Predicting the Price of AirBnb in New York

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Overview

Company Background:

Airbnb is a community-based, two-sided online platform that facilitates the process of booking private living spaces for travelers.

Website is a competitive marketplace, and the onus of pricing is left to the lister

Client Problem:

 Our client is AirBnb. Our goal is to help AirBnb give their customers recommendation on pricing.

Our Role:

 Our team built a model using data from the AirBnb website to predict the price of listings

Data

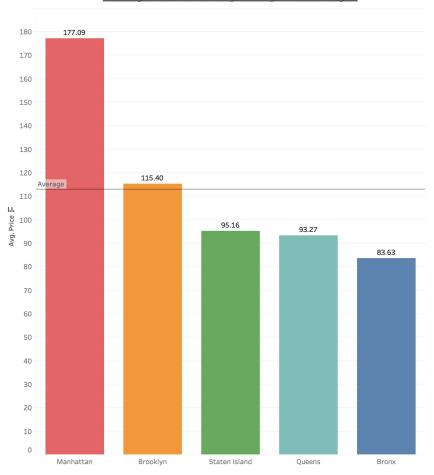
Our dataset had 17 columns not including price. We believe that there are three broad categories that would drive the price of AirBnb listings

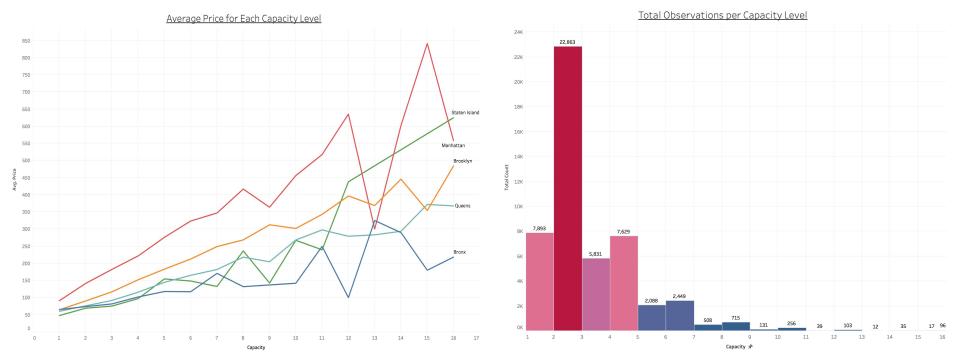
	Туре	Potential Variables
Location	Qualitative	Boroughs Neighborhoods
Type of Listing	Qualitative	Property Type Room Type
Capacity	Quantitative	Beds Bedrooms Accomodations
Bathrooms	Quantitative	Bathrooms



1. Boroughs and Price

Average Price of Listings along NYC Boroughs





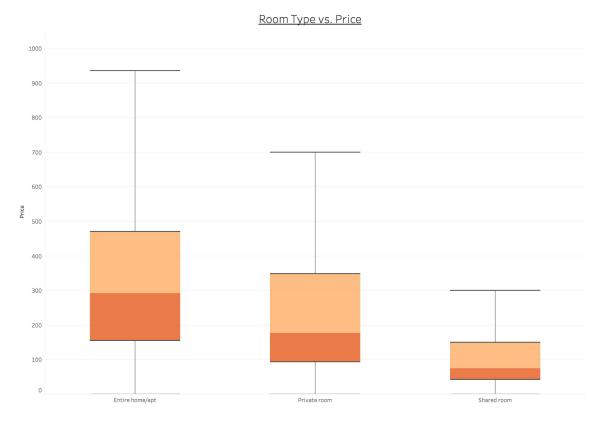
2. Capacity and Price

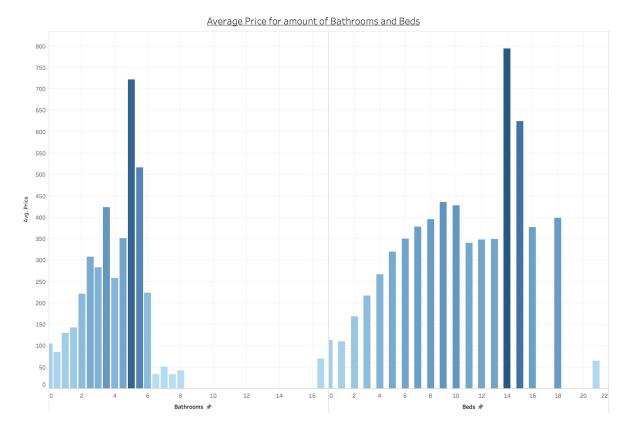
3. Capacity and Price

Entire Home/Apartment

Private Room

Shared Room





4. Bathrooms

Mostly positive correlation

Did not include Beds because of multicollinearity

Model - QDA

	Predictions					
Actual observations		\$0 - \$69	\$69 - \$106	\$106 - \$175	\$175 +	
	\$0 - \$69	803	1538	109	72	
	\$69 - \$106	349	1542	333	345	
	\$106 - \$175	81	685	537	1145	
	\$175 +	38	219	239	2101	

Table 1. Confusion matrix

QDA Evaluation

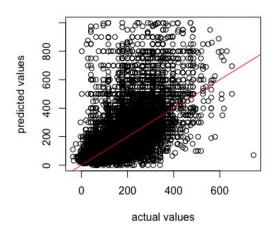
	\$0 - \$69	\$69 - \$106	\$106 - \$175	\$175 +	
sensitivity	0.3183981	0.6002336	0.2193627	0.8090104	
specificity	0.5489887	0.4547377	0.5783039	0.3822788	
accuracy	0.491614				

Table 2. Evaluation based on sensitivity, specificity, and accuracy.

Model - Linear Regression

Model Formula: Price ~ Accommodates + Bathrooms + Room_Type + Boroughs

Visualization



Model Statistics

Residuals:

Min 1Q Median 3Q Max -676.46 -43.23 -9.75 23.28 933.38

Coefficients:

	Estimate	Std. Error	t value	Pr(>ItI)	
(Intercept)	15.2704	2.9473	5.181	2.21e-07	***
accommodates	21.8186	0.2520	86.595	< 2e-16	***
bathrooms	44.9478	0.9597	46.833	< 2e-16	***
neighbourhood_group_cleansedBrooklyn	18.7340	2.7170	6.895	5.45e-12	***
neighbourhood_group_cleansedManhattan	72.8619	2.7167	26.820	< 2e-16	***
neighbourhood_group_cleansedQueens	6.0630	2.8810	2.104	0.0353	*
neighbourhood_group_cleansedStaten Island	-10.5061	5.3256	-1.973	0.0485	*
room_typePrivate room	-60.1517	0.8959	-67.138	< 2e-16	***
room_typeShared room	-80.5925	2.5859	-31.166	< 2e-16	***

Residual standard error: 83.88 on 50656 degrees of freedom Multiple R-squared: 0.4495, Adjusted R-squared: 0.4494 F-statistic: 5170 on 8 and 50656 DF, p-value: < 2.2e-16

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1

Final Product

For an AirBnB listing, the price should be calculated following this formula:

- Start at base price of \$15
- Add \$22 for every person you can accommodate
- Add \$45 for every bathroom in the listing
- Add \$70 if the listing is located in Manhattan
- Add \$20 if the listing is located in Brooklyn
- Add \$6 if the listing is located in Queens
- Discount \$10 if the listing is located in Staten Island
- If the listing is a private room (not an entire home), discount \$60
- If the listing is a shared room (not an entire home), discount \$80

Validating the Model

AirBnB Pricing

Rules:	Base Price: \$15.00		
Per Person	\$22.00	Queens	\$6.00
Per Bathroom	\$45.00	Staten Island	-\$10.00
Manhattan	\$70.00	Private Room	-\$60.00
Brooklyn	\$20.00	Shared Room	-\$80.00

Property Name	Location	Room Type	Capacity	Bathroom	Model Prediction	Actual Value	Percentage Error
1	Manhattan	Entire Home	4	1	\$218.00	\$225.00	3.11%
2	Bronx	Private Room	2	1	\$44.00	\$42.00	-4.76%
3	Staten Island	Entire Home	8	2	\$271.00	\$200.00	-35.50%
4	Queens	Entire Home	1	1	\$88.00	\$75.00	-17.33%
5	Brooklyn	Entire Home	8	1	\$256.00	\$260.00	1.54%
6	Brooklyn	Private Room	2	1	\$64.00	\$79.00	18.99%
7	Manhattan	Entire Home	2	1	\$174.00	\$185.00	5.95%
8	Bronx	Entire Home	3	1	\$126.00	\$125.00	-0.80%
9	Queens	Entire Home	3	1	\$132.00	\$100.00	-32.00%
10	Staten Island	Entire Home	5	1	\$160.00	\$221.00	27.60%
	1 2 3 4 5 6 7 8	1 Manhattan 2 Bronx 3 Staten Island 4 Queens 5 Brooklyn 6 Brooklyn 7 Manhattan 8 Bronx 9 Queens	1 Manhattan Entire Home 2 Bronx Private Room 3 Staten Island Entire Home 4 Queens Entire Home 5 Brooklyn Entire Home 6 Brooklyn Private Room 7 Manhattan Entire Home 8 Bronx Entire Home 9 Queens Entire Home	1 Manhattan Entire Home 4 2 Bronx Private Room 2 3 Staten Island Entire Home 8 4 Queens Entire Home 1 5 Brooklyn Entire Home 8 6 Brooklyn Private Room 2 7 Manhattan Entire Home 2 8 Bronx Entire Home 3 9 Queens Entire Home 3	1 Manhattan Entire Home 4 1 2 Bronx Private Room 2 1 3 Staten Island Entire Home 8 2 4 Queens Entire Home 1 1 5 Brooklyn Entire Home 8 1 6 Brooklyn Private Room 2 1 7 Manhattan Entire Home 2 1 8 Bronx Entire Home 3 1 9 Queens Entire Home 3 1	1 Manhattan Entire Home 4 1 \$218.00 2 Bronx Private Room 2 1 \$44.00 3 Staten Island Entire Home 8 2 \$271.00 4 Queens Entire Home 1 1 \$88.00 5 Brooklyn Entire Home 8 1 \$256.00 6 Brooklyn Private Room 2 1 \$64.00 7 Manhattan Entire Home 2 1 \$174.00 8 Bronx Entire Home 3 1 \$126.00 9 Queens Entire Home 3 1 \$132.00	1 Manhattan Entire Home 4 1 \$218.00 \$225.00 2 Bronx Private Room 2 1 \$44.00 \$42.00 3 Staten Island Entire Home 8 2 \$271.00 \$200.00 4 Queens Entire Home 1 1 \$88.00 \$75.00 5 Brooklyn Entire Home 8 1 \$256.00 \$260.00 6 Brooklyn Private Room 2 1 \$64.00 \$79.00 7 Manhattan Entire Home 2 1 \$174.00 \$185.00 8 Bronx Entire Home 3 1 \$126.00 \$125.00 9 Queens Entire Home 3 1 \$132.00 \$100.00

Recommendation

- We would recommend that AirBnb adopt our model to recommend estate pricing for their users
- Very low R2 only ~50 of variation in price is explained by our model.
- Validation method showed that most predictions are within 30%
 - Not concerning given context
- Value to consumers would outweigh the error

Limitations

- Can only use the model for New York
- Income in the area or proximity to tourist destinations major confounding variable, especially in New York
- Seasonality, customer sentiment, related costs, etc.

Next steps

- 1. Evaluate the model in the field
- 2. Consider and gather new variables (regional income, parking)
- 3. Evaluate seasonality in the model
- 4. Evaluate data progression through time for more accurate results if there is knowledge of Time Series
- 5. Extrapolate the model to other cities

Questions?

