of municipalities in a given country**)
  - a) *e.g.: In Chile there are 343 municipalities, we have to deal with a 343x343 matrix*
2. Necessity to update the dataset (Possibility to use API's for data scrapping)
3. Apply different restrictions
  - a) minimum distance between towns
  - b) or a minimum population per town



# Why python?

4. Automatize its calculation to facilitate its application
  - a) Allow non-python users to take advantage of Stata's power
  - b) Attract python users to use the best Stata features vs python (Econometrics)
5. Explore Stata 17's newest python features (writing Stata code in Jupyter Lab)



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# The code: Calculating the Improved Local Autonomy Ratio (Vst)

- The structure:
  1. The data: Excel and structure of data
  2. Stata: Ask the user about the dataset to use and the restrictions to apply
  3. Python: read the data, apply restrictions, calculate V for each municipality.
  4. Stata: Use the SFI module to load data back into Stata's Data Editor
  5. Stata: apply econometrics models to reveal possible correlations.



# The code: Calculating the Improved Local Autonomy Ratio (Vij)

1. Stata: Prompt the user about the dataset to use (Excel file's path) and the restrictions to apply

\*Note that dataset has to include: 1)Distance matrix, 2)Decentralization Index (FCM), 3) Population for every municipality

\*\* Python's input("question?") function doesn't work in Stata

```
display "Introduce file's name (no commas needed) " _request(file_stata)
display "Introduce number of columns with no distances" _request(number_stata)
display "Introduce name of the column with the FCM to study" _request(distance_stata)
display "Introduce name of the column with the population to study" _request(distance_stata)
display "Introduce maximum distance between cities" _request(distance_stata)
display "Introduce maximum number of population per town: " _request(population_stata)
```



Company Overview



Industry Overview



The code



Transaction Opportunities



Team Overview

# The code: Calculating the Improved Local Autonomy Ratio (Vij)

## 2. A) Python: import modules, read data, set distance matrix and checks

*\*We invoke necessary modules and packages*

```
import pandas as pd
import numpy as np
import Dataframe as df
```

*\*We read the data and check its shape*

```
file_python= "$file_stata"
data = pd.read_csv(file_python)
print(data.head())
```

*\*We set the distance matrix \*We need to split the columns with values of variable FCM or Population from distance's value by writting down the number of columns that does not contain distances*

```
number_python= "$number_stata"
n = int(number_python)
distances = data.iloc[:len(data.columns)-1,1:len(data.columns)-n]
```

*\*We can check matrix's dimension with:*

```
print(distances.shape)
```



Company Overview



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The code



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# The code: Calculating the Improved Local Autonomy Ratio (Vij)

## 2. B) Python: Applying restrictions and calculating inverse

*\*We apply the chosen maximum population restriction*

```
population_1 = "$population_stata" #This is a variable already defined in Stata in the second line of code  
population = int(population_1)  
populationcolumn_python = "$populationcolumn_stata"  
distances.loc[data[populationcolumn_python]<=population]
```

*\*We apply the maximum distance restriction*

```
y_stata = "$distance_stata"  
y = int(y_stata)  
matrix_minor_y = (np.where(distances<=y, distances,0))  
print(matrix_minor_y)  
*Inverse  
inverse = 1/matrix_minor_y  
from numpy import inf  
inverse[inverse==inf]=0 #Correcting mistake for 1/0 (equal to "if error")  
print('INVERSE')  
print(inverse)
```

# The code: Calculating the Improved Local Autonomy Ratio (Vij)

## 2. C) Python: We compute the formula

\*We get COLUMN C =

```
FCM_python = "FCM_stata"
```

```
product_fcm = np.matmul(inverse,data[FCM_python])
```

```
print(pd.Series(product_fcm))
```

```
print(product_fcm.shape) #We use pd.Series so results are shown in a single column
```

\*We get column D =

```
sum_inverses = np.sum(inverse,0)
```

```
print('Sum inverses per municipality')
```

```
print(pd.Series(sum_inverses))
```

```
print(sum_inverses.shape)
```

```
inverse_sum_inverses = 1/sume_inverses
```

```
print(pd.Series(inverse_sum_inverses))
```

\*We finally get Vst

```
v = inverse_sum_inverses*product_fcm
```

```
v_df = pd.DataFrame(v)
```

```
v_df.columns = ["v value"]
```

```
print(v_df)
```



# The code: Calculating the Improved Local Autonomy Ratio (Vst)

\*We clean the results, removing municipalities that did not satisfy our restrictions

```
results = pd.Series(v, index = data['MUNICIPALITY']).dropna()
print(results)
results = pd.DataFrame(results)
results.columns = ["v value"]
print(results)
```

# The code: Calculating the Improved Local Autonomy Ratio (Vst)

4) Now we use SFI to load the data back in Stata and save results

```
from sfi import Data
Data.setObsTotal(len(results))
Data.addVarStr("MUNICIPALITY",len(data['MUNICIPALITY']))
Data.addVarDouble("v_value")

Data.store("MUNICIPALITY", None, data['MUNICIPALITY'],None)
Data.store("v_value",None, v ,None)
end
*Rename v_value for FCM_2011
rename v_value_ v_value_2011
*Generate variables year and id
gen year = 2011
gen id = _n
*Save v_value_2011
save /Users/juanmorales/Desktop/Stata/v_value_2011.dta , replace
```

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# APPLICATIONS

- The **instrumental variable  $V_{st}$**  is a **covariate of the happiness model** that Letelier and Saez-Lozano (2020) estimated for the case of Chile, in the years 2011 and 2013.
- Since happiness is a latent variable, the **level of satisfaction** with individual life, as a **proxy** variable for happiness. The authors specified a multilevel model, in which one of the covariates was  $V_{st}$ .
- They used the Newton-Raphson algorithm to maximize the likelihood function, which is done according to the adaptive quadrature procedure proposed by Rabe-Hesketh et. to the. (2005). The main conclusion that this research provides **evidence in support of the hypothesis that decentralization ( $V_{st}$ ) increases the level of happiness of the Chilean population.**



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# REFERENCES

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Martinez-Vasquez, J., Vulovic, V., & Liu, Y. (2011). *Direct versus Indirect Taxation: Trends, Theory, and Economic Significance*, Georgia State University. International Studies Program, 09–11: 37–92.

Rabe-Hesketh, S., Skrondal, A. and Pickles, A. (2005). *Maximum likelihood estimation of limited and discrete dependent variable models with nested random effects*. Journal of Econometrics, 128(2): 301-323.

Sanogo T (2019) *Does fiscal decentralization enhance citizens access to public services and reduce poverty? Evidence from Côte d'Ivoire municipalities in a conflict setting*. World Development 113: 204-221.



# Bonus: A hand for creating an ado?

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- Attempts to create an ado from this do file has been unsuccessful
- When using Stata's program function, inability to connect Stata's variable with Python's variable
  - E.g.: `file_python= "$file_stata"`
- How could we compile all of these code into a program?

# Thanks for your attention!

Special appreciations to:

Stata

Statalist community

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