A geo-referenced micro-data set of real estate listings for Spain's three largest cities - Response to referees

Referee: 1

Comments to the Author

Dear Authors, Thank you for the opportunity to review this paper. I would appreciate it if you would highlight the changed text in the next submission. Otherwise it is difficult to identify which parts have changed. The unknown quality of this one year randomised data is the main issue for potential users and I repeat my suggestion that this data should be validated for the benefit potential readers and users. Below is a list of comments and questions which need to be addressed.

Thank you for your constructive comments. Below we describe how we improved the paper in response.

1. Please clearly state the nature of the data in the abstract. It is worthwhile to mention the relocated and randomised dataset in the abstract. This house price dataset only has 189,923 records, which could be not really be described as a big dataset as you describe it in the abstract. Except the authors could provide an evidence/reference on why this is big dataset.

Response:

We appreciate this comment. There are a few publicly available datasets for hedonic price analysis, but they tend to be smaller. For example, Bonifaci and Copiello (2015) includes n=1,042 observations for Padua, in Italy; Del Giudice, De Paola, and Forte (2018) share a dataset with n=576 observations relating to rental prices in Naples, Italy; Solano S'anchez et al. (2019) present a dataset with n=1,623 daily rental prices in Seville, Spain; and the dataset of Fuerst and Haddad (2020) includes n=4,201 property prices geocoded to the level of nine regions in England and Wales. The most similar in terms of geographical disaggregation and sample size is the dataset of Song et al. (2021) which includes transactions for four cities in South Korea, namely Busan (n=61,152), Daegu (n=32,363), Daejeon (n=21,114), and Gwangju (n=25,984). In response to your comment we drop the terminology of "big data" and describe our dataset as "large" instead.

2. Please check the typographical errors before your next submission. For example, the "thirdr" in your table1.

Response:

Thank you for highlighting this. We have carefully proof-read the document for typos and other errors.

3. Please use part of your answer to my previous fourth question in the manuscript. Your answer is quite useful for your future users, especially those researchers who are not familiar with house price data in Spain. I quote your previous answer below:

"idealista covers fairly well all segments in the Spanish market, both individual and professional advertisers. This dataset comprehends information about listing prices, thus it contains the market situation from the offer. Nevertheless, actual sale prices information is not publicly available as it is information that can only be accessed by paying high fees to Colegio de Registradores. However listing prices reflect quite well the (transaction) reality of real estate markets using the

offer perspective, they keep strong correlation between idealista and transaction prices can be established see Banco de España. Even though a great extent of correlation does exist between official transactional and asking records, both databases can be taken as complementary and are of great interest when studying asking-transaction price gaps or the relation between listing site demand variables (ie. ad contacts or ad views) and price gaps. The main real estate portals in Spain are idealista and Fotocasa, further away are Habitaclia and Milanuncios, the latter focusing almost exclusively on private individuals. In September 2021, according to data from Similarweb (a site specialized in sites' traffic volume comparison), in Spain, there were a total of 103 million page views on real estate portals, out of page views on real estate portals, of which the four main portals, idealista, Fotocasa, Habitaclia, and Pisos.com, in that order, accounted for 94% of the traffic. Another relevant feature of this sector is that it is highly concentrated, idealista being the leader by far with 58.6 monthly million visits (57% of the total traffic) versus its immediate competitor with 19.9 million visits (19.3 % of total traffic). In terms of the evolution over time of interest in the content of each portal, based on data from Google Trends, the leadership of idealista Trends, the leadership of idealista"

Response:

Thank you for this suggestion. We have adapted this answer to expand the text in the section "Data Description".

4. For data description, it is worth showing solid evidence of the quality of this dataset. For example, what is the data coverage of this data compared to all the residential properties in the city? Listing the advertised volume is not very helpful for users who wish to understand the quality.

Response:

To better place in context the coverage of the {idealista18} dataset, Table 1 now shows the number of listings with respect to the total residential stock in each city (in 2018). As seen in the table, the number of listings ranges between 6.1% of the total number of properties (in Madrid) and 8.1% (in Valencia). Information from Instituto Nacional de Estadistica¹ for 2018 also shows that the number of transactions in 2018 are equivalent to 81.3% of idealista listings in Barcelona, 80.8% in Madrid, and 91.1% in Valencia. While it is not possible to link actual transactions to idealista listings, this gives a quantitative sense of the coverage of the data package.

Table 1: Total properties and transactionst three Spanish cities. Year 2018

City	Total properties (P)	Total transactions (T)	Listings (L)	L/P	$\mathrm{T/L}$
Barcelona	789,740	56,012	61,329	7.8%	81.3%
Madrid	1,545,397	76,603	94,802	6.1%	80.8%
Valencia	416,004	30,615	$33,\!593$	8.1%	91.1%
Total	2,751,141	163,230	189,724	6.9%	86.0%

Sources

Total properties (P): Ministerio Español de Hacienda y Función Pública

Total transactions (T): Instituto Nacional de Estadística

5. Given the asking price is a resized version of the original asking price with a random percentage between -2.5% and +2.5%, please elaborate on how this influences your quantitative research results and findings. Please provide critical evidence to explain how this randomised price will not influence academic research. For example, how does the house price variation at local level change after the price is randomised. To what degree will this influence a typical

¹https://www.ine.es/jaxiT3/Tabla.htm?t=6150&L=1

hedonic house price research approach? I am concerned about whether this randomised data could bias conclusions.

Response:

We appreciate the opportunity to delve more deeply into this issue, as it may be of interest to users of the data package. The short version of this story is that adding a small amount of random noise to the prices $(\pm 2.5\%)$ does not introduce bias but does increase the variance of the masked variable. This can be shown with some standard algebra of random variables.

Now the longer story.

The masked prices (P) in the dataset are calculated as the sum of two random variables, the original raw prices (RP) and a random masking variable (η) . The masking variable is generated as follows:

$$\eta = RP \cdot \epsilon$$

The masking variable uses values drawn at random from the uniform distribution with parameters b = 0.025 and a = -0.025:

$$f(\epsilon) = \begin{cases} \frac{1}{b-a} & \text{for } a \le \epsilon \le b\\ 0 & \text{otherwise} \end{cases}$$

The effect of ϵ is to produce a random percentage of the raw price ($\pm 2.5\%$). The expectation of ϵ given these parameters is:

$$E[\epsilon] = \frac{a+b}{2} = \frac{-0.025 + 0.025}{2} = 0$$

and the variance is:

$$V[\epsilon] = \frac{(b-a)^2}{12} = \frac{25}{120000}$$

The masked prices P are the raw prices plus the masking variable:

$$P = RP + RP \cdot \epsilon = RP + n$$

Since the mean of ϵ is zero, and RP and ϵ are independent, it turns out that the mean of the masking variable is zero:

$$E[\eta] = E[RP \cdot \epsilon] = E[RP] \cdot E[\epsilon] = 0$$

It follows that the masked prices are unbiased since their expectation is identical to the expectation of the raw prices:

$$E[P] = E[RP] + E[\eta] = E[RP]$$

Now let us examine the covariance. Again, since RP and ϵ are independent, the variance of the masking variable η is:

$$V[\eta] = V[RP \cdot \epsilon] = V[RP] \cdot V[\epsilon] + V[RP] \cdot (E[\epsilon])^2 + V[\epsilon] \cdot (E[RP])^2$$

Considering that $E[\epsilon] = 0$, then:

$$V[\eta] = V[RP] \cdot V[\epsilon] + V[\epsilon] \cdot (E[RP])^2 = V[\epsilon] \cdot (V[RP] + (E[RP])^2) = \frac{25}{120000} (V[RP] + (E[RP])^2)$$

Accordingly, the variance of the masked prices is:

$$V[P] = V[RP + \eta] = V[RP] + V[\eta] + 2Cov[RP, \eta]$$

If we recall that $E[\eta] = 0$ and $E[\epsilon] = 0$, the covariance of RP and η can be shown to be zero:

$$\operatorname{Cov}[RP,\eta] = \operatorname{E}[RP \cdot \eta] - \operatorname{E}[RP] \cdot \operatorname{E}[\eta] = \operatorname{E}[RP \cdot \eta] = \operatorname{E}[RP^2 \cdot \epsilon] = \operatorname{E}[RP^2] \cdot \operatorname{E}[\epsilon] = 0$$

From the above we can conclude that the variance of the masked prices is:

$$V[P] = V[RP] + V[\eta] = V[RP] + \frac{25}{120000} \left(V[RP] + (E[RP])^2 \right) = \frac{120025}{120000} \cdot V[RP] + \frac{25}{120000} \cdot (E[RP])^2$$

which is 0.0208 greater than the variance of the raw prices, plus 0.0208 of the square of the mean of the raw prices.

What does this mean for users of the data? The masked prices are an unbiased version of the raw prices, but they have a somewhat larger variance than the original (unreported) prices.

6. In figure 1, is it possible to standardise this data rather than showing only the number of dwelling records in your house price data? I guess the city centre always has high advertised volumes but it also has a high density of dwellings. It is worthwhile to show values standardised by the number of dwellings.

Response:

After this comment we included a new figure in the paper to display the percentage of listings relative to the total number of dwellings by district (see Figure 1 below).

Thank you again for your generous feedback. We hope that our responses will prove satisfactory and you will agree that the paper makes a worthwhile contribution to growing catalog of datasets to support research on real estate markets.

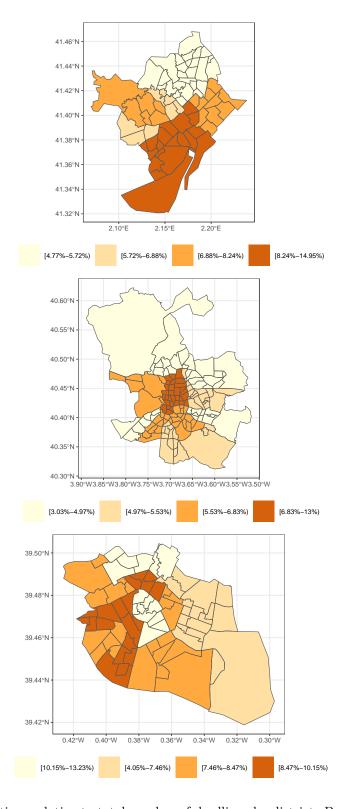


Figure 1: Percentage of listings relative to total number of dwellings by district. Boundary for Barcelona (Top), Madrid (Center), and Valencia (Bottom).

References

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