# Computing Functional Urban Areas Using a Hierarchical Travel Time Approach: An Applied Case in Ecuador Obaco M.\*, Royuela V.\*, Vítores X.

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- Introduction
- The proposal
- Sensitivity test
- Robustness checks
- Conclusions

## Administrative boundaries ≠ Economic boundaries

## SMA, LLMA (TTWA), FUA, FUR, Regionalization.....

#### -To collect information

- -To develop public policies
- -Normative use



Modifiable area unit problem (MAUP) is an inseparable part of almost any spatial analysis... (Klapka et al., 2014)

The correct identification of MAUP should reduce problems associated to mismeasurement of the size of the local economy (Briant et al. 2010)

#### Administrative boundaries ≠ Economic boundaries

- -There is not a consensus of the best approach (Halás et al., 2015)
- -Different approaches give different results and the same approach can give sharply different at different thresholds (Klapka et al., 2014)



#### Introduction



## **OECD Methodology: 3 steps**

## 1. Identifying urban cores:

✤Grid cells of high population density (1,000 – 1,500 inhab./km<sup>2</sup>).

✤Clusters of contiguous high population density (50,000 – 100,000 inhab. to be an urban core)

Municipality of reference (at least 50% of population)



## **OECD Methodology: 3 steps**

2. Connecting non-contiguous urban cores that belong to the same FUA: among areas of reference of urban cores (at least 15-50% of commuting flow)



## Introduction

**3. The hinterland**: surrounding areas that are mot urban cores to each urban core (at least 15-50% of commuting flow)



## The problem....?



# We are able to identifying FUAs in a suitable way using GIS data and travel time



## Following OECD methodology

- 1. Identifying Urban Cores:
  - ✓ High density grid cells in 1 km<sup>2</sup>
  - ✓ Cluster of high population density (Extra rules)
  - ✓ Minimum size of self-containment
- 2. Connecting non-contiguous Urban Cores:
  - An algorithm that uses travel time applied in a hierarchical procedure on the road network.
- **3. Defining hinterland:**

Radius of influence from the center of each urban core: $t_i = \frac{2}{3} * \sqrt{A_i \pi} / v$ ; (Ahlfeldt & Wendland, 2015)

## Application

#### ECUADOR:

- Small Open Developing Economy
- Not other important transportation system
- Average of population size and geographical characteristics





## DATA:

- 1<sup>st</sup> step) LandScan 2013 database –used OECD (Raster data of 1 km<sup>2</sup> in SHP) → QGIS
- 2<sup>nd</sup> step) Google map service (Stratification Algorithm, road information) → STATA\*
- 3<sup>rd</sup> step) Open Street Map (Isochrones-road information)
  → QGIS
- Administrative level: INEC (Parishes-level3)

## Application

## Setting the minimum thresholds

#### **MINIMUM THRESHOLD FOR URBAN CORES:**

- Half values applied in developed economies as starting point
- 500 inhb/km<sup>2</sup> and 25,000 inhb. urban core  $\rightarrow$  3% of total grids cells

#### TRAVEL TIME THRESHOLD:

- SHLC 2014; 1 hour by public transport (60%)
- 30 minutes by private car
- Fixed velocity of 45km/h





## 1<sup>st</sup> step)





## 1<sup>st</sup> step)



	Ν	Cells	Рор	Mean	Median	Max	Min	S.D.
	1	310	2553993	8238.69	5008.5	39800	0	9150.31
	2	523	2166700	4142.83	1753	41536	3	4950.62
lotal	3	97	347371	3581.14	1770	39473	92	4809.74
34	4	80	294618	3682.73	1910.5	21696	11	4337.59
urhan	5	32	286186	8943.31	5531	31110	58	9217.87
uivan	<b>6</b> 123 27	276507	2248.02	729	19390	7	3589.86	
cores	7	41	250088	6099.71	4272	43145	91	8935.1
	8	49	212192	4330.45	1891	35823	112	7233.95
	9	42	180342	4293.86	1318	36652	392	7853.18
	10	37	174433	4714.41	1849	19467	28	5388







26/10/2016











## Application

3<sup>rd</sup> step)



#### Sensitivity test of urban cores based on travel time

			Initial	Results (trave		s / FUAs el time)		
Threshold	Grid cells	Threshold	Cores	1/2 h	1h	1h30	2h	
500 inhab./km²		25,000	34	30	23	16	13	
	3,699 (3%)	50,000	21	20	16	14	12	
		100,000	16	15	13	12	11	
1,000 inhab./km²		25,000	29	28	22	15	13	
	2,114 (1.75%)	50,000	20	20	16	14	12	
	(111070)	100,000	16	15	13	12	2h 13 12 11 13 12 11 12 11 14 12 12 11	
1,500 inhab./km²		25,000	33	31	22	15	14	
	1,532	50,000	21	20	16	14	12	
	(1.2070)	100,000	16	15	13	12	11	

# **Robustness checks**

- -Commuting patterns: Survey HLC 2014
- 50,000 workers; 6,800 commuters; 2,800 pairs of parishes
- -Gravity equation: Rescale SHLC & National Census of

Population 2010.

$$Flow_{o,d} = \beta_1 M_o \ \beta_2 M_d e^{-\beta_3 D}$$

-Radiation model\*: National Census of population 2010.

$$F_{ij} *= F_i * \frac{Pop_i * Pop_j}{(Pop_i + w_{i,j}) (Pop_i + Pop_j + w_{i,j})}$$

-Internal migration: National Census of population 2010.

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# **Robustness checks**

#### Programmed in Stata;



Works with: 
$$Share\_commuters = \frac{F_{ij}}{Pop_i}$$



#### Descriptive statistics of commuters

## Estimated commuters (Full matrix size)

	OBS.	MINIMUM	MAXIMUM	MEAN	MEDIAN	ST.DEV.
SHLC	558,902	0	277	0.04	0	1.51
SHLC (RESCALED)	558,902	0	91,403	2.99	0	161.88
GRAVITY EQUATION	1.024.140	0	4.537	1.54	0	28.71
RADIATION MODEL	1,024,140	1.09E-12	7,563	0.94	5.49E-08	29.91
INTERNAL MIGRATION	1,024,140	1	13,453	12.03	2	98.55

#### **Comparison table**

	FUAs (1)	Min (2)	Max (3)	Mean (4)	Median (5)	St. Dev. (6)	TOTAL (7)	CV (8)
Travel time (30 minutes)	30	25,603	2,809,089	339,962	144,927	641,762	10,166,220 (64.5%)	53%
Commuting SHLC (10 %)	31	53,237	2,930,848	340,763	150,258	658,285	10,222,899 (65.15%)	52%
Commuting Gravitational (10 % )	33	37,663	2,769,539	295,143	107,129	618,271	9,739,748 (62.07%)	48%
Commuting Radiation (10%)	32	33,186	2,492,869	296,305	161,022	572,811	9,481,786 (60.05%)	52%
Migration (15 %)	29	59,312	2,558,798	417,070	280,325	634,405	11,260,940 (71.77%)	66%
26/10/2016 Based on nemulation 2013							29	

## **Conclusions**:

- Using GIS, we have enough available information to approximate integrated cities
- Travel time seems a good proxy to commuting patterns
- There is not a consensus among the best minimum threshold to work in developing countries. Although, low thresholds fit better in developing countries.
- Results become stables at very high thresholds. However, it might make invisible urban cores that can be important (e.g. Amazon region).
- The hinterland seems to be the most sensible and difficult to define.

# THANKS...