

The Demand for Air Quality: Evidence from the Housing Market in Bogotá, Colombia

Fernando Carriazo^{1*} and John Alexander Gomez²

¹Universidad del Rosario, Calle 12C No. 6-25 - Bogotá D.C. Colombia and ²Universidad de Los Andes, Bogotá, Colombia.

*Corresponding author. Email: fernando.carriazo@gmail.com

ONLINE APPENDIX

1. Theoretical construction of the Hedonic Price Model (HPM)

Taylor (2003) offers the following description of the HPF: if Z is a ‘differentiated product’ which is determined by its characteristics, this means that

$$Z = (z_1, \dots, z_n), \quad (A1)$$

where n is the number of characteristics of the differentiated good.

At the same time, it is assumed that Z is sold in a competitive market. The price schedule or price function for a characteristic is denoted as $p(Z): \mathbb{R}^n \rightarrow \mathbb{R}$. Kolstad (2000) affirms that, “The price function is an equilibrium concept, resulting from the interaction of supply and demand. In much the same way that conventional supply and demand interact to produce a single market price, forces of supply and demand interact to produce a price function...” Thus, the consumer is price acceptant because $p(Z)$ is exogenous.

The utility of the consumers depends on the differentiated good Z and on a composite good X^1 (Palmquist, 2005). Additionally, this utility is affected by the socioeconomic characteristics of the agent, which are denoted as α_i . Thus, the utility for the individual i is defined by

$$U^i(X, Z, \alpha_i): \mathbb{R}^{n+1} \rightarrow \mathbb{R}. \quad (A2)$$

At the same time, in the model it is assumed that the consumer buys just one unit of the good Z , so the income restriction can be expressed like:

$$y^i = X + p(Z) \quad (A3)$$

Palmquist (2005) affirms that in the presence of consumers with more than one residence, it is assumed that each one of them has a different purpose. Accordingly, they will enter separately in the utility function.

¹ Palmquist (2005) describes X as a “Hicksian composite good”.

Thus, the consumer's problem is to

$$\max_{X,Z} U^i(X, Z, \alpha_i) \text{ s.t. } y^j = X + p(Z), \quad (A4)$$

so that n of the n+1 first order conditions could be summarized in the following expression

$$\frac{\partial p(Z)}{\partial z_j} = p_j = \frac{\frac{\partial U^i(X, Z, \alpha_i)}{\partial z_j}}{\frac{\partial U^i(X, Z, \alpha_i)}{\partial X}} \text{ para } j = 1, \dots, n \quad (A5)$$

The consumer has to decide which is his optimal bid for the house; this means its willingness to pay. The consumer's "*Bid Function*" describes the relationship between the bid that the consumer would make for Z, based on changes on one or more of its characteristics, keeping constant the utility and its income (Taylor, 2003). Thus, if $y_0^i - \theta^i = X$, the utility of the agent i can be expressed as:

$$U^i(y_0^i - \theta^i, Z, \alpha_i) \equiv U_0^i, \quad (A6)$$

from (A6):

$$\theta^i = \theta^i(Z, y_0^i, U_0^i, \alpha_i). \quad (A7)$$

From the previous equations, the consumer problem can be expressed as

$$\max_Z U^i(y^i - \theta^i(Z, y^i, U^i, \alpha_i), Z, \alpha_i), \quad (A8)$$

the first order conditions of the maximization problem in (A8)

$$\frac{\partial U^i(\cdot)}{\partial X} (-1) \frac{\partial \theta^i(\cdot)}{\partial z_j} + \frac{\partial U^i(\cdot)}{\partial z_j} = 0 \text{ para } j = 1, \dots, n \quad (A9)$$

This implies,

$$\frac{\partial \theta^i(Z, y^i, U^i, \alpha_i)}{\partial z_j} = \frac{\frac{\partial U^i(X, Z, \alpha_i)}{\partial z_j}}{\frac{\partial U^i(X, Z, \alpha_i)}{\partial X}} \text{ para } j = 1, \dots, n. \quad (A10)$$

By comparing (A5) to (A10), in the optimum

$$\frac{\partial p(Z)}{\partial z_j} = \frac{\partial \theta^i(Z, y^i, U^i, \alpha_i)}{\partial z_j} = \frac{\frac{\partial U^i(X, Z, \alpha_i)}{\partial z_j}}{\frac{\partial U^i(X, Z, \alpha_i)}{\partial X}} \quad (A11)$$

for each consumer and characteristic. This means that at the optimum, the consumer's offer for one more unit of the z_j characteristic ($\frac{\partial \theta^i(Z, y^i, U^i, \alpha_i)}{\partial z_j}$) is exactly equal to the marginal effect of this change on the price of Z . In other words, what the consumer pays is equivalent to what the market will charge him (her).

2. Robustness Checks

There are at least two identification issues to consider: the evidence in favor of the existence of three submarkets defined by type of dwelling, and how results might change using an alternative measure of air quality, or different functional forms for the HPF.

A Chow test was used to check for structural changes in the HPF among the housing submarkets. By calculating an equivalent Chow test for the frontier model of the general case, we found that a structural change exists ($\chi^2 = 354.7$) between the three submarkets based on the specified functional form. Another exercise was to use the Two Step Clustering method as in Poudyal *et al.* (2009). We estimated the optimum number of clusters and the cluster structure using cadastral data, which captures the census of all Bogotá's dwellings. The methodology predicts three submarkets: apartments (cluster 1), houses without horizontal property (cluster 2), and houses with horizontal properties, similar to the market structure that this document uses.²

We also used as a robustness check different functional forms of the hedonic function as well as different air quality measures. As suggested by Cropper *et al.* (1988) and Kuminoff *et al.* (2010), simple functional forms such as semi-log or log models are preferred. Air quality

² Please see table A7 in the online appendix reporting the output of the cluster methodology.

parameters were negative and significant for most of the functional forms' specifications and submarkets. We also compared the results of the FS using OLS estimations to the FS frontier model. OLS results are also consistent with a negative and statistically significant effect of PM10 concentration on housing rents; the implicit price elasticity for air quality based on OLS estimations were 21 per cent higher than in the benchmark frontier model specified to mitigate Omitted Variable Bias.

Regarding various air quality measures, table RC1 reports the air quality estimated parameter using the annual PM10 concentration average, interpolated by VNA (column 1, the same results are reported in table 2), daily mean concentration interpolated by Inverse Distance Weighting (IDW) (column 4), which is a more standard interpolation method, mean maximum daily concentration (column 5), and mean daily concentration of PM10 during the EMB time framework. For all variables and submarkets the price-pollution relationships are still negative and almost all are statistically significant at least at the 10% level.³

Table RC1. *First Stage with different pollution measures*

	β	IC-	IC+	β - IDW	β - Mean Survey	β - Annual Max
General	-0.01537	-0.00979	-0.02095	-0.00914 (0.00189)	-0.00649 (0.00549)	-0.00303 (0.00021)
Apartment	-0.00612	-0.00045	-0.01180	-0.00187 (0.51649)	-0.00050 (0.82776)	-0.00081 (0.39885)
Condos	-0.02764	0.02033	-0.07561	-0.05101 (0.00206)	-0.02935 (0.01607)	-0.00830 (0.10124)
No-condos	-0.02165	-0.00944	-0.03386	-0.02299 (0.00319)	-0.01344 (0.04163)	-0.00812 (0.00007)

Relevance test p-value in parentheses

³ Table A9 in the online appendix goes further by exploring the robustness of results to different functional forms.

The SS demand function is robust to the measurement of air quality (using IDW results), to changes of functional form (log-log specification), to model specification for FS estimation (using results of OLS with Spatial fixed effects) and to changes of instruments for the SS (using only precipitation as the instrument). None of the mentioned changes had major consequences on the significance of the negative relationship between price and quantity of the estimated demand.⁴ Our main goal was the estimation of Total Willingness to Pay (TWP) for PM10 reductions, thus, robustness should be seen primarily in terms of these results. Table RC2 below summarizes the welfare estimations for the general case of the proposed estimation and the robustness exercises. In some cases, results converge either to similar elasticities or to similar welfare measures. For example, when using data from the IDW interpolation method, welfare measures for achieving both the national standard and the WHO standard are higher but not much higher than those derived by the proposed model. Interestingly, welfare measures are quite robust to different instrument specifications. For example, welfare measures and elasticity estimations do not change significantly between the benchmark specification and a specification using only wind speed but not raining data.

Table RC 2. Total Willingness to Pay (robustness exercises)

	Elasticity (%)	TWP 20µg/m ³	for	TWP 50µg/m ³	for
Proposed Model	-1.306023	189.64	[21.04]	12.16	[1.35]
Log – Log	-1.911331	198.29	[22]	32.88	[3.65]
Log – Lin	-2.007038	115.51	[12.82]	6.33	[.7]
IDW	-.9417697	207.47	[23.02]	13.62	[1.51]
OLS with spatial fix effects	-1.582098	94.26	[10.46]	5.97	[.66]
Instrumented by wind speed and spatial lag	-1.311581	189.38	[21.02]	12.14	[1.35]
Instrumented just by wind speed	-1.030766	206.35	[22.9]	13.5	[1.5]
Percentage of monthly income in square brackets					

⁴ Tables A9 to A14 in the online appendix show different estimations, changing the measurement of air quality, functional form, and instruments.

3. Additional tables and figures

Table A1. *Dwelling and neighborhood characteristics*

Variables	Description	N	Mean	SD	Min.	Max
<i>Structural and Neighborhood Characteristics</i>						
arriendo	Living in rent (Yes or no)	11,643	.3865	.487	0	1
numCuartos	Number of rooms	11,643	3.63	1.324	1	15
banios	Number of bathrooms	11,631	1.593	.8211	1	9
garaje	Garage (Yes or no)	11,643	.4109	.492	0	1
arriendoReal	Rent for renters (US\$)	4,500	251.3	219.5	12	3,790
arriendoAuto	Hypothetical rent (US\$)	7,143	395.6	380.7	14	3,790
PArrien	Aggregated rent (US\$)	11,643	339.8	335.4	12	3,790
ascensor	Elevator (Yes or no)	11,643	.1328	.3394	0	1
terrazza	Terrace (Yes or no)	11,643	.2854	.4516	0	1
zonVerdes	Green areas (Yes or no)	11,643	.2949	.456	0	1
numTelFijo	Number of fixed phones	11,643	.8012	.5068	0	5
internet	Internet (Yes or no)	11,643	.4738	.4993	0	1
conjunCerr	In a residential complex (Yes or no)	11,643	.3007	.4586	0	1
jardin	Garden (Yes or no)	11,643	.4084	.4916	0	1
pisos	Number of levels in the building	11,643	3.479	2.909	1	30
plancha	Plate (Yes or no)	11,544	.8482	.3588	0	1
humedad	Moistness (Yes or no)	11,619	.3581	.4795	0	1
inseguridad	Insecurity (Yes or no)	11,643	.757	.4289	0	1
homicidios	Mean number of murders during last ten years in the block	11,643	330.8	239.4	0	1,900
precipitacion	2011 mean rainfall (mm/month)	11,643	.3341	.8821	0	14
ruido	Noise (Yes or no)	11,643	.3673	.4821	0	1
malOlor	Odor (Yes or no)	11,643	.3832	.4862	0	1
indComSer	Near industry, commerce or services (Yes or no)	11,643	.3728	.4836	0	1
anunciosExc	Advertising (Yes or no)	11,643	.08142	.2735	0	1
invasionAndenes	Advertising (Yes or no)	11,643	.1611	.3677	0	1
cuartDormi	Number of rooms available for sleeping	11,643	2.2107	0.9413	1	9
calentador	Water heater (Yes or no)	11,643	0.5424	0.4982	0	1
aireLib	Access to terrace or Green areas (Yes or no)	11,643	0.7882	0.4086	0	1
cocinaExcl	Private kitchen (Yes or no)	11,643	1.9861	1	0	2
fallViv	Dwelling present moistness, sinking or pipeline failures (Yes or no)	11,628	0.3967	0.4892	0	1
puntaje	Mean building's quality score in the block level	11,643	41.32	15.03	10	91
distFuenCont	Distance to fix sources of pollution (m)	11,643	751.5	571.5	0	3,365
<i>Categorical variables</i>						
matPisos	Floors predominant material					
matParedes	Walls predominant material					
ubiAgua	Water source in the house					
dondCocina	Cooking area					
combCocina	Cooking combustible					
viaEstado	Quality of dwellings' roads of access					
viallumin	Quality of ilumintaion in the road of access					
localidad	Localitie (Bogota's principal political division) where the dwelling is located					

Table A2. *Description of socio-demographic variables*

Variable	Description	N	Mean	SD	Min.	Max.
sexoJH	Sex of HH (1=man)	11,643	.616	.4864	0	1
edadJH	Age of HH	11,643	48.23	15.09	16	99
meanAge	Mean age within families	11,643	35.97	15.49	8	92
tiemBogJH	Time living in Bogotá (years)	11,643	37.14	19.77	0	99
perHog	Number of people per house	11,643	3.382	1.64	1	19
hog_hijos	Number of sons	11,643	1.297	1.127	0	9
ingresos	Income (US\$)	10,993	901.1	1,554	0	54,143
menores_5	Number of children under 5y	11,643	.2286	.4987	0	4
men5Tos	Children under 5y with cough during the last week	11,643	.0931	.3228	0	3
men5DifRes	Children under 5y with breath difficulties during the last week	11,643	.02499	.1683	0	3
hog_orgAmb	Number of family members in environmental organization	11,643	.01589	.1369	0	3
trabaja	Marital status of HH	11,507	.717	.4505	0	1

Table A3. *PM10 Air Quality data*

Variable	Description	N	Mean	SD	Min.	Max
meanStock	PM10 Annual Concentration (ug/m3)	11,643	52.84	9.294	38	84
meanEnc	PM10 mean dailly Concentration while EMB (ug/m3)	11,643	59.3	9.724	40	88
maxStock	PM10 Max Concentration (ug/m3)	11,643	255.4	39.48	176	376
inter2meanPM	PM10 concentration by VNA (ug/m3)	6,077	48.64	7.718	38	84
SO2	SO ₂ mean annual concentration (PPB)	11,643	2.9	1.142	0.934	8.365
CO	CO mean annual concentration (PPM)	11,643	0.9	0.128	0.657	1.507
O3	O ₃ mean annual concentration (PPB)	11,643	11.09	1.925	7.97	16.398

Table A4. *Correlation matrix between pollutants*

	meanStock	inter2meanPM	CO	O3	SO2
meanStock	1				
inter2meanPM	0.9573 (0.0000)	1			
CO	0.8383 (0.0000)	0.718 (0.0000)	1		
O3	-0.603 (0.0000)	-0.5992 (0.0000)	-0.3357 (0.0000)	1	
SO2	0.9072 (0.0000)	0.8397 (0.0000)	0.9453 (0.0000)	-0.4303 (0.0000)	1

P-value of individual significance test in parentheses

Table A5. *Hedonic Price Function estimation*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
lnrent				
PM10	-.015372*** (0.0028)	-.0061217** (0.0029)	-.024383* (0.0131)	-.021776*** (0.0062)
inter2meanPM_arr	.0029179** (0.0013)	.0021535 (0.0015)	.0053419 (0.0060)	.0025825 (0.0033)
arriendo	-.5351** (0.2471)	-.40505 (0.2627)	-1.1933** (0.5495)	-.87396 (0.5404)
numCuartos	.065031*** (0.0059)	.069829*** (0.0082)	.029764 (0.0218)	.049269*** (0.0100)
pisos	-.010301*** (0.0031)	.0071155*** (0.0024)		-.0019179 (0.0221)
banios	.17178*** (0.0089)	.16237*** (0.0116)	.1287*** (0.0317)	.12296*** (0.0169)
cuartDormir	.0042259 (0.0068)	-.0011395 (0.0086)	-.0047237 (0.0236)	.0063024 (0.0124)
calentador	.042664*** (0.0140)	.013347 (0.0170)		
aireLib	.064505*** (0.0153)	.034332** (0.0156)	.0904 (0.0681)	.038869 (0.0425)
ascensor	.090834*** (0.0238)			
numTelFijo	.043708*** (0.0144)	.071417*** (0.0164)		.022028 (0.0293)
cocinaExcl	.22167** (0.1008)	.14947 (0.1078)		.14607 (0.1948)
internet	.063874*** (0.0134)	.059659*** (0.0155)		.097931*** (0.0267)
DmatPisos3	-.022391 (0.0162)	-.027149 (0.0175)	-.092397** (0.0443)	-.012973 (0.0410)
DmatPisos4	-.028231 (0.0224)	-.0045639 (0.0242)	-.33872*** (0.0783)	-.018855 (0.0540)
DmatPisos5	-.010598 (0.0257)	.0067745 (0.0285)	-.14214* (0.0836)	-.026525 (0.0584)
DmatPisos6	-.17272*** (0.0470)	-.27322*** (0.0743)	-.088762 (0.1198)	-.12429 (0.0792)
DmatPisos7	-.070204 (0.2817)	.30109 (0.2794)		-.16928 (0.2851)

Table A5. *Hedonic Price Function estimation (continued)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
fallViv	-.014298 (0.0122)	-.015083 (0.0143)		-.0003099 (0.0246)
DmatParedes2	-.019264 (0.0264)	-.037212 (0.0307)		
DmatParedes3	-.0011155 (0.0123)	-.0099306 (0.0141)		
DmatParedes4	-.019484 (0.0444)	-.0029611 (0.0651)		
DmatParedes5	-.37901*** (0.1310)	-.27076 (0.2551)		
DmatParedes6	.08459 (0.1717)	.0035294 (0.3793)		
DmatParedes7	-.88938* (0.4971)			
DubiAgua2	-.0027236 (0.0337)	-.013971 (0.0397)	.172* (0.0956)	-.054736 (0.0677)
DubiAgua3	-.012483 (0.1081)	-.012386 (0.1068)		-.02587 (0.3698)
DubiAgua4	-.086798 (0.1494)	-.098687 (0.1337)		-.17851 (0.2703)
DdondCocina2	-.014937 (0.1109)	.13692 (0.1291)		-.23998 (0.2103)
DdondCocina3	-.10774 (0.0988)	.17592 (0.1388)		.07899 (0.2397)
DdondCocina4	-.065143 (0.2645)	-.047105 (0.2191)		-.26143 (0.3911)
DdondCocina5	-.024636 (0.2645)	.025241 (0.2213)		-.16083 (0.2449)
DdondCocina6	-.22511 (0.2773)			-.31676
homicidios	-.017381** (0.0075)	-.011348 (0.0082)	-.099515** (0.0390)	-.006772 (0.0174)
inseguridad	-.0083373 (0.0134)	-.022874 (0.0147)	.070555* (0.0404)	.011359 (0.0308)
distFuenCont	.0000258 (0.0000)	.0000296 (0.0000)		-.0000264 (0.0000)

Table A5. *Hedonic Price Function estimation (continued)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
puntaje	.011197*** (0.0008)	.011308*** (0.0009)	.015223*** (0.0024)	.016965*** (0.0022)
indComSer	-.0069347 (0.0124)	-.017009 (0.0143)		
anunciosExc	-.0079866 (0.0199)	-.040132* (0.0232)		
invasionAndenes	.021109 (0.0151)	-.007327 (0.0168)		
DviaIllumin2	.0034486 (0.0132)	-.020031 (0.0154)		
DviaIllumin3	-.0097587 (0.1194)	-.0052178 (0.1498)		
DviaEstado2	-.021764 (0.0155)	-.038892** (0.0188)	.04645 (0.0537)	
DviaEstado3	-.031181 (0.0259)	.0087599 (0.0319)	-.2043 (0.1275)	
DviaEstado4	-.041126 (0.0546)	.020925 (0.0587)	-.037357 (0.1515)	
numCuartos_arr	.017406 (0.0111)	.020936 (0.0129)	-.0020945 (0.0615)	.0016248 (0.0251)
pisos_arr	.011789** (0.0049)	.001501 (0.0038)		-.024327 (0.0391)
banios_arr	-.02623 (0.0165)	-.016885 (0.0189)	-.049486 (0.0639)	-.012749 (0.0414)
cuartDormir_arr	.047792*** (0.0133)	.032765** (0.0147)	.1548*** (0.0586)	.062115* (0.0324)
calentador_arr	.0115 (0.0221)	.026481 (0.0242)		
aireLib_arr	-.068423*** (0.0227)	-.065297*** (0.0226)	.070971 (0.1551)	.047757 (0.0728)
ascensor_arr	-.016743 (0.0395)			
numTelFijo_arr	-.042799** (0.0214)	-.073748*** (0.0229)		-.040944 (0.0543)
cocinaExcl_arr	.033198 (0.1152)	.034166 (0.1219)		.26662 (0.2330)

Table A5. *Hedonic Price Function estimation (continued)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
internet_arr	.0012088 (0.0220)	.0031749 (0.0233)		-.056659 (0.0568)
DmatPisos3_arr	.031005 (0.0290)	.018358 (0.0288)	.21272* (0.1199)	.0651 (0.1111)
DmatPisos4_arr	-.011956 (0.0377)	-.021389 (0.0379)	.3904** (0.1566)	-.054408 (0.1285)
DmatPisos5_arr	-.0063366 (0.0438)	-.01774 (0.0450)	.3961* (0.2130)	-.013484 (0.1410)
DmatPisos6_arr	.071817 (0.0709)	.20341** (0.0985)	.043658 (0.3134)	-.05533 (0.1547)
DmatPisos7_arr	-.18871 (0.3485)			-.28053 (0.4626)
fallViv_arr	-.035629* (0.0199)	-.026462 (0.0214)		-.062963 (0.0515)
DmatParedes2_arr	.013932 (0.0418)	.015672 (0.0442)		
DmatParedes3_arr	.0007494 (0.0199)	-.0066011 (0.0212)		
DmatParedes4_arr	.0034505 (0.0747)	.010016 (0.0919)		
DmatParedes5_arr	.14741 (0.2158)	.20993 (0.3420)		
DmatParedes6_arr	-.31457 (0.3102)			
DubiAgua2_arr	-.033876 (0.0561)	-.037619 (0.0623)		.087847 (0.1179)
DubiAgua3_arr	.018871 (0.2123)	.069673 (0.2039)		
DubiAgua4_arr	-.016596 (0.1914)	-.023827 (0.1791)		
DdondCocina2_arr	-.010989 (0.1630)	-.11327 (0.1824)		.045054 (0.3284)
DdondCocina3_arr	-.032122 (0.1293)	-.24595 (0.1762)		-.32627 (0.3467)
DdondCocina4_arr	.056529 (0.3621)			

Table A5. *Hedonic Price Function estimation (continued)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
DdondCocina5_arr	-.16757 (0.3055)	.11238 (0.3047)		
homicidios_arr	.018275* (0.0101)	.0095691 (0.0106)	.15455** (0.0613)	.019083 (0.0268)
inseguridad_arr	-.0075902 (0.0211)	.0076646 (0.0219)	-.27237*** (0.1039)	-.030031 (0.0598)
distFuenCont_arr	2.69e-06 (0.0000)	-.0000196 (0.0000)		.0000316 (0.0001)
puntaje_arr	.0017793 (0.0012)	.000856 (0.0012)	-.0003746 (0.0061)	-.00222 (0.0044)
indComSer_arr	.012978 (0.0198)	.032735 (0.0211)		-.013946 (0.0473)
anunciosExc_arr	-.0097475 (0.0334)	.018952 (0.0357)		.032984 (0.0750)
invasionAndenes_arr	-.041807 (0.0259)	.0006611 (0.0267)		-.0753 (0.0657)
DviaIllumin2_arr	.031968 (0.0215)	.05043** (0.0230)		.076804 (0.0478)
DviaIllumin3_arr	.097818 (0.1539)	.10475 (0.1803)		.20111 (0.2309)
DviaEstado2_arr	.011309 (0.0247)	.0019734 (0.0270)	-.14168 (0.1460)	.095804* (0.0538)
DviaEstado3_arr	-.027709 (0.0435)	-.09546** (0.0486)	.19338 (0.2032)	-.05722 (0.0878)
DviaEstado4_arr	.06769 (0.0829)	.037084 (0.0890)	.21725 (0.3710)	-.072165 (0.1833)
CO	.67184*** (0.1779)	.54141*** (0.2021)	-.96809 (1.0685)	.90066** (0.4120)
SO2	-.050516** (0.0240)	-.066538** (0.0282)	.10485 (0.1333)	-.037957 (0.0484)
O3	.012656** (0.0059)	.017121*** (0.0060)	-.074452* (0.0423)	-.01454 (0.0211)
1.localidad	0 (.)	0 (.)	0 (.)	0 (.)
2.localidad	.085845* (0.0446)	.087635** (0.0407)		-.16339 (0.3603)

Table A5. *Hedonic Price Function estimation (continued)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
3.localidad	-.18233*** (0.0627)	-.22474*** (0.0652)	-.97632*** (0.3539)	-.22744 (0.4052)
4.localidad	-.3062*** (0.0922)	-.3488*** (0.0931)		-.44226 (0.3831)
6.localidad	-.14898* (0.0848)	-.15492** (0.0737)		-.16098 (0.3635)
8.localidad	-.099101 (0.0880)	-.42125*** (0.0804)	-.35 (0.4285)	.0919 (0.3577)
9.localidad	-.0084299 (0.0649)	-.06586 (0.0588)	-.45089 (0.2949)	-.15214 (0.3197)
10.localidad	-.32114*** (0.0584)	-.36906*** (0.0617)	-.67777*** (0.2357)	-.31368 (0.3130)
11.localidad	-.19098*** (0.0449)	-.23917*** (0.0443)	-.23457 (0.1755)	-.16362 (0.3051)
12.localidad	-.13262*** (0.0492)	-.27799*** (0.0508)	-.14947 (0.2148)	.039731 (0.3165)
13.localidad	-.030834 (0.0517)	-.1951*** (0.0566)	.011039 (0.3384)	.43007 (0.3094)
14.localidad	-.17126*** (0.0574)	-.2603*** (0.0561)		-.23845 (0.3440)
15.localidad	-.10404* (0.0584)	-.28902*** (0.0581)	.43844 (0.3711)	-.066819 (0.3399)
16.localidad	-.16758** (0.0725)	-.23037*** (0.0661)	-.45653 (0.3555)	-.26788 (0.3492)
17.localidad	.086969 (0.0842)	.14668* (0.0803)		-.035157 (0.5840)
18.localidad	-.20175*** (0.0642)	-.28428*** (0.0665)	-.58612* (0.3341)	-.24641 (0.3495)
tanqAgua		.031344** (0.0140)	-.15112*** (0.0418)	
conjunCerr		-.027873* (0.0155)		

Table A5. *Hedonic Price Function estimation (continued)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
DcombCocina2		.0092477 (0.0180)	.012336 (0.0697)	
DcombCocina3		-.030222 (0.0360)	.18355 (0.1561)	
DcombCocina4		.26886 (0.3092)		
tanqAgua_arr		-.010668 (0.0206)	.22473*** (0.0866)	
conjunCerr_arr		.046354* (0.0239)		
DcombCocina2_arr		.02818 (0.0277)	.23527 (0.1808)	
DcombCocina3_arr		-.046722 (0.0470)	-.10424 (0.2535)	
DcombCocina4_arr		-.4383 (0.3386)		
lavadero			-.18407* (0.1014)	
lavadero_arr			.14308 (0.1816)	
garaje				.14267*** (0.0292)
garaje_arr				.064829 (0.0605)
Constant	4.6848*** (0.2765)	4.543*** (0.2986)	8.3833*** (1.0803)	5.0767*** (0.7562)

Table A5. *Hedonic Price Function estimation (continued) – Asymmetric Variance Function (Hedonic Frontier Model)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
Constant	-2.3132*** (0.0360)	-2.6197*** (0.0393)	-3.547*** (0.2904)	-2.0072*** (0.0487)
lnsig2u				
PM10	.021559 (0.1867)	.263* (0.1564)	.069413 (0.1155)	-.61459 (1.7330)
1.localidad	0 (.)	0 (.)	0 (.)	0 (.)
2.localidad	.70105 (0.4691)	.28982 (0.3668)		-16.967 (4360.1195)
3.localidad	-1.1231* (0.6530)	-1.5068** (0.6488)	.39925 (1.6642)	3.1378 (12.6043)
4.localidad	-.84964 (0.7670)	-1.4013* (0.7209)		1.3068 (12.7945)
6.localidad	-.19284 (1.1874)	.22377 (0.5730)		-6.5506 (174.4877)
8.localidad	.52588 (1.2311)	-32.757 (5371.3371)	.098763 (2.1778)	4.5069 (12.5208)
9.localidad	1.0777* (0.6385)	.20519 (0.4718)	-.14683 (1.2337)	-18.763 (1003.3409)
10.localidad	-2.0998* (1.0845)	-2.0721*** (0.7640)	-2.5977 (.)	-2.7955 (18.6521)
11.localidad	.087188 (0.4538)	-.60339 (0.4096)	1.432 (0.9503)	2.7346 (12.5811)
12.localidad	.080845 (0.4288)	-1.2473** (0.4876)	-.77322 (0.7453)	4.0724 (12.5257)
13.localidad	1.1687** (0.4690)	-.83868 (0.6101)	.078414 (1.1609)	5.9173 (12.3719)
14.localidad	.35479 (0.5350)	-.81068* (0.4715)		2.2339 (12.7364)
15.localidad	.13053 (0.4818)	-1.4713*** (0.4846)	2.5441** (1.2911)	3.8431 (12.5336)
16.localidad	1.0721* (0.6475)	-.18329 (0.5453)	.22243 (1.5290)	2.0502 (12.7637)
17.localidad	.33337 (0.5272)	.093245 (0.4568)		4.201 (12.6907)
18.localidad	.79989 (0.5381)	-.36995 (0.4792)	-.45648 (1.7249)	3.6518 (12.5565)

Table A5. *Hedonic Price Function estimation (continued) – Asymmetric Variance Function (Hedonic Frontier Model)*

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
2.estrFinal#c.inter2m	-.15481	-.36473**		.5513
eanPM	(0.1890)	(0.1566)		(1.7393)
3.estrFinal#c.inter2m	-.18605	-.31125**	-.12485*	.46649
eanPM	(0.1877)	(0.1546)	(0.0734)	(1.7386)
4.estrFinal#c.inter2m	-.18186	-.33191**	-.19171	.51779
eanPM	(0.1867)	(0.1540)	(0.1301)	(1.7371)
5.estrFinal#c.inter2m	-.25185***	-.31439***	-2.2839	
eanPM	(0.0294)	(0.0236)	(1311.7614)	
1.estrFinal	0	0		0
	(.)	(.)		(.)
2.estrFinal	4.755	14.54**	0	-27.987
	(8.2416)	(6.8384)	(.)	(78.8039)
3.estrFinal	4.9464	10.855	6.5396	-24.943
	(8.1508)	(6.7280)	(4.3028)	(78.6711)
4.estrFinal	3.8944	11.003	9.5183	-27.502
	(8.1180)	(6.6990)	(6.4596)	(78.5452)
5.estrFinal	5.7356	9.3095	75.945	-7.1616
	(.)	(.)	(54882.7625)	(15.4630)
6.estrFinal	-32.943	-30.686	-9.1389	
	(618.7509)	(418.2796)	(8.2966)	
Constant	-.1478	-10.399	-6.1786	25.702
	(8.1466)	(6.7895)	(6.0575)	(69.8236)
R ²				
N	6,061	4,153	319	1,562
Est. Var.	oim	oim	oim	oim

Standard errors in parentheses

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$

Table A6. Demand for PM10 reduction estimation by 2SLS

	(1) First Stage PM10	(2) Second Stage
PM10		-.057522*** (0.0133)
Rainfall	-.0057113** (0.0026)	
Wind speed	19.996*** (5.9213)	
Wind speed ²	-9.3602*** (3.1687)	
Wind speed ³	1.4206*** (0.5423)	
WPM10	1.0011*** (0.0176)	
Income	-5.03e-06 (0.0000)	.0001854*** (0.0001)
educMasAlta	.0039801 (0.0099)	.13915*** (0.0512)
sexoJH	.0072044 (0.0294)	.16051 (0.1262)
edadJH	.0008034 (0.0013)	.030495*** (0.0064)
tiemBogJH	.000333 (0.0008)	-.0009991 (0.0047)
hog_hijos	-.0041393 (0.0144)	-.18256* (0.0987)
hog_orgAmb	.036098 (0.0772)	.19915 (0.4925)
menores_5	-.0045431 (0.0283)	-.34139** (0.1723)
men5Tos	.0048016 (0.0387)	-.057857 (0.2458)
men5DifRes	-.0053862 (0.0552)	.27923 (0.4358)

Table A6. Demand for PM10 reduction estimation by 2SLS (continued)

	(1) First Stage PM10	(2) Second Stage
perHog	-.0053309 (0.0103)	.19829** (0.0805)
arriendo	-.009277 (0.0411)	-.98919*** (0.1280)
numCuartos	.0014222 (0.0188)	1.4776*** (0.0835)
puntaje	-.0040868 (0.0040)	.017732** (0.0083)
aireLib	-.069127* (0.0367)	1.0549*** (0.1435)
homicidios	.037244 (0.0292)	-.20033** (0.0912)
inseguridad	.018274 (0.0435)	.037545 (0.1444)
Constant	-13.31*** (4.3914)	-2.3394*** (0.8696)
R ²		0.279
N	5,694	5,694
Hausman pValue		0.013
Relevance pValue		0.000
Hansen pValue		0.731
Var. Est.	robust cluster	robust cluster

Standard errors in parentheses

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$

Table A7. *Two Step Clustering results*

Cluster	1	2	3
Size	40.7%	12.3%	47.0%
Entries	Apartments 0 (100%) Build Area 160.94 Land Area 104.65 Horizontal property 0 (100%) Year 1984.10	Apartments 0 (100%) Build Area 72.27 Land Area 69.80 Horizontal property 1 (100%) Year 1999.58	Apartments 1 (100%) Build Area 71.31 Land Area 35.96 Horizontal property 1 (100%) Year 1995.59

Notes: Table A7 replicates SPSS output for the Two-step Clustering algorithm. The first row identifies the proposed clusters, and assigns to each of them an id number. The second row presents the size of each cluster, as the percentage that each of them represents in the *CATASTRO* database. Finally, from row three onward, the algorithm shows the variable with which clusters were created. Results suggest that the variables *Apartments* and *Horizontal property* completely define the clusters. Hence the first group is defined by no-condos houses (*Apartments* = 0 and *Horizontal property* = 0), the second group corresponds to houses in condominiums (*Apartments* = 0 and *Horizontal property* = 1), and the third one corresponds to apartments (*Apartments* = 1 and *Horizontal property* = 1).

Table A8. OLS estimation of Hedonic Price Function with spatial fixed effects

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
PM10	-.0046074** (0.0020)	-.0045677* (0.0024)	-.01429* (0.0084)	-.0082987** (0.0036)
inter2meanPM_arr	.0030557** (0.0014)	.0019761 (0.0016)	.0021704 (0.0053)	.0029045 (0.0029)
arriendo	-.32736 (0.4181)	-.013404 (0.5171)	-.61886 (0.4769)	-.67975 (0.4641)
numCuartos	.066669*** (0.0078)	.070988*** (0.0107)	.019992 (0.0410)	.057939*** (0.0122)
pisos	-.012212*** (0.0042)	.0057968* (0.0033)		-.0068527 (0.0259)
banios	.18049*** (0.0124)	.18344*** (0.0147)	.20956*** (0.0515)	.11915*** (0.0203)
cuartDormir	.0029746 (0.0081)	-.0030892 (0.0097)	.028503 (0.0310)	-.0064286 (0.0155)
calentador	.049251*** (0.0176)	.018587 (0.0207)		.089458*** (0.0270)
aireLib	.061304*** (0.0179)	.033993* (0.0195)	.10334* (0.0597)	.074419* (0.0448)
ascensor	.11889*** (0.0310)			
numTelFijo	.049584*** (0.0187)	.067907*** (0.0217)		.039872 (0.0366)
cocinaExcl	.302 (0.1893)	.31327 (0.2293)		.12621 (0.1174)
internet	.072865*** (0.0146)	.067502*** (0.0173)		.10401*** (0.0293)
DmatPisos3	-.033537* (0.0190)	-.036487* (0.0203)	-.081618 (0.0621)	-.020914 (0.0490)
DmatPisos4	-.035186 (0.0255)	-.019081 (0.0289)	-.37962*** (0.1236)	-.01592 (0.0598)
DmatPisos5	-.0087276 (0.0297)	-.0005532 (0.0337)	-.10635 (0.0759)	-.040926 (0.0703)
DmatPisos6	-.20627*** (0.0658)	-.35215*** (0.0852)	-.10763 (0.1120)	-.11745 (0.1029)
DmatPisos7	-.17144 (0.3180)	.32722*** (0.0697)		-.13321 (0.2698)

Table A8. OLS estimation of Hedonic Price Function with spatial fixed effects (continued)

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
fallViv	-.017488 (0.0148)	-.020818 (0.0172)		-.017115 (0.0296)
DmatParedes2	-.023076 (0.0355)	-.032813 (0.0376)		.0029843 (0.0545)
DmatParedes3	-.0028973 (0.0155)	-.014343 (0.0182)		.038275 (0.0264)
DmatParedes4	-.023559 (0.0559)	.02188 (0.0754)		-.032552 (0.0634)
DmatParedes5	-.46975*** (0.1161)	-.62146*** (0.1681)		-.344** (0.1461)
DmatParedes6	.10008 (0.2358)	-.0025441 (0.0585)		.069315 (0.2037)
DmatParedes7	-.81796*** (0.2327)			
DubiAgua2	-.013633 (0.0439)	-.017833 (0.0548)	.11002 (0.1009)	-.029702 (0.0919)
DubiAgua3	.045924 (0.0443)	.038111 (0.0526)		.022608 (0.0744)
DubiAgua4	-.10434 (0.1425)	-.09405 (0.1338)		-.112 (0.0866)
DdondCocina2	-.058544 (0.1675)	.17759 (0.1412)		-.28539 (0.2452)
DdondCocina3	-.15588 (0.1873)	.25748 (0.2688)		.059556 (0.3425)
DdondCocina4	-.11323 (0.1509)	.017653 (0.0688)		-.26873** (0.1310)
DdondCocina5	.01456 (0.0600)	.06901 (0.0563)		-.22682 (0.2050)
DdondCocina6	.0071264 (0.7128)			-.32741 (0.9404)
homicidios	-.018552** (0.0085)	-.012564 (0.0081)	-.10655** (0.0507)	-.016679 (0.0235)
inseguridad	-.013314 (0.0151)	-.027869 (0.0178)	.13612** (0.0575)	-.018984 (0.0336)
distFuenCont	-5.34e-06 (0.0000)	.0000101 (0.0000)		-.0000963* (0.0001)

Table A8. OLS estimation of Hedonic Price Function with spatial fixed effects (continued)

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
puntaje	.015034*** (0.0012)	.015435*** (0.0013)	.013798*** (0.0033)	.021132*** (0.0023)
indComSer	-.0041321 (0.0158)	-.0043714 (0.0180)		.0063251 (0.0317)
anunciosExc	-.0075159 (0.0237)	-.035551 (0.0244)		.0021211 (0.0481)
invasionAndenes	.022417 (0.0184)	.0031325 (0.0193)		.061316 (0.0413)
DviaIllumin2	-.0013441 (0.0159)	-.024637 (0.0186)		.037351 (0.0294)
DviaIllumin3	.0050059 (0.1137)	.026662 (0.1278)		.089711 (0.2044)
DviaEstado2	-.021227 (0.0187)	-.031037 (0.0212)	.050405 (0.0821)	-.01417 (0.0322)
DviaEstado3	-.040266 (0.0348)	.0053685 (0.0390)	-.24811* (0.1247)	-.079389 (0.0617)
DviaEstado4	-.048171 (0.0583)	.0060945 (0.0659)	-.10789 (0.1281)	-.10222 (0.1239)
numCuartos_arr	.018082 (0.0126)	.02054 (0.0146)	-.010019 (0.0532)	-.0066991 (0.0253)
pisos_arr	.011778*** (0.0043)	.0022599 (0.0038)		-.021216 (0.0372)
banios_arr	-.02805 (0.0187)	-.032711 (0.0214)	-.086961 (0.0674)	.0088846 (0.0393)
cuartDormir_arr	.049644*** (0.0141)	.039296** (0.0165)	.12211** (0.0462)	.074932** (0.0297)
calentador_arr	.0082057 (0.0245)	.018317 (0.0278)		
aireLib_arr	-.060264** (0.0240)	-.053465** (0.0256)	.06119 (0.1130)	-.027039 (0.0605)
ascensor_arr	-.0059993 (0.0366)			
numTelFijo_arr	-.051335** (0.0250)	-.067801** (0.0267)		-.09695 (0.0623)
cocinaExcl_arr	-.054572 (0.1994)	-.13317 (0.2464)		.22116 (0.1859)

Table A8. OLS estimation of Hedonic Price Function with spatial fixed effects (continued)

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
internet_arr	-.0066045 (0.0229)	-.0041574 (0.0264)		-.056454 (0.0529)
DmatPisos3_arr	.025568 (0.0278)	.0092123 (0.0293)	.17044* (0.0938)	.082256 (0.1165)
DmatPisos4_arr	-.014361 (0.0381)	-.024708 (0.0429)	.43798*** (0.1564)	-.063362 (0.1272)
DmatPisos5_arr	-.008578 (0.0416)	-.017037 (0.0468)	.33499*** (0.1194)	.01017 (0.1334)
DmatPisos6_arr	.065471 (0.0858)	.21089* (0.1177)	-.093935 (0.2144)	-.059241 (0.1584)
DmatPisos7_arr	-.14444 (0.3298)			-.4349 (0.3620)
fallViv_arr	-.034185* (0.0201)	-.024431 (0.0228)		-.05428 (0.0464)
DmatParedes2_arr	.010377 (0.0482)	.004583 (0.0467)		
DmatParedes3_arr	.0040624 (0.0218)	-.0051925 (0.0246)		
DmatParedes4_arr	.032694 (0.0800)	.01321 (0.0974)		
DmatParedes5_arr	.19028 (0.1165)	.25136 (0.2107)		
DmatParedes6_arr	-.30028 (0.2595)			
DubiAgua2_arr	-.049697 (0.0632)	-.072379 (0.0726)		.08555 (0.1449)
DubiAgua3_arr	-.0026143 (0.1454)	.0064831 (0.1707)		
DubiAgua4_arr	.014592 (0.1593)	-.018885 (0.1613)		
DdondCocina2_arr	.05488 (0.1862)	-.11733 (0.1716)		.16549 (0.2506)
DdondCocina3_arr	.0077964 (0.2181)	-.30549 (0.2989)		-.25552 (0.3585)
DdondCocina4_arr	.1391 (0.1763)			

Table A8. OLS estimation of Hedonic Price Function with spatial fixed effects (continued)

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
DdondCocina5_arr	-.23676 (0.1970)	.071583 (0.1189)		
DdondCocina6_arr	.58991 (0.7588)			1.3917 (0.9915)
homicidios_arr	.015581* (0.0090)	.0069284 (0.0091)	.11972* (0.0659)	.012195 (0.0302)
inseguridad_arr	-.0098122 (0.0209)	.010179 (0.0237)	-.36481*** (0.0989)	-.0067433 (0.0536)
distFuenCont_arr	5.47e-06 (0.0000)	-.0000268 (0.0000)		.0000819 (0.0001)
puntaje_arr	.0011955 (0.0013)	.0007431 (0.0014)	.0002766 (0.0060)	-.0066609 (0.0042)
indComSer_arr	.0090264 (0.0214)	.025042 (0.0246)		-.016384 (0.0507)
anunciosExc_arr	-.0094896 (0.0330)	.0089116 (0.0376)		.038112 (0.0710)
invasionAndenes_arr	-.038893 (0.0270)	-.0064112 (0.0274)		-.14766* (0.0771)
DviaIllumin2_arr	.040736* (0.0219)	.055063** (0.0254)		.044555 (0.0478)
DviaIllumin3_arr	.071015 (0.1713)	.046867 (0.1885)		.1512 (0.2122)
DviaEstado2_arr	.016982 (0.0247)	.0081085 (0.0281)	-.14332 (0.1169)	.12474** (0.0574)
DviaEstado3_arr	-.013112 (0.0473)	-.073962 (0.0574)	.16996 (0.1840)	.045214 (0.0884)
DviaEstado4_arr	.061208 (0.0831)	.025753 (0.0913)	-.017154 (0.2709)	.17438 (0.1919)
1.localidad	0 (.)	0 (.)	0 (.)	0 (.)
2.localidad	.065296 (0.0473)	.07192 (0.0514)		-.17687 (0.2469)
3.localidad	-.15691*** (0.0573)	-.15608** (0.0613)	-.22579 (0.1953)	-.3393* (0.1946)
4.localidad	-.35096*** (0.0783)	-.34846*** (0.0971)		-.40483** (0.1574)

Table A8. OLS estimation of Hedonic Price Function with spatial fixed effects (continued)

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
6.localidad	-.21581*** (0.0680)	-.25234*** (0.0771)		-.15411 (0.1559)
8.localidad	-.21036*** (0.0736)	-.2558*** (0.0834)	.010669 (0.2622)	-.12283 (0.1712)
9.localidad	-.15619*** (0.0505)	-.15377*** (0.0561)	.12204 (0.1689)	-.23104 (0.1513)
10.localidad	-.21163*** (0.0481)	-.23024*** (0.0537)	-.18333 (0.1735)	-.25463** (0.1246)
11.localidad	-.2415*** (0.0473)	-.25436*** (0.0526)	-.16653 (0.1652)	-.27757** (0.1296)
12.localidad	-.14998*** (0.0477)	-.18682*** (0.0498)	.085702 (0.1603)	-.21062* (0.1277)
13.localidad	-.098824** (0.0445)	-.11888** (0.0483)	.36625** (0.1655)	-.13996 (0.1337)
14.localidad	-.2077*** (0.0515)	-.228*** (0.0595)		-.23207* (0.1269)
15.localidad	-.17202*** (0.0510)	-.25237*** (0.0557)	.28988 (0.1948)	-.22146* (0.1274)
16.localidad	-.21451*** (0.0547)	-.25573*** (0.0612)	-.048224 (0.2020)	-.2276* (0.1347)
17.localidad	-.10328 (0.0810)	-.018197 (0.0890)		-.29488 (0.2926)
18.localidad	-.34332*** (0.0560)	-.35432*** (0.0623)	.0084572 (0.2041)	-.3958*** (0.1372)
tanqAgua		.04161** (0.0169)	-.14652*** (0.0553)	
conjunCerr		-.037806* (0.0202)		
DcombCocina2		.012703 (0.0223)	.078877 (0.1291)	
DcombCocina3		-.013896 (0.0431)	.24407 (0.1555)	
DcombCocina4		.37804*** (0.0517)		
tanqAgua_arr		-.018931 (0.0226)	.21515*** (0.0754)	

Table A8. OLS estimation of Hedonic Price Function with spatial fixed effects (continued)

	(1) All	(2) Apart.	(3) House in Condos	(4) House no Condos
conjuncerr_arr		.034201 (0.0257)		
DcombCocina2_arr		.021562 (0.0306)	.13669 (0.1649)	
DcombCocina3_arr		-.0707 (0.0523)	-.13712 (0.1806)	
DcombCocina4_arr		-.56338*** (0.1249)		
lavadero			-.1094 (0.0918)	
lavadero_arr			.11454 (0.1059)	
garaje_arr				.19606*** (0.0437)
Constant	4.2851*** (0.4039)	4.2016*** (0.4867)	5.4872*** (0.4977)	4.7775*** (0.3469)
R ²	.67091	.7262	.73803	.56368
N	6,061	4,153	319	1,562
Est. Var.	cluster	cluster	cluster	cluster

Standard errors in parentheses

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$

Table A9. *Hedonic Price Function estimation with different functional form and different air quality variables*

	Mean daily PM10 interpolated by VNA		Mean daily PM10 interpolated by IDW		Mean Dailly Maximum	
	Log-Lin	Log-Log	Log-Lin	Log-Log	Log-Lin	Log-Log
All	-0.01537 (0.00000)	-0.86020 (0.00000)	-0.00914 (0.00189)	-0.41898 (0.00681)	-0.00303 (0.00021)	-0.72675 (0.00033)
Apart	-0.00612 (0.03449)	-0.30677 (0.04848)	-0.00187 (0.51649)	-0.04345 (0.76441)	-0.00081 (0.39885)	-0.11974 (0.60767)
House in Condos	-0.02764 (0.25870)	0.49005 (0.47102)	-0.05101 (0.00206)	0.38448 (0.57686)	-0.00830 (0.10124)	0.75956 (0.48149)
House no Condos	-0.02165 (0.00051)	-0.48458 (0.42851)	-0.02299 (0.00319)	-0.67507 (0.28500)	-0.00812 (0.00007)	-1.03388 (0.29131)

P-value of individual significance test in parentheses

Table A10. *Demand for Air Quality Estimation using Log-Log Function*

	(1) First Stage	(2) Second Stage
log(PM10)		-.5232*** (0.1296)
log(Rainfall)	-.0012634 (0.0021)	
log(wind speed)	.006998** (0.0034)	
log(WPM10)	1.009*** (0.0057)	
Income	-1.75e-07 (0.0000)	.0000274*** (0.0000)
R ²		0.400
N	5,693	5,693
Hausman pValue		0.189
Relevance pValue		0.000
Hansen pValue		0.177
Var. Est.	robust cluster	robust cluster

Standard errors in parentheses

** $p < 0.05$, *** $p < 0.01$ Table A11. *Demand for Air Quality Estimation using IDW interpolation*

	(1) First Stage	(2) Second Stage
meanStock		-.074849** (0.0295)
Rainfall	.010088** (0.0041)	
Wind speed	48.459*** (13.6817)	
Wind speed ²	-21.376*** (6.9379)	
Wind speed ³	3.0544*** (1.1381)	
WPM10	.94512*** (0.0219)	
Income	-2.85e-06 (0.0000)	.0001252 (0.0001)
R ²		0.177
N	5,694	5,694
Hausman pValor		0.153
Relevancia pValor		0.000
Hansen pValor		0.327
Var. Est.	robust cluster	robust cluster

Standard errors in parentheses

** $p < 0.05$, *** $p < 0.01$

Table A12. *Demand for Air Quality Estimation using like instruments Wind Speed, Wind Speed² and Wind Speed³ and Winter mean of PM10*

	(1) First Stage	(2) Second Stage
PM10		-.057278*** (0.0132)
Wind speed	16.535** (6.6133)	
Wind speed ²	-7.9354** (3.4826)	
Wind speed ³	1.2422** (0.5852)	
WPM10	1.0106*** (0.0146)	
Incone	-5.36e-06 (0.0000)	.0001854*** (0.0001)
R ²		0.279
N	5,694	5,694
Hausman pValue		0.016
Relevance pValue		0.000
Hansen pValue		0.708
Var. Est.	robust cluster	robust cluster

Standard errors in parentheses

** $p < 0.05$, *** $p < 0.01$

Table A13. *Demand for Air Quality Estimation using like instruments Wind Speed, Wind Speed² and Wind Speed³*

	(1) First Stage	(2) Second Stage
PM10		-.072883** (0.0310)
Wind speed	-442.35*** (38.0815)	
Wind speed ²	235.76*** (19.6906)	
Wind speed ³	-39.694*** (3.3036)	
Incone	.0000392 (0.0000)	.0001849*** (0.0001)
R ²		0.279
N	5,694	5,694
Hausman pValue		0.451
Relevance pValue		0.000
Hansen pValue		0.603
Var. Est.	robust cluster	robust cluster

Standard errors in parentheses

** $p < 0.05$, *** $p < 0.01$

Table A14. *Demand for Air Quality Estimation using FS estimation with localities*

	(1) First Stage	(2) Second Stage
PM10		-.025757*** (0.0054)
Rainfall	-.0057113** (0.0026)	
Wind speed	19.996*** (5.9213)	
Wind speed ²	-9.3602*** (3.1687)	
Wind speed ³	1.4206*** (0.5423)	
WPM10	1.0011*** (0.0176)	
Income	-5.03e-06 (0.0000)	.0001404*** (0.0000)
R ²		0.345
N	5,694	5,694
Hausman pValue		0.030
Relevance pValue		0.000
Hansen pValue		0.305
Var. Est.	robust cluster	robust cluster

Standard errors in parentheses

** $p < 0.05$, *** $p < 0.01$

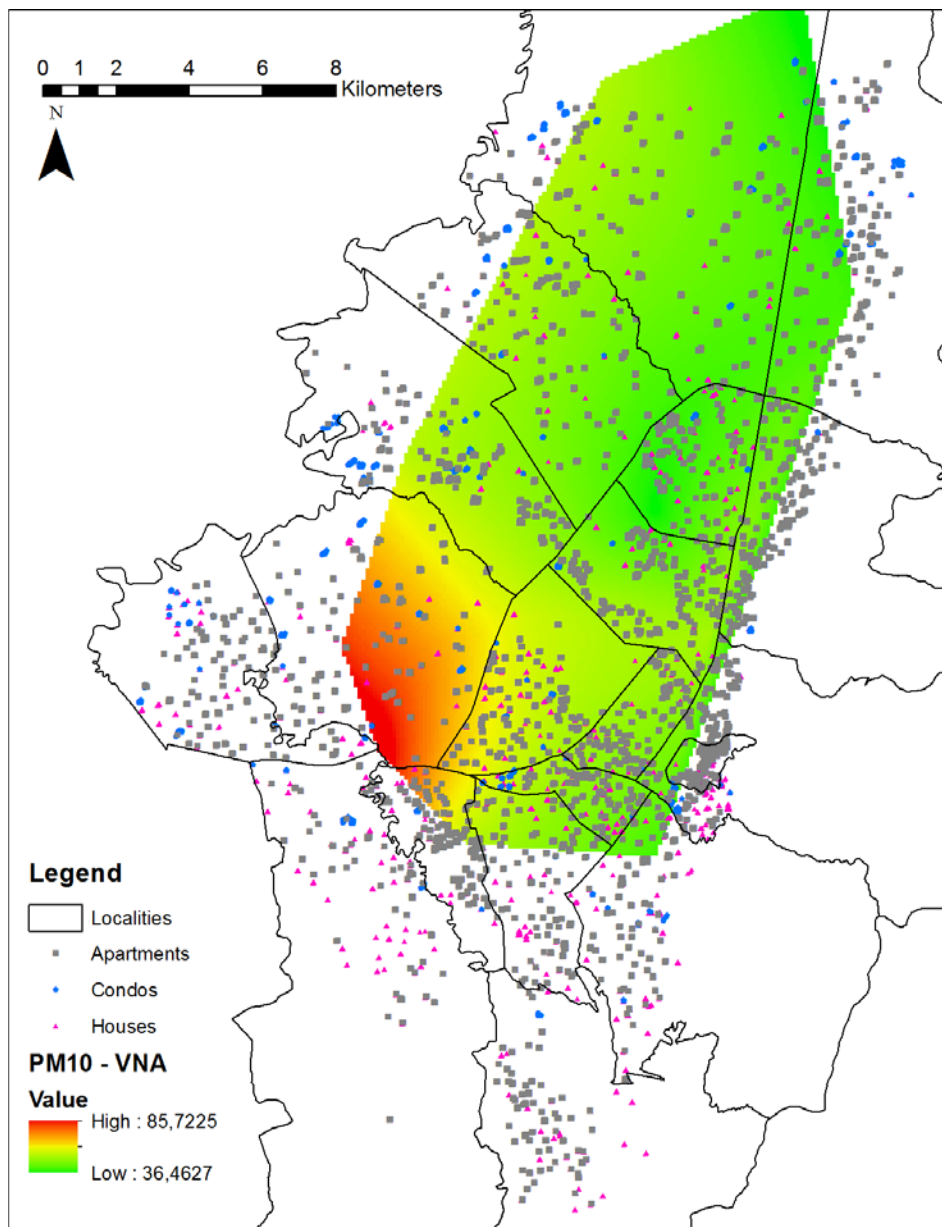


Figure A1. Types of dwellings

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