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# The valorisation of urban residential space. Questions and approaches for the quantitative geographer.

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## The housing market on the French Riviera

- Essential to understand the capacity of our cities to allow upward social mobility (access to home ownership), social mix and generational mix.
- A particularly tense housing market in the last decades (second only to the one in Paris)
- Not a market, but several markets, according to housing typology (studios, 2-room apts, 3-room apts, houses, etc.)
  - 114 K transactions on the French Riviera from 2010 to 2017 (out of 150 K, French notaries BD)



## From the value of dwellings to the value of spaces

Our research topic: Micro-geography of residential valorization Carpenter & Peponis (2010): the interest of fine-grained local approaches.

Census areas: too large, do not correspond to morphological entities







#### • The street segment:

- essential morphological component of the physical city (Conzen 1960, Caniggia & Maffei 2001)
- the basis of the functioning of the city as a social system (Jacobs 1960, Hillier 1996, Porta et al. 2006).
- A local ecosystem of the residential function

# Local statistical description of residential valorization

Geolocation of transactions on the urban street network.

Normalization of transactions values by housing type and year to understand « **value of the street** »: market statistics on each street segment.



ventile:	1	2	3	4	5	6	7	8	9	 
T1 2013			*			**				
T1 2016				*						
T2 2012				*				*		
T2 2016			*							
T3 2015		*								
T4 2012			*		*					
pm 2014					*					

D5 (median) = 4 D9 = 6 D1 = 3 D9-D1 (dispersion) = 3

Only 14,300 units with at least 10 transactions

## Median residential valorization on the French Riviera



## Median residential valorization in Nice



## **Dispersion of valorization**, a potential for social mix



A. Venerandi, G. Fusco, UMR ESPACE 2019

## **Social Mix potential in Nice**



## **Bayesian Clustering**

**Score optimization** (log-likelihood + penalization for number of clusters... in order to avoid overfitting).

Random walk in solution space under constraints :

- Min % of street segments in each cluster (1%)
- Max number of clusters (20)
- Minimum average cluster purity (90%)
- Pseudo-random initialization with given seed

Different search parametrizations → different clustering results Sensitivity analysis to parametrization

Final result: 10 clusters

(contingency table fit score of 29,4%)

## Combining valorization and housing types: social and generational mix



## Social and generational mix in Nice



## What explanatory model for the value of the street?

Urban economics: the hedonic regressive model.

- Based on theory of price formation of a dwelling
- Difference between intrinsic and extrinsic characteristics
- Linearity (overcome through Box-Cox transforms), monotonicity, additivity (substitutability), independence
- Hofman et al. (2017): model specification based on unbiased estimation of significative parameters respecting the theoretical model – no attention to predictive power.

#### A different stance (urban micro-geography):

- Use transaction prices to obtain a statistical description of valorisation of spatial units.
- Explain valorisation statistics using only extrinsic characteristics (intrinsic ones of are part of the phenomenon to be explained).

## Hypotheses behind median valorisation

Configurational properties of the street (through-movement, accessibility, tranquility)





# Form of the urban fabric

Prevalence

of small houses



Topography and orientation

#### Proximity to infrastructures





Position within the spatial structure of functions (jobs, commerce, service, transit)

Proximity and accessibility to the Sea



## A Machine Learning Approach

Valorization of street segments:

- Absence of well-established theory (no formal model to adjust, only hypotheses)
- Multi-dimensionality (with interdependencies)
- Non-linearity, non-substitutability



## **Ensemble ML method:**

- Sequential Forward Selection
- Gradient-Boosting Decision Trees
- Network-based Moran I
- SHapley Additive exPlanations

### **Feature Selection**

#### **Sequential Forward Selection**

- From 116 metrics to 28 factors (optimum required between 10 and 30)
- 14,319 units (80% train, 20% test), 5-fold cross validation
- Average mean absolute error 2.38 (target variable between 1 and 20)



## **Gradient Boosting Model**

#### A predictive model based on nested decision trees.

- 14,319 units (80% train, 20% test)
- Hyper-parameter learning, max 128 DT, 10-fold crossvalidation
- Average mean absolute error : 1.05 (train) 2.43 (test)
- R<sup>2</sup> :
  - 0.95 (train) 0.75 (test)



## **Network-based Moran I**

Need to account for all possible spatial structure of target variable.



No spatial autocorrelation of model deviates

20 topological steps: I = 0.08 Significant (p-value = 0.0099) 10 topological steps: I = 0.13 Significant (p-value = 0.0099)

- Helpful to identify missing factors (ex. Distance to University Campus identified at this stage)
- Residual spatial structure at finer scale (not all data were available at the street-segment level)

## Making the model understandable

#### **SHapley Additive exPlanations (SHAP)**

- Local analysis the I/O relationships in the model
- Possibility to analyse SHAP interaction values





## Conclusions

- The value of streets, a question rooted in urban geography, no intrinsic/extrinsic dichotomy.
- The proposed ensemble method: effort to make ML approaches interpretable, without renouncing to predictive power.
- French Riviera: % second homes, distance to the sea, density of commerce/services and concentration of social housing, the most important factors (the first showing a positive relationship, the others an inverse one).
- Perspective of comparative analyses with other urban areas.

## **Limitations & Future Work**

- Inaccuracies from the PERVAL DB
- Loss of information from streets with too few transactions
- Hyperparameters of the GB model set through trial and error by visual evaluation of learning curves... need of algorithmic solution.
- Residual spatial structure at local scale

## Forthcoming:

 Integrate information from streets having too few transactions. Approaches of uncertain knowledge (Bayesian probabilities, possibility theory).

# Thanks for your attention

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