#### **SEGREGATION OF SYRIAN REFUGEES IN TURKEY:**

#### EVIDENCE FROM MOBILE PHONE DATA

\*\* PRELIMINARY – DO NOT CITE, COPY WITHOUT PERMISSION \*\*

2nd LSE Workshop on Political Economy of Turkey

LSE, September 11, 2020

Simone Bertoli, Çağlar Özden & Michael Packard

\*The usual World Bank disclaimer applies. All comments, results are the opinions of the presenter, not of the Board, senior management or the member countries.



- How can we use the spatial and temporal distribution of Syrian Refugees in Turkey?
- Relevant for the analysis of:

- Labor markets
- Product markets
- Real estate markets
- Political markets



### CHALLENGE:

- Absence of high quality/frequency data on the distribution of refugees
- Existing data too aggregated both temporally and spatially



- As part of Data for Refugees (D4R) challenge managed by TUBITAK & Bosphorus University
- Based on Call Detail Records (CDR) of incoming and outgoing calls of Turk Telekom (25% market share)
- CDR: caller ID, receiver ID, time/date, tower location
- ~ 1 million customers, 185k tagged as refugees.
- Refugees receive lower rates and need to show ID cards
- Multiple data files. We use mainly datafile #1

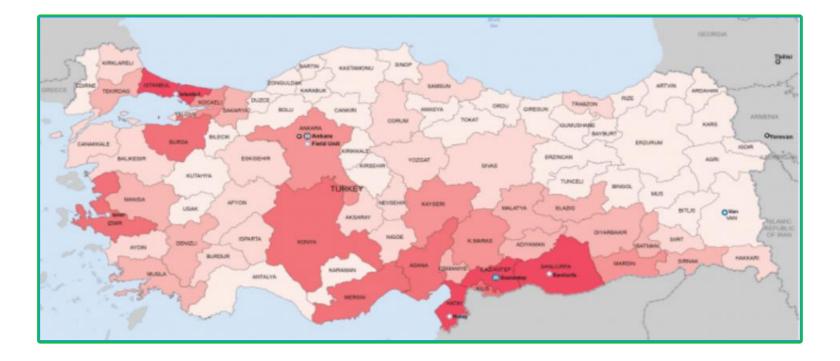


- Datafile #1 includes <u>all</u> refugees and <u>sample</u> of Turkish customers, chosen from the same spatial distribution at the province level .
- Number of native <u>and</u> refugee calls for <u>each</u> tower (30k+) for <u>each</u> hour (8760 hours) in 2017.
- We use the call volume as a <u>proxy</u> for refugee and native populations' <u>temporal and spatial</u> distributions

### **DEFINITION OF IDENTITY**



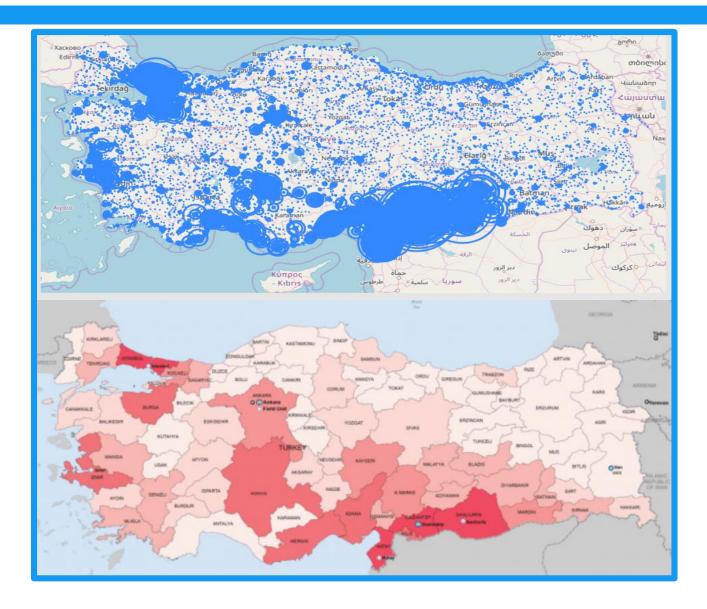
### **Syrian Refugees Distribution Across Provinces - UNHCR**



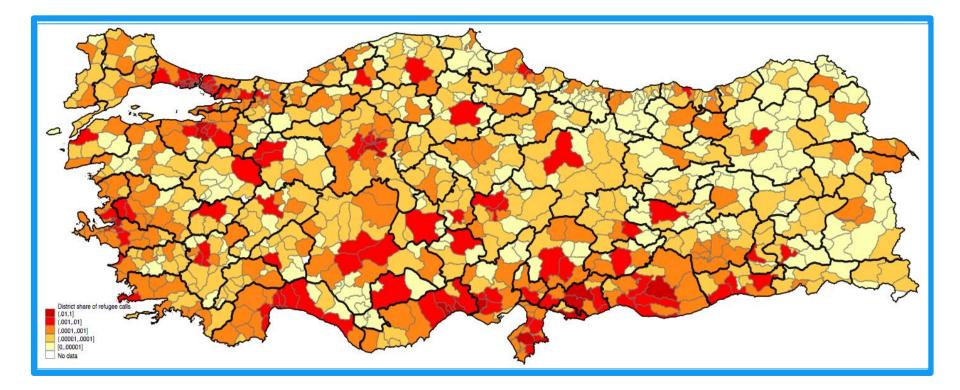
### Syrian Refugees Cellular Phone Call Volumes



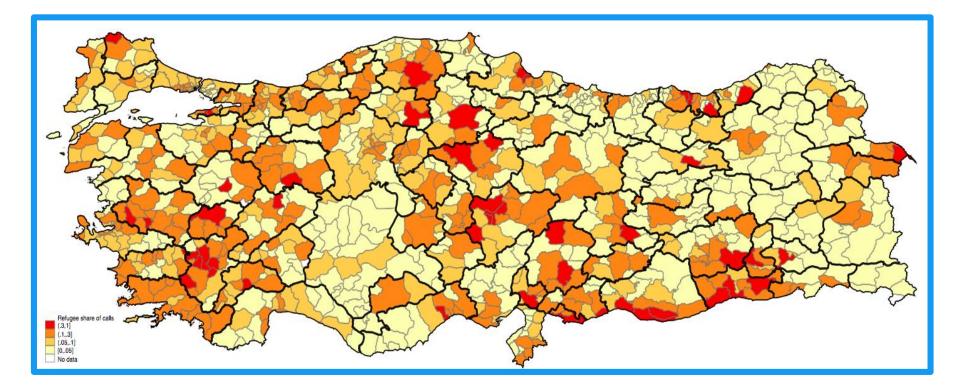
### Phone Data versus Administrative Data



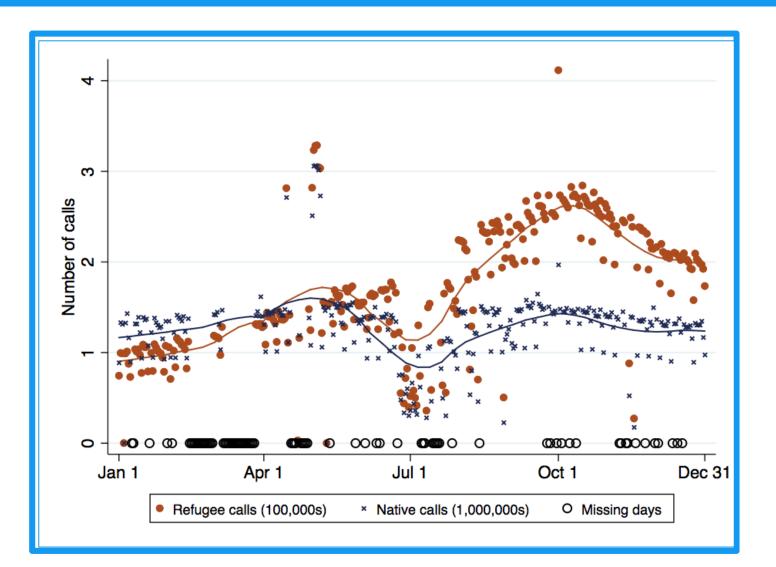
### DISTRIBUTION OF REFUGEE CALLS ACROSS THE DISTRICTS



### **REFUGEE SHARE OF CALLS IN EACH DISTRICT**



### **CALL VOLUME - OVER TIME**





- Identify the patterns of distribution so can use it for impact analysis
- Main Tools from the Vast Sociology Literature
- Several Segregation indices:
  - Dissimilarity Index
  - (adjusted) Assimilation Index

### **SEGREGATION INDICES**



#### Douglas S. Massey

Professor of Sociology and Public Affairs, <u>Princeton University</u> Verified email at princeton.edu - <u>Homepage</u>

Sociology Demography

TITLE	CITED BY	YEAR
American apartheid: Segregation and the making of the underclass DS Massey, NA Denton Harvard University Press	12202	1993
Theories of international migration: A review and appraisal DS Massey, J Arango, G Hugo, A Kouaouci, A Pellegrino, JE Taylor Population and development review 19 (3), 431-466	6812	1993
The dimensions of residential segregation DS Massey, NA Denton Social forces 67 (2), 281-315	3412	1988

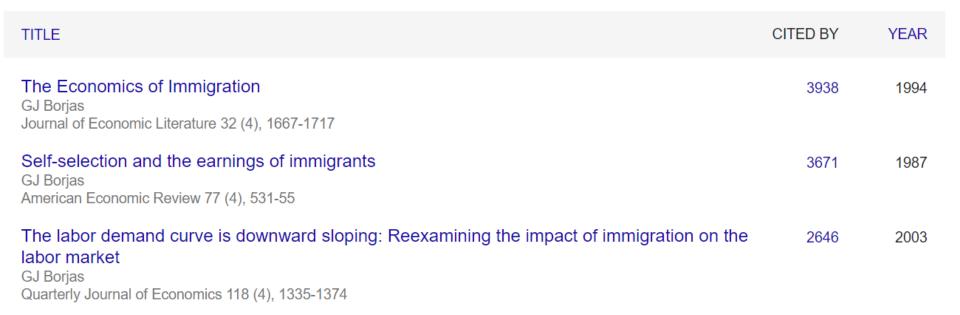
M FOLLOW

### **SEGREGATION INDICES**



#### George Borjas

Harvard Kennedy School Verified email at harvard.edu - <u>Homepage</u> Labor Economics Economics



**FOLLOW** 

## **DISSIMILARITY INDEX**

$$D = \frac{1}{2(1-P)} \sum_{i=1}^{K} \left| \frac{m_i}{M} - \frac{t_i}{T} \right|$$
$$= \frac{1}{2} \sum_{i=1}^{K} \left| \frac{m_i}{M} - \frac{n_i}{N} \right|$$

- K number of areas
- m<sub>i</sub> minority population in areas i
- n<sub>i</sub> non-minority population in areas i
- t<sub>i</sub> total population in areas i
- M total minority population
- N total non-minority (native) population
- T total population

### **DISSIMILARITY INDEX**

- D ∈ [0,1],
  - D=0 (1) is complete integration (segregation)
- D : share of the minority (refugee) population that needs to be relocated from high to low concentration areas to match their average distribution across the country
- D : <u>insensitive</u> to an identical proportional increase in the size of the minority group across all areas – since "m<sub>i</sub>/M" stays constant.

### **ADJUSTED ISOLATION INDEX**

$$I \equiv \sum_{i=1}^{K} \left( \frac{m_i}{t_i} \frac{m_i}{M} \right) = \sum_{i=1}^{K} \left( p_i \frac{m_i}{M} \right)$$
$$I^{\text{adj}} \equiv \frac{I - P}{\min\left\{ 1, M/(\min_i t_i) \right\} - P}$$

- m<sub>i</sub> minority population in areas i
- t<sub>i</sub> total population in areas i
- p<sub>i</sub> share of minority population in areas i
- M total minority population
- T total population
- P share of minority population in the total

# **ADJUSTED ISOLATION INDEX**

- I ∈ [0,1]
- Regular isolation index: <u>sensitive</u> to proportional increase in the size of the minority group across the regions.
- So construct the <u>adjusted</u> isolation index to <u>reduce</u>, not <u>eliminate</u>, its dependency on P.
- A uniform increase in the share of the refugees across all districts will increase the adjusted isolation index, I.

### **EMPIRICAL CHALLENGES**

- Objective: using the phone call distribution as a proxy for population distribution
- Problem #1: Different call propensities for Turkish and refugee customers
- Problem #2: Differences in market shares of Turkish Telekom within refugee population across provinces.
- Problem #3: Different degrees of geographic partitioning creates difference in indices

# PROBLEM #1 : DIFFERENT CALL PROPENSITIES BETWEEN GROUPS

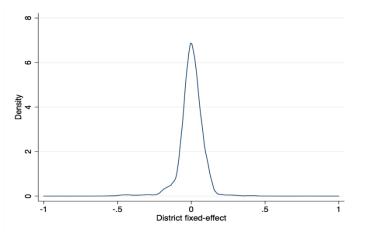
$$D = \frac{1}{2(1-P)} \sum_{i=1}^{K} \left| \frac{m_i}{M} - \frac{t_i}{T} \right| \\ = \frac{1}{2} \sum_{i=1}^{K} \left| \frac{m_i}{M} - \frac{n_i}{N} \right|$$

- No impact on Dissimilarity Index D if identical across geographic areas
- ... but, isolation index, I, becomes bigger if the refugees have a higher propensity (since p<sub>i</sub> will be bigger)

$$I \equiv \sum_{i=1}^{K} \left( \frac{m_i}{t_i} \frac{m_i}{M} \right) = \sum_{i=1}^{K} \left( p_i \frac{m_i}{M} \right) \qquad I^{\text{adj}} \equiv \frac{I - P}{\min\left\{1, M/(\min_i t_i)\right\} - P}$$

# PROBLEM #1 : DIFFERENT CALL PROPENSITIES BETWEEN GROUPS

- Use D4R #2 to see if propensity is different across provinces
- Smaller sample of individuals, identify time and location of ALL of their call throughout the year
- Calculate daily call volume of each individual, assign to a district for each day, regress the call volume on date and district fixed effects.



» These have mean zero with small standard deviation. 90% of observation are within 10 % of the average district. The rest are small sample districts.

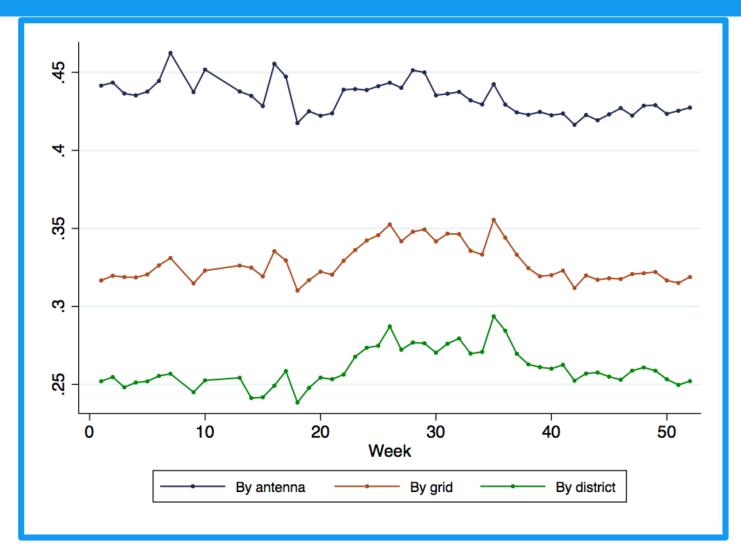
# **PROBLEM #2 : DIFFERENT MARKET SHARES**

- Different market shares across provinces!
- Take the official statistics from Refugee Administration at the province level for January 1 and July 1, 2017.
- Use these numbers as the weights to calculate the national indices, D and I.

## **PROBLEM #3: GEOGRAPHIC PARTITIONING**

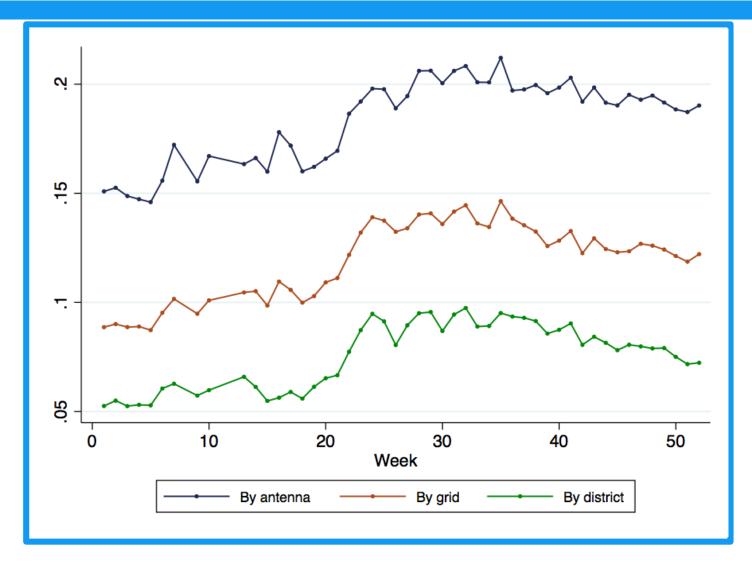
- 82 Provinces
- 957 districts
- Urban areas at least 15 cells within 5 km radius
- Catchment areas of the 30,000+ towers

# DISSIMILARITY INDEX EVOLUTION OVER TIME FOR THE WHOLE COUNTRY

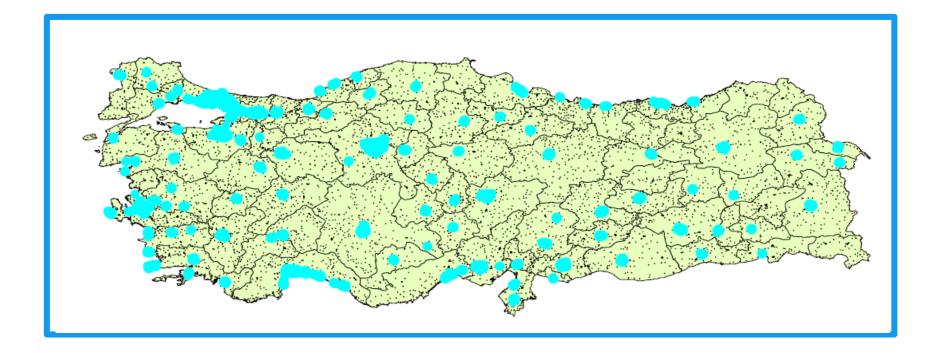


#### Note Weeks #26 and 35!

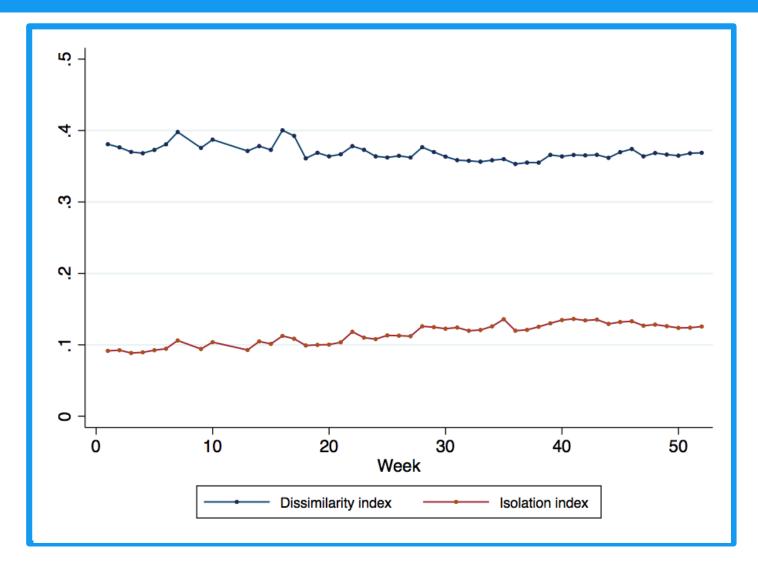
# ISOLATION INDEX EVOLUTION OVER TIME FOR THE WHOLE COUNTRY



### **URBAN AREAS**



# DISSIMILARITY AND ISOLATION INDICES URBAN AREAS



### **DETERMINANTS OF MOBILITY - NATIVES**

Table 1: Gravity Estimates for Natives (Dataset 3)							
	(1)	(2)	(3)	(4)	(5)	(6)	
$Contiguity_{jk}$	0.672	0.635	0.623	0.624	0.615	0.611	
	(0.144)***	(0.138)***	(0.138)***	(0.139)***	(0.141)***	(0.141)***	
$Ln(Distance_{jk})$	-0.368	-0.307	-0.292	-0.289	-0.290	-0.294	
	(0.073)***	(0.077)***	(0.076)***	(0.076)***	(0.078)***	(0.078)***	
$Ln(Population_k)$		0.956	0.920	0.921	0.916	0.914	
		(0.040)***	(0.040)***	(0.040)***	(0.041)***	(0.041)***	
$Ln(GDP/capita_k)$		0.418	0.544	0.527	0.513	0.533	
		(0.095)***	(0.099)***	(0.106)***	(0.104)***	(0.107)***	
$Refugeeshare_k$			2.083	1.963	2.032	2.220	
			(0.731)***	(0.828)**	(0.787)***	(0.855)**	
$Dissimilarity index_k$				-0.238		0.369	
				(0.647)		(0.824)	
$Isolation index_k$					-0.839	-1.033	
					(0.626)	(0.800)	
Origin FE	Yes	Yes	Yes	Yes	Yes	Yes	
Dest FE	Yes	No	No	No	No	No	
Pseudo- $R^2$	0.59	0.58	0.58	0.58	0.58	0.58	
Origin-destination pairs	2,400	2,400	2,400	2,400	2,400	2,400	
Aggregate sample	13,142	13,142	13,142	13,142	13,142	13,142	

Table shows results from Poisson pseudo-maximum likelihood (PPML) regressions of cross-province

Non-refugee migration flows calculated from dataset 3. Population and GDP variables are calculated for

2013. Province level refugee shares are taken from Turkish administrative data and correspond to stocks as

### **DETERMINANTS OF MOBILITY - REFUGEES**

	(1)	(2)	(3)	(4)	(5)	(6)
$Contiguity_{jk}$	0.428	0.447	0.422	0.381	0.417	0.408
	(0.154)***	(0.157)***	(0.152)***	(0.151)**	(0.156)***	(0.160)**
$Ln(Distance_{jk})$	-0.295	-0.401	-0.275	-0.266	-0.275	-0.260
	(0.064)***	(0.092)***	(0.069)***	(0.067)***	(0.069)***	(0.069)***
$Ln(Population_k)$		1.164	1.065	1.042	1.062	1.061
		$(0.054)^{***}$	$(0.044)^{***}$	$(0.045)^{***}$	(0.043)***	(0.040)***
$Ln(GDP/capita_k)$		0.066	0.644	0.582	0.639	0.591
		(0.115)	(0.112)***	(0.110)***	(0.115)***	(0.114)**
$Refugeeshare_k$			7.905	7.719	7.949	7.033
			(0.365)***	(0.373)***	(0.409)***	(0.499)**
$Dissimilarity index_k$				-2.187		-4.650
				(0.536)***		(1.380)**
$Isolation index_k$					-0.214	2.966
					(0.701)	(1.440)**
Origin FE	Yes	Yes	Yes	Yes	Yes	Yes
Dest FE	Yes	No	No	No	No	No
Pseudo- $R^2$	0.66	0.58	0.63	0.64	0.63	0.64
Origin-destination pairs	2,400	2,400	2,400	2,400	2,400	2,400
Aggregate sample	15,497	15,497	15,497	15,497	15,497	15,497

Table shows results from Poisson pseudo-maximum likelihood (PPML) regressions of cross-province Refugee migration flows calculated from dataset 3. Population and GDP variables are calculated for 2013. Province level refugee shares are taken from Turkish administrative data and correspond to stocks as of

### **DETERMINANTS OF MOBILITY - NATIVES**

Table 1: Gravity Estimates for Natives (Dataset 3)							
	(1)	(2)	(3)	(4)	(5)	(6)	
$Contiguity_{jk}$	0.672	0.635	0.623	0.624	0.615	0.611	
	(0.144)***	(0.138)***	(0.138)***	(0.139)***	(0.141)***	(0.141)***	
$Ln(Distance_{jk})$	-0.368	-0.307	-0.292	-0.289	-0.290	-0.294	
	(0.073)***	(0.077)***	(0.076)***	(0.076)***	(0.078)***	(0.078)***	
$Ln(Population_k)$		0.956	0.920	0.921	0.916	0.914	
		$(0.040)^{***}$	$(0.040)^{***}$	$(0.040)^{***}$	(0.041)***	(0.041)***	
$Ln(GDP/capita_k)$		0.418	0.544	0.527	0.513	0.533	
		$(0.095)^{***}$	$(0.099)^{***}$	$(0.106)^{***}$	$(0.104)^{***}$	(0.107)***	
$Refugeeshare_k$			2.083	1.963	2.032	2.220	
			(0.731)***	(0.828)**	(0.787)***	(0.855)***	
$Dissimilarity index_k$				-0.238		0.369	
				(0.647)		(0.824)	
$Isolation index_k$					-0.839	-1.033	
					(0.626)	(0.800)	
Origin FE	Yes	Yes	Yes	Yes	Yes	Yes	
Dest FE	Yes	No	No	No	No	No	
Pseudo- $R^2$	0.59	0.58	0.58	0.58	0.58	0.58	
Origin-destination pairs	2,400	2,400	2,400	2,400	2,400	2,400	
Aggregate sample	13,142	13,142	13,142	13,142	13,142	13,142	

Table shows results from Poisson pseudo-maximum likelihood (PPML) regressions of cross-province

Non-refugee migration flows calculated from dataset 3. Population and GDP variables are calculated for

2013. Province level refugee shares are taken from Turkish administrative data and correspond to stocks as

### **DETERMINANTS OF MOBILITY - REFUGEES**

	(1)	(2)	(3)	(4)	(5)	(6)
Continuitor	0.428	0.447	0.422	0.381	0.417	0.408
$Contiguity_{jk}$	(0.420 (0.154)***	(0.157)***	(0.152)***	(0.151)**	(0.156)***	(0.160)**
L (D:						
$Ln(Distance_{jk})$	-0.295	-0.401	-0.275	-0.266	-0.275	-0.260
	$(0.064)^{***}$	(0.092)***	$(0.069)^{***}$	(0.067)***	$(0.069)^{***}$	(0.069)**
$Ln(Population_k)$		1.164	1.065	1.042	1.062	1.061
		(0.054)***	(0.044)***	(0.045)***	(0.043)***	(0.040)**
$Ln(GDP/capita_k)$		0.066	0.644	0.582	0.639	0.591
		(0.115)	(0.112)***	(0.110)***	(0.115)***	(0.114)**
$Refugeeshare_k$			7.905	7.719	7.949	7.033
			(0.365)***	(0.373)***	(0.409)***	(0.499)**
$Dissimilarity index_k$				-2.187		-4.650
				(0.536)***		(1.380)**
$Isolation index_k$					-0.214	2.966
					(0.701)	(1.440)**
Origin FE	Yes	Yes	Yes	Yes	Yes	Yes
Dest FE	Yes	No	No	No	No	No
Pseudo- $R^2$	0.66	0.58	0.63	0.64	0.63	0.64
Origin-destination pairs	2,400	2,400	2,400	2,400	2,400	2,400

Table shows results from Poisson pseudo-maximum likelihood (PPML) regressions of cross-province Refugee migration flows calculated from dataset 3. Population and GDP variables are calculated for 2013. Province level refugee shares are taken from Turkish administrative data and correspond to stocks as of

### **DETERMINANTS OF MOBILITY - NATIVES**

	(1)	(2)	(3)	(4)	(5)	(6)
$Contiguity_{jk}$	0.672	0.635	0.623	0.624	0.615	0.611
	(0.144)***	(0.138)***	(0.138)***	(0.139)***	(0.141)***	(0.141)***
$Ln(Distance_{jk})$	-0.368	-0.307	-0.292	-0.289	-0.290	-0.294
	(0.073)***	(0.077)***	(0.076)***	(0.076)***	(0.078)***	(0.078)***
$Ln(Population_k)$		0.956	0.920	0.921	0.916	0.914
		(0.040)***	(0.040)***	$(0.040)^{***}$	(0.041)***	(0.041)***
$Ln(GDP/capita_k)$		0.418	0.544	0.527	0.513	0.533
		$(0.095)^{***}$	$(0.099)^{***}$	$(0.106)^{***}$	(0.104)***	(0.107)***
$Refugeeshare_k$			2.083	1.963	2.032	2.220
			(0.731)***	(0.828)**	(0.787)***	(0.855)***
$Dissimilarity index_k$				-0.238		0.369
				(0.647)		(0.824)
$Isolation index_k$					-0.839	-1.033
					(0.626)	(0.800)
Origin FE	Yes	Yes	Yes	Yes	Yes	Yes
Dest FE	Yes	No	No	No	No	No
Pseudo- $R^2$	0.59	0.58	0.58	0.58	0.58	0.58
Origin-destination pairs	2,400	2,400	2,400	2,400	2,400	2,400
Aggregate sample	13,142	13,142	13,142	13,142	13,142	13,142

Table shows results from Poisson pseudo-maximum likelihood (PPML) regressions of cross-province

Non-refugee migration flows calculated from dataset 3. Population and GDP variables are calculated for

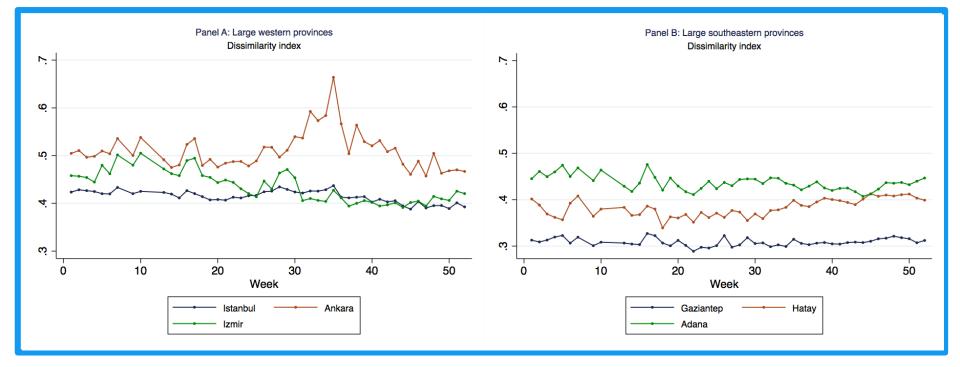
2013. Province level refugee shares are taken from Turkish administrative data and correspond to stocks as

### **DETERMINANTS OF MOBILITY - REFUGEES**

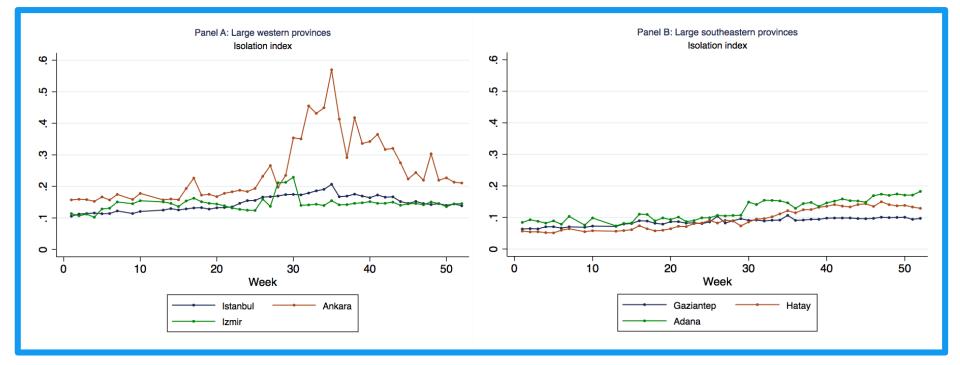
Table 2: Gravity Estimates for Refugees (Dataset 3)						
	(1)	(2)	(3)	(4)	(5)	(6)
$Contiguity_{jk}$	0.428	0.447	0.422	0.381	0.417	0.408
	(0.154)***	(0.157)***	(0.152)***	(0.151)**	(0.156)***	(0.160)**
$Ln(Distance_{jk})$	-0.295	-0.401	-0.275	-0.266	-0.275	-0.260
	(0.064)***	(0.092)***	(0.069)***	(0.067)***	(0.069)***	(0.069)***
$Ln(Population_k)$		1.164	1.065	1.042	1.062	1.061
		(0.054)***	(0.044)***	$(0.045)^{***}$	(0.043)***	(0.040)***
$Ln(GDP/capita_k)$		0.066	0.644	0.582	0.639	0.591
		(0.115)	$(0.112)^{***}$	$(0.110)^{***}$	(0.115)***	(0.114)***
$Refugeeshare_k$			7.905	7.719	7.949	7.033
			$(0.365)^{***}$	$(0.373)^{***}$	(0.409)***	$(0.499)^{***}$
$Dissimilarity index_k$				-2.187	Г	-4.650
				$(0.536)^{***}$		(1.380)***
$Isolation index_k$					-0.214	2.966
					(0.701)	(1.440)**
Origin FE	Yes	Yes	Yes	Yes	Yes	Yes
Dest FE	Yes	No	No	No	No	No
Pseudo- $R^2$	0.66	0.58	0.63	0.64	0.63	0.64
Origin-destination pairs	2,400	2,400	2,400	2,400	2,400	2,400
Aggregate sample	$15,\!497$	$15,\!497$	15,497	15,497	$15,\!497$	15,497

Table shows results from Poisson pseudo-maximum likelihood (PPML) regressions of cross-province Refugee migration flows calculated from dataset 3. Population and GDP variables are calculated for 2013. Province level refugee shares are taken from Turkish administrative data and correspond to stocks as of

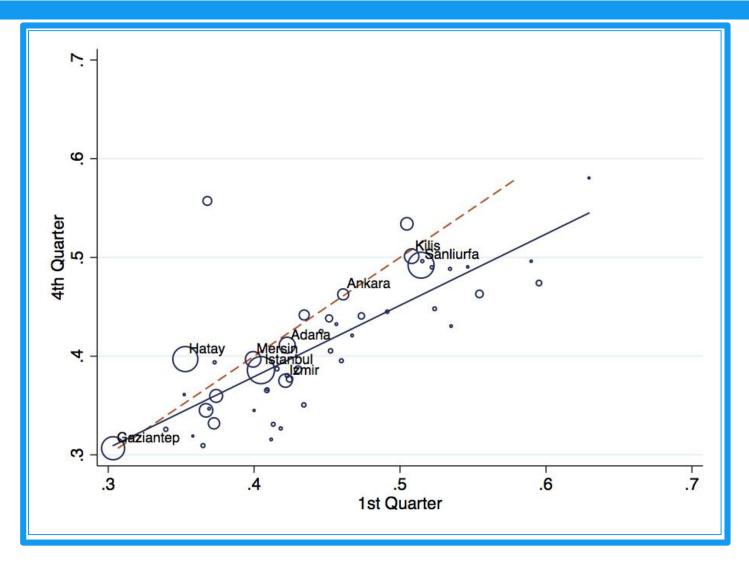
## DISSIMILARITY INDEX EVOLUTION OVER TIME FOR MAJOR PROVINCES



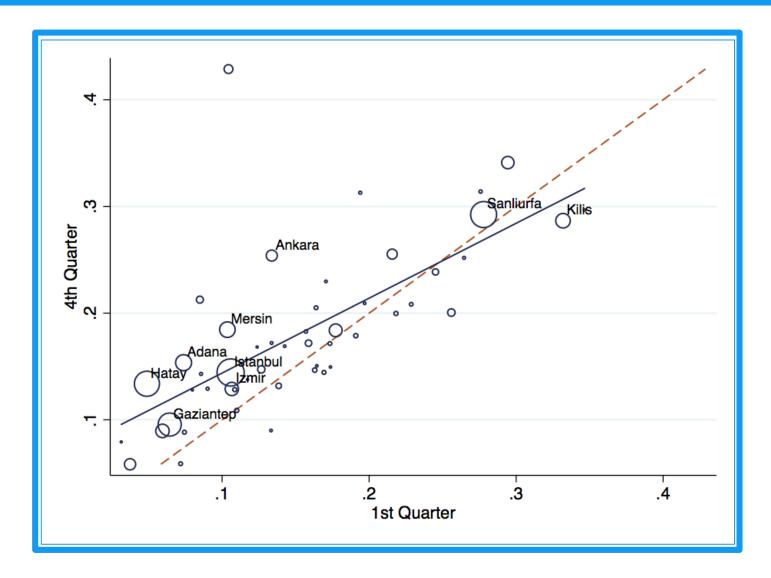
# ISOLATION INDEX EVOLUTION OVER TIME FOR MAJOR PROVINCES



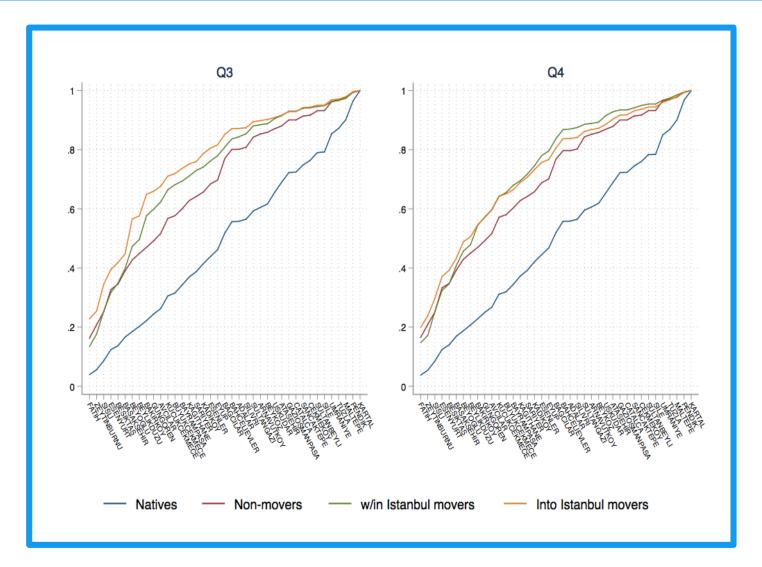
# DISSIMILARITY INDEX FIRST QUARTER VS. FOURTH QUARTER



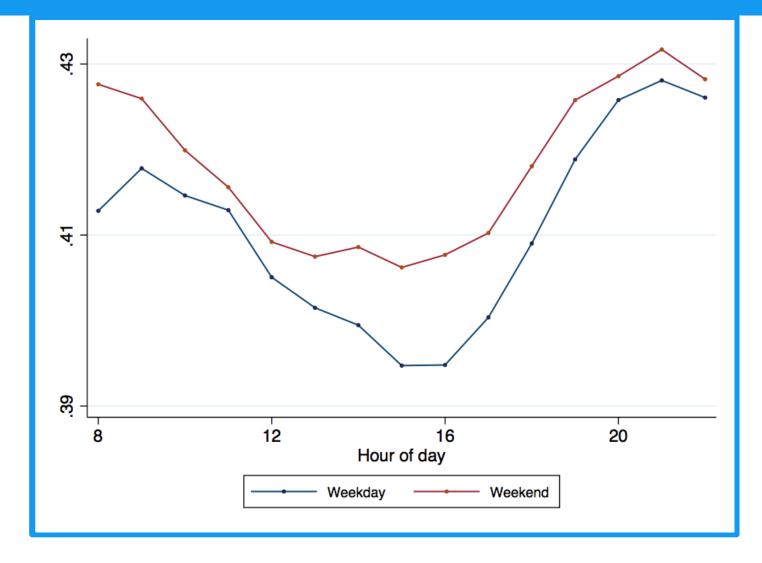
# ISOLATION INDEX FIRST QUARTER VS. FOURTH QUARTER



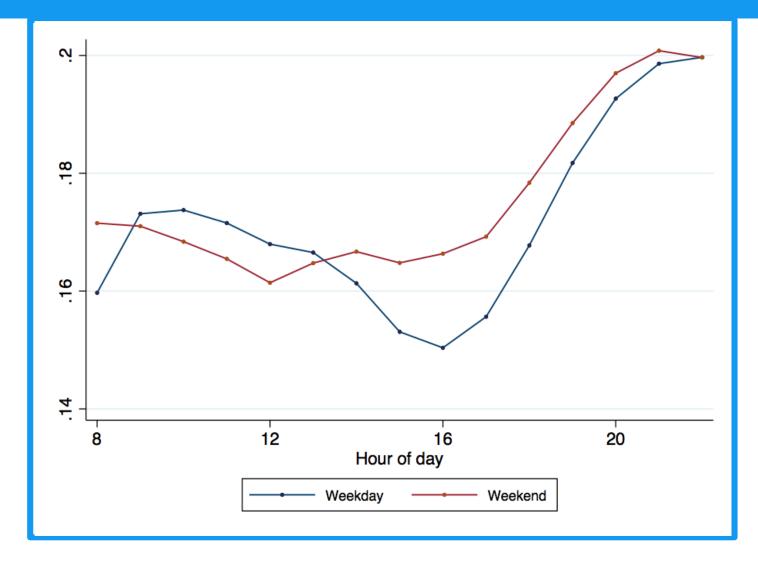
# DISTRIBUTION OF REFUGEES AND NATIVES ISTANBUL



# DISSIMILARITY INDEX HOUR OF THE DAY



# **ISOLATION INDEX HOUR OF THE DAY**



# CONCLUSION

- Big data can answer many questions
- Careful in how we use it
  - high frequency does not mean there are no measurement or sampling problems.
- Segregation measures depend on the partitioning
- Different segregation indices measure different things
- Time dimension is critical

# **CONCLUSION**

• There are large differences in provinces

• Segregation is actually lower in larger cities in the west as opposed to border cities

• Segregation is declining over time

• There are large differences between residential and labor market segregation

# THANK YOU!

cozden@worldbank.org