Do Airbnb properties affect house prices?

Stephen Sheppard and Andrew Udell

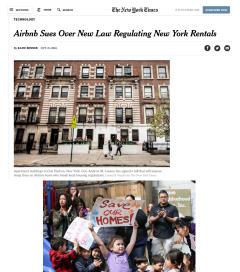
Williams College / Dropbox, Inc.



ASSA/American Real Estate and Urban Economics Association Meetings
Philadelphia, PA

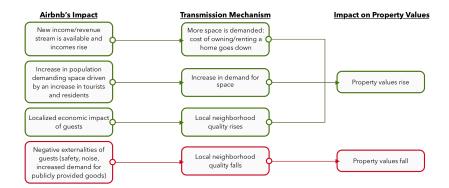
January 5, 2018

POLICY CONCERN IN RESPONSE TO AIRBNB GROWTH

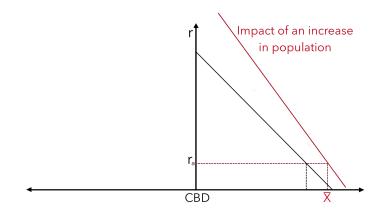


- 'Prohibited' in Barcelona, Berlin & Santa Monica
- Recent limits in NYC, Airbnb sues
- Controversial elsewhere
- Competition with hotels
- Short vs long term?
- Housing prices?
- Gentrification?

TRANSMISSION MECHANISMS

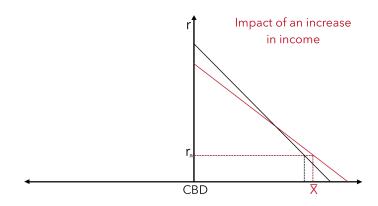


ACCOMMODATING MORE POPULATION



Equilibrium utility declines

INCREASING INCOME



Equilibrium utility increases

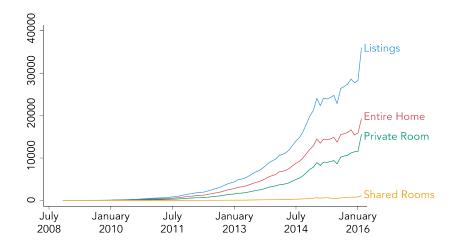
DATA SOURCES

Source	Description & Use
InsideAirbnb	InsideAirbnb (released by Murray Cox) contains information such as
	pricing, reviews, and location of each listing on Airbnb that was available
	on the date the Airbnb website was crawled (12 times in 2015-2016).
Department of Finance	The Department of Finance releases information on all sales in New York
Annualized Sales Data: January	City. The data are available from 2003-2015 and contain information such
2003 - August 2015	as sale price, square footage, and sale date.
Department of Finance "Places"	The Department of Finance releases information on areas of interest, such
or "Areas-of-Interest" Map	as parks, cemeteries, and airports, available in GIS format.
Department of City Planning	The Department of City Planning releases detailed information about each
Pluto TM	tax lot in New York City (of specific use for this analysis was square
	footage information).
American Community Survey	The American Community Survey contains information available at the
2010-2014	Census Tract level such as education, racial and ethnic demographics, and
	employment-related measures.
New York Police Department	The New York Police Department reports annual counts for different
Crime Statistics	crimes (major felonies, non-major felonies, and misdemeanors) by precinct.
Census Geography Maps	In order to merge sales with local Census demographics, Census
	geographies needed to be identified and spatially joined to the sales dataset.
Metropolitan Transportation	Information provided by the Metropolitan Transportation Authority was
Authority Map of Subway	made into a map of subway entrances in New York City.
Entrances	

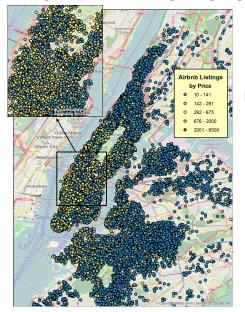
EXTENDING TIMELINE BASED ON RECENT SNAPSHOT



NEW YORK AIRBNB LISTINGS OVER TIME



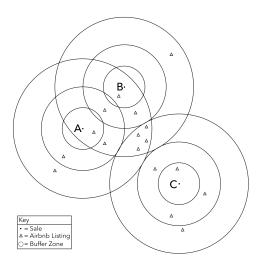
NEW YORK AIRBNB LISTING LOCATIONS



Extensive coverage:

- Manhattan
- Brooklyn
- Queens
 - Astoria
 - Long Island City
 - Jackson Heights

DISTANCE BUFFERS



DESCRIPTIVE STATISTICS

VARIABLES	μ	σ	max
Listing Counts, by Total and Type			
Listings Count (150m)	1.221	5.217	133
Listings Count (300m)	4.644	19.06	439
Listings Count (500m)	11.99	47.75	1,034
Listings Count (1000m)	40.99	157.5	2,899
Listings Count (2000m)	133.4	490.3	6,170
Listing Price			
Avg. Nightly Price (150m)	23.09	65.18	5,000
Avg. Nightly Price (300m)	29.34	69.00	5,000
Sales Unit			
Sale Price	683,922	913,580	10,000,000
Square Footage of Unit	1,183	577.0	18,590

HOW MUCH IMPACT SHOULD WE EXPECT?

$$P = \frac{R(v,x)}{u}$$

$$u = r_{\rm rf} + \omega - \tau \cdot (r_{\rm m} + \omega) + \delta - g + \gamma - (1-\tau) \cdot \alpha$$

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Assume the expected capital gain equals the owner risk premium $\Rightarrow g = \gamma = 0$ and that:

Variable	Value	Interpretation
$r_{\sf rf}$	0.02	Risk free annual interest rate
ω	0.025	Property tax rate as a percent of market price
au	0.29	Effective tax rate on personal income
r_{m}	0.04	Annual mortgage interest rate
δ	0.025	Maintenance costs as a percent of market price
α	0.01373	Airbnb rental as a percent of market price

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 \Rightarrow 17.7% **increase** in the price of housing

Hedonic estimates: impact on $ln({\it sale price})$

Variable	(1)	(2)	(3)
Total Accommodations Total Rooms	0.0646***	0.0814***	
Total Rents			0.0323***

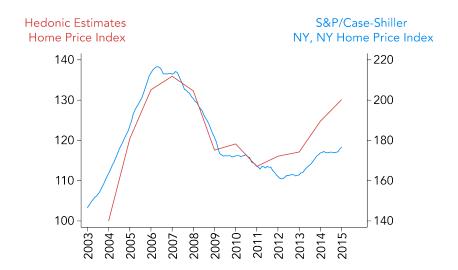
Hedonic estimates: impact on $ln({\tt SALE\ PRICE})$

Variable	(1)	(2)	(3)
Total Accommodations Total Rooms	0.0646***	0.0814***	0.0000***
Total Rents			0.0323***
Square Feet	0.402***	0.402***	0.402***
Felonies	-0.0458***	-0.0455***	-0.0517***
Pre-war	0.0843***	0.0842***	0.0847***
Distance to AOI	-0.103	-0.103	-0.102
Distance to subway	-0.00875	-0.00897	-0.00897
Elevator	0.0858***	0.0863***	0.0847***

Hedonic estimates: impact on $ln({\it sale price})$

Variable	(1)	(2)	(3)
Total Accommodations Total Rooms Total Rents	0.0646***	0.0814***	0.0323***
Square Feet Felonies Pre-war Distance to AOI Distance to subway Elevator	0.402*** -0.0458*** 0.0843*** -0.103 -0.00875 0.0858***	0.402*** -0.0455*** 0.0842*** -0.103 -0.00897 0.0863***	0.402*** -0.0517*** 0.0847*** -0.102 -0.00897 0.0847***
R-squared Sale-Year FE Local Nbd FE Clustered Standard Errors	0.524 YES Block-Group Tract	0.524 YES Block-Group Tract	0.524 YES Block-Group Tract

ESTIMATED HPI AND CASE-SHILLER



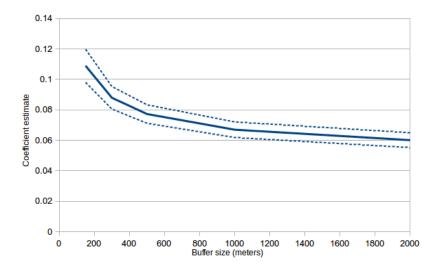
HEDONIC ESTIMATES WITH INCREASING BUFFERS

Variable	(1)	(2)	(3)
Airbnb ₅₀₀	0.0773***		
$Airbnb_{1000}$		0.0670***	
Airbnbann			0.0601***

HEDONIC ESTIMATES WITH INCREASING BUFFERS

Variable	(1)	(2)	(3)
$\begin{array}{l} Airbnb_{500} \\ Airbnb_{1000} \\ Airbnb_{2000} \end{array}$	0.0773***	0.0670***	0.0601***
Sauare Feet Felonies Pre-war Distance to AOI Distance to subway Elevator	0.403*** -0.0268* 0.0839*** -0.100 -0.00984 0.0907***	0.402*** -0.0243 0.0842*** -0.0997 -0.0111 0.0903***	0.402*** -0.0307* 0.0840*** -0.0991 -0.0113 0.0902***
R^2 Local Nbd FE Clustered σ	0.525 Block Broup Tract	0.525 Block Broup Tract	0.525 Block Broup Tract

IMPACT OF INCREASING BUFFER SIZE



EFFECT OF AIRBNB TREATMENT ON ln PRICE

	Nearest	Propensity	Regression
	Neighbor	Score	Adjustment
All distances	0.2094***	0.3171***	0.3493***
Less than 7 km	0.3252***	0.5142***	0.5490***
7 km to 11.5 km	0.2405***	0.3113***	0.2854***
11.5 km to 17 km	0.1734***	0.2774***	0.1678***
More than 17 km	0.1051***	0.0490***	0.0428***

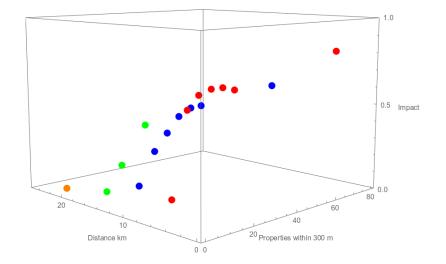
*** - significant at 1 percent

INCREASING INTENSITY OF AIRBNB TREATMENT

# Airbnb	All	Less than	7km to	11.5km to
	Distances	7km	11.5km	17km
1 to 5	0.1596***	0.1413***	0.1653***	0.0855***
6 to 10	0.4553***	0.5838***	0.3371***	0.2202***
11 to 15	0.5399***	0.6486***	0.4228***	-0.1171
16 to 20	0.6636***	0.6712***		0.4179*
21 to 25	0.6403***	0.6699***	0.5361***	-0.0761
26 to 30	0.6436***	0.6499***		-0.3063
31 or more	0.7748***	0.7934***	0.5903***	-0.0325

^{*** -} significant at 1 percent, * - significant at 10 percent

INCREASING INTENSITY OF AIRBNB TREATMENT



CONCLUDING REMARKS

- Airbnb listings do affect local house prices
- Greater in more central locations, less at periphery
- Analogous to increase in population
- Local impact between 5% and 30%
- Similar to expected from simple capitalization
- Similar to that found in Barron, Kung and Proserpio [2017]

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- Airbnb listings do affect local house prices
- Greater in more central locations, less at periphery
- Analogous to increase in population
- Local impact between 5% and 30%
- Similar to expected from simple capitalization
- Similar to that found in Barron, Kung and Proserpio [2017]
- Welfare impacts are less clear
- Presumably positive for Airbnb hosts and guests
- Potential reduced utility for original residents
- Political economy of policy responses