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Data Article

idealista18: A data package with real estate information in three major Spanish markets from the Idealista database*

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ABSTRACT

This dataset contains three items for each of the three major cities in Spain: Madrid, Barcelona and Valencia. First real estate listings published on idealista portal in 2018. All listings have been enriched with cadastral information (i.e. building year of construction, built quality materials grade) plus some geographical features such as distance to relevant city areas and the coordinates themselves. To comply with european personal protection laws, we have processeed some sensible variables yet preserving their spatial properties. The second and third items are a list of points of interest and the administrative boundaries for each municipality. This dataset is suitable to house market analysis, hedonic house price models and other spatial research related with real estate markets.

¡falta cita/referencia al artículo;

Specifications Table

Every section of this table is mandatory. Please enter information in the right-hand column and remove all the instructions

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¹This is author footnote for second author.

Subject	Geography, Economics		
Specific subject area	Spatial analysis, machine learning, hedonic price analysis		
Type of data	Table		
How data were acquired	[State how the data were acquired: E.g. Microscope, SEM, NMR, mass spectrometry, survey* etc. Instruments: E.g. hardware, software, program Make and model and of the instruments used:		
	* if you conducted a survey you must submit a copy of the survey(s) used (either provide these as supplementary material file or provide a URL link to the survey in this section of the table). If the survey is not written in English, please provide an English-language translation.]		
Data format	Spatially masked		
Parameters for data	[Provide a brief description of which conditions were considered for data collection. Max		
collection	400 characters]		
Description of data collection	[Provide a brief description of how these data were collected. Max 600 characters]		
Data source location Data accessibility	Institution: Idealista City/Town/Region: Madrid, Barcelona, Valencia Country: Spain Latitude and longitude (and GPS coordinates, if possible) for collected samples/data: If you are describing secondary data, you are required to provide a list of the primary data sources used in the section. Primary data sources:] Repository name: GitHub Direct URL to data: https://github.com/paezha/idealista18		
Related research article	D. Rey Blanco, P. González Arbues, F. López Hernández, A. Páez, Using machine learning to identify spatial market segments: A reproducible study of major Spanish markets, Comput Environ Urban Syst. In Press.		

Value of the Data

- A cleaned and enriched dataset consisting of real estate listings for three major cities in Spain, constructed to analyze the impacto of using machine learning models to identify spatial market segments to build house price hedonic models.
- The dataset can be used to extend the topic of market segments automatic or semi-automatic identification.
- Official boundaries combined with spatial patterns can be used to analyse the suitability of these boundaries for real estate value purposes.
- The dataset can be enlarged with complementary information to develop hedonic models.
- The data can be processed by quantitative analysis and statistical modeling to study the different factors that affect house prices in the three areas.
- Identification of spatial patterns in the real estate scope using the geo-referenced data points. For either value or urban patterns discovery.

Data Description

The data set is composed by three items for three major spanish cities: quarterly single family listings, neighboorhood polygons and a set of Points of Interest from Open Street Map [1]. All

spatial features, such as polygons and points, are expressed in geodetic coordinates using the *EPSG:4326* coordinate reference system.

The first block integrates property published ads,the files contain the complete offering for each of the four quarters in 2018 on idealista web site [2]. Idealista is the major real estate listing portal in Spain and also present in other southern european countries as Italy and Portugal. Each record contains all attributes the listing ad has plus a number of additional attributes from the Spanish cadastre [3], the names for all these variables start with the prefix *CAD*. Cadastral features assignment is done by assigning the features of the nearest parcel to the coordinates *LATITUDE* and *LONGITUDE*.

Variable M	lesurement scale	Description
ASSETID	Identifier	Unique identifier of the advertisement
PERIOD	Nominal (Date)	Expressed as YYYYMM, indicates the quarter when the ad was extracted we used YYYY03 for the 1st Q, YYYY06 the 2nd, YYYY09 for the 3rd and YYYY12
		for the 4th
PRICE	Interval	Asking price for the ad at idealista expressed in euros
UNITPRICE	Interval	Asking price in euros per square meter (constructed area)
ADTYPOLOGYID	Nominal	Residential building type: multi-family: <i>home</i> , single-familiy: <i>chalet</i>
ADOPERATIONID	Nominal	Operation type for the ad: sale or rent
ROOMNUMBER	Ordinal	Number of bedrooms
BATHNUMBER	Ordinal	Number of bathrooms
HASTERRACE	Nominal	Dummy variable for terrace (takes 1 if there is a terrace, 0 otherwise
HASLIFT	Nominal	Dummy variable for lift (takes 1 if there is a lift in the building, 0 otherwise)
HASAIRCONDITIONING	Nominal	Dummy variable for AA (takes 1 if there is a AA, 0 otherwise)
AMENITYID	Nominal	Indicates the amenities included (1 - no furniture, no kitchen amenities, 2 - kitchen amenities, no furniture, 3 - kitchen amenities, furniture)
HASPARKINGSPACE	Nominal	Dummy variable for parking (takes 1 if parking is included in the Ad, 0 otherwise)
ISPARKINGSPACEINCLUDEDINPRIC	E Nominal	Dummy variable for parking (takes 1 if parking is included in the Ad, 0 otherwise)
PARKINGSPACEPRICE	Interval	Price of parking space in euros
HASNORTHORIENTATION	Nominal	Dummy variable for orientation (takes 1 if orientation is North in the Ad, 0 otherwise) - Important note: orientation features are not orthogonal features, a house oriented to the north can be also oriented to the east
HASSOUTHORIENTATION	Nominal	Dummy variable for orientation (takes 1 if orientation is South in the Ad, 0 otherwise) - Important note: orientation features are not orthogonal features, a house oriented to the north can be also oriented to the east
HASEASTORIENTATION	Nominal	Dummy variable for orientation (takes 1 if orientation is East in the Ad, 0 otherwise) - Important note: orientation features are not orthogonal features, a house oriented to the north can be also oriented to the east
HASWESTORIENTATION	Nominal	Dummy variable for orientation (takes 1 if orientation is West in the Ad, 0 otherwise) - Important note: orientation features are not orthogonal features, a house oriented to the north can be also oriented to the east

HASBOXROOM	Nominal	Dummy variable for boxroom (takes 1 if boxroom is included in the Ad. 0 otherwise)
HASWARDROBE	Nominal	Dummy variable for wardrobe (takes 1 whether the property has wardrobes, 0 otherwise)
HASSWIMMINGPOOL	Nominal	Dummy variable for swimming pool (takes 1 if swimming pool is included in the Ad, 0 otherwise)
HASDOORMAN	Nominal	Dummy variable for doorman (takes 1 if there is a doorman in the building, 0 otherwise)
HASGARDEN	Nominal	Dummy variable for garden (takes 1 if there is a garden in the building, 0 otherwise)
ISDUPLEX	Nominal	Dummy variable for bachelor apartment (referred as stu- dio in Spain) (takes 1 if it is a bachelor apartment, 0 oth- erwise)
ISINTOPFLOOR	Nominal	Dummy variable indicating if the apartment is located in the top floor (takes 1 on the top floor 0 otherwise)
CONSTRUCTIONYEAR	Interval	Construction year (source: advertiser)
FLOORCLEAN	Ordinal	Indicates flat floornumber starting from the 0 value for ground floor (source: advertiser)
FLATLOCATIONID	Nominal	Indicates the kind of views the flat has (1 - external, 2 - internal)
CADCONSTRUCTIONYEAR	Interval	Construction year as of cadastral source (source: cadastre), note this figure can differ from the one given by the advertiser
CADMAXBUILDINGFLOOR	Ordinal	Max building floor (source: cadastre)
CADDWELLINGCOUNT	Interval	Dwelling count in the building (source: cadastre)
CADASTRALQUALITYID	Ordinal	Cadastral quality (source: cadastre). 0 Best - 10 Worst
BUILTTYPEID_1	Nominal	Dummy value for flat condition: 1 new development and 0 otherwise
BUILTTYPEID_2	Nominal	Dummy value for flat condition: 1 second hand to be restored 0 otherwise (<i>source</i> : <i>advertiser</i>)
BUILTTYPEID_3	Nominal	Dummy value for flat condition: 1 second hand in good condition 0 otherwise (<i>source: advertiser</i>)
DISTANCE_TO_CITY_CENTER	Interval	Distance to center of city in Km
geometry	Geometry	Geometry for the elements. A point with X, Y coordinates

In addition to the common features each city will have a set of spatial additional features:

City	Variable	Mesurement scale	Description
Madrid	DISTANCE_TO_CASTELLANA	Interval	Distance in km to the Paseo de la Castellana Street
Valencia	DISTANCE_TO_METRO DISTANCE_TO_BLASCO	Interval Interval	Distance in km to the nearest subway station Distance in km to the Blasco Ibáñez Avenue
Barcelona	DISTANCE_TO_DIAGONAL	Interval	Distance in km to the Diagonal Avenue

The second part contains the polygons for the different neighborhoods for such cities as we can see in the figure 1. This boundaries are not the official boundaries but the official boundaries used by idealista portal. In practical terms we can assume they are the same, since the website simply collapsed those areas if they are small enough in terms of number of ads. In the case of Madrid they just collapse four areas in two new ones.

We have a total of 73 neighborhoods for Barcelona, 135 for Madrid and 73 for Valencia. Each neighborhood has also two additional variables:

Fig. 1. Neighborhood boundaries for Madrid, Barcelona and Valencia. Source: own elaboration



Variable	Mesurement scale	Description
LOCATIONID	nominal	Unique identifier for the neighborhood
LOCATIONNAME	nominal	Neighborhood name

Experimental Design, Materials and Methods

The data encompassed three major cities in Spain: Madrid, Barcelona and Valencia, for each municipality we provide listing prices for multi-familiy homes on a quarterly basis. The files contain the complete offering for each of the four quarters in 2018 on idealista web site [2]. Idealista is the major real estate listing portal in Spain and also present in other southern european countries as Italy and Portugal.

Given regulatory restriction idealista listings are slighly anonymized, in order to agree regulatory requirements guaranteeing their attributes and spatial properties. This process take two steps, the first consists of the obfuscation of prices adding or substracting a random percentage of their original values ranging from -2.5% to +2.5%. Since asking prices are not normally a completely continuous magnitude (sale prices are usually multiples of 1000 and rent prices are of 10), after the first price modification we finally align prices to multiples of 1000.

Finally we carry out a spatial masking process that intend to keep spatial properties of the original data set. We essentially displace the coordinates of each listing with a minimum distance and a maximum distance with the restriction that every new location must fall in within the original neighborbood of the add. The process is described in the algorithm below:

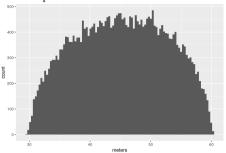


Fig. 2. Coordinate displacement in meters Valencia. Source: own elaboration

Data: all idealista listings

Result: all idealista listings with masked coordinates

initialization;

for each listing L do

take geographical location of L as (X, Y) repeat

take a random angle α from 0 to 360 degrees take a distance R as a random value from 30 to 60 meters determine a new point (X', Y') calculated as a point located R with the angle α until this stop condition;

set (X', Y') as the new location for the listing L

rend

Algorithm 1: Coordinate displacement process for anonymisation purposes

In the figure 2 we display the histogram with the displacement in meters for all listings in the city of Valencia, as average the displacement is 45 meters

Ethics Statement

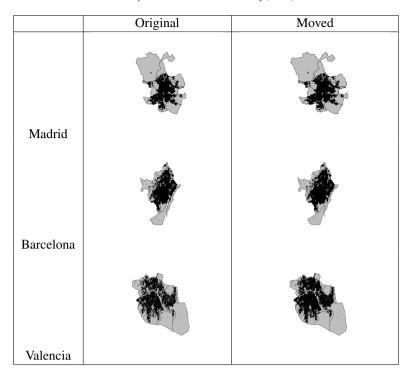
[Please refer to the journal's Guide for Authors for more information on the ethical requirements for publication in Data in Brief. In addition to these requirements:

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If the work involved animal experiments: please include a statement confirming that all experiments comply with the ARRIVE guidelines and were be carried out in accordance with the U.K. Animals (Scientific Procedures) Act, 1986 and associated guidelines, EU Directive 2010/63/EU for animal experiments, or the National Institutes of Health guide for the care and use of Laboratory animals (NIH Publications No. 8023, revised 1978)]

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Acknowledgments should be inserted at the end of the paper, before the references, not as a footnote to the title. Use the unnumbered Acknowledgements Head style for the Acknowledgments heading.



Declaration of Competing Interest

[All authors are required to report the following information:

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- (2) All financial relationships with any entity that could be viewed as relevant to data described in this manuscript;
- (3) All sources of revenue with relevance to this work where payments have been made to authors, or their institutions on their behalf, within the 36 months prior to submission;
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Note

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References

- [1] OpenStreetMap contributors, Planet dump retrieved from https://planet.osm.org , https://www.openstreetmap.org, 2017.
- [2] idealista, http://www.idealista.com, http://www.idealista.com, 2018.
- [3] Registro Central del Catastro, https://www.sedecatastro.gob.es/, 2021.