# Predicting Craigslist NYC Apartment Rental Prices with Linear Regression

Ami Tian

## Project Goal: Do an apartment's features help predict its rental price?





https://www.insider.com/worst-apartment-new-vork-city-tiktok-reactions-2021-2

#### Web Scraping and Data Cleaning

- Selenium/BeautifulSoup for web scraping
- Features:
  - # of bedrooms (BR)
  - # of bathrooms (BA)
  - Location
  - Square footage (SQFT)
  - Pets allowed
  - Application Fee
  - Broker's Fee
  - Laundry
  - Parking
- Cleaning: Dealing with outliers, incorrect/misplaced information, removing duplicate posts

\$1 / 1br - 800ft<sup>2</sup> - \$1,500+ (3-Units) w/Elevator+Laundry (near Queens Cente

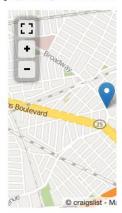


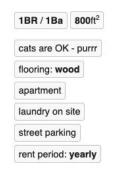


Good Income Required

STUDIO \$1,500 1BED 1BATH \$1,700 1BED 1BATH \$1,800 Move in ASAP

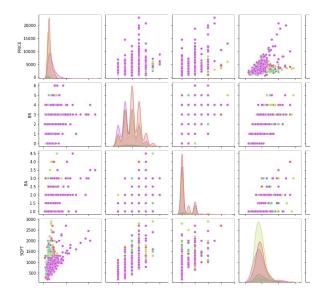
51st Avenue off Broadway Elmhurst, NY 11373 2 Blocks to R+M trains

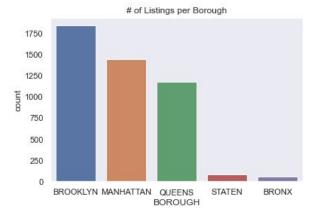




#### Data Prep and Feature Engineering

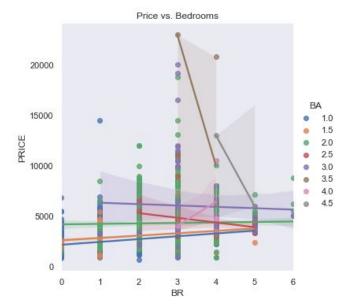
- Out of 4609 listings, only 969 listed square footage.
  - Imputed missing values based on median, borough, bedrooms.
- Created dummy variables for categoricals
  - o Boolean: pets, application fee, broker's fee
  - Washer/Dryer: none, in-unit, in-building
  - Parking: none, off-street (includes carports and garages),
     valet
  - Borough: Brooklyn, Manhattan, Queens, the Bronx, Staten Island

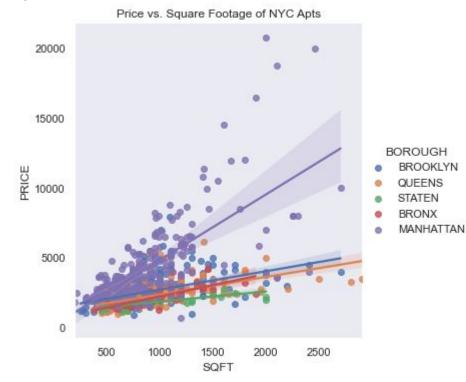




#### More Data Prep and Feature Engineering

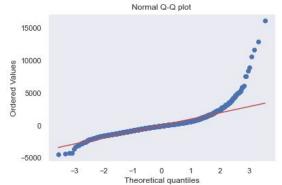
- Removed some features (application fee, pets)
- Added interaction terms
  - o BR/BA
  - SQFT \* MANHATTAN

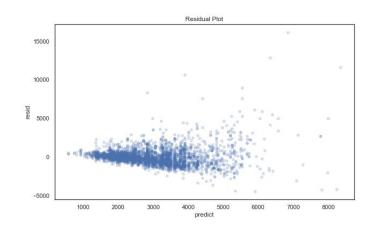




#### Issues

- Multicollinearity
- Heteroskedasticity







Correlation Heatmap 0.14 -0.26 -0.096 PRICE -0.058 -0.0052 0.35 -0.0052 -0.076 -0.019 -0.09 0.057 -0.01 -0.09 0.2 -0.16 -0.044 -0.011 BR -0.0088 -0.005 0.22 0.091 -0.005 0.047 -0.0049 -0.041 -0.015 SQFT -0.089 0.0074 0.13 0.1 0.0074 0.045 -0.03 -0.046 0.055 APP\_FEE -0.058 -0.019 -0.0088 -0.089 0.19 -0.023 -0.032 0.19 0.12 -0.11 -0.036 -0.021 BROK\_FEE -0.0052 -0.09 -0.005 0.0074 0.19 -0.14 -0.12 -0.16 0.031 0.14 -0.0015 WD 0.057 0.22 0.13 -0.023 -0.14 0.21 -0.14 0.038 0.18 -0.045 -0.032 -0.12 -0.12 0.079 0.03 PARKING 0.14 -0.01 0.091 0.1 -0.12 0.21 0.047 -0.14 -0.12 0.14 PETS -0.0052 -0.09 -0.005 0.0074 0.19 -0.16 0.031 -0.0015 -0.48 BOROUGH\_BROOKLYN -0.076 0.2 0.047 0.045 0.12 -0.16 0.038 -0.12 -0.16 -0.11 -0.16 -0.0049 -0.03 -0.11 0.031 0.18 0.079 0.031 -0.093 BOROUGH\_MANHATTAN BOROUGH\_QUEENS -0.044 -0.041 -0.046 -0.036 0.14 0.03 0.14 -0.48 -0.081 BOROUGH\_STATEN -0.096 -0.011 -0.015 0.055 -0.021 -0.0015 -0.045 <u>0.047</u> -0.0015 -0.11 -0.093 -0.081 H PARKING PETS BOROUGH\_STATEN

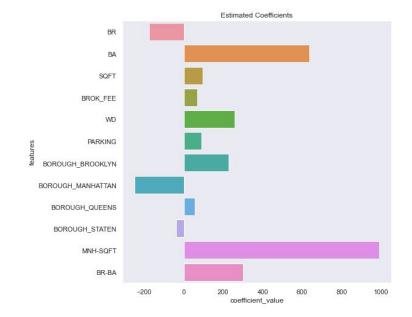
			- 1.00
vif	variables		
24.626210	BR	0	- 0.75
8.557337	ВА	1	- 0.50
4.497295	SQFT	2	
1.078687	BROK_FEE	3	- 0.25
1.184902	WD	4	- 0.00
1.087153	PARKING	5	
19.939513	BOROUGH_BROOKLYN	6	0.25
29.935476	BOROUGH_MANHATTAN	7	0.50
16.101196	BOROUGH_QUEENS	8	0.75
2.444164	BOROUGH_STATEN	9	0.75
12.682606	MNH-SQFT	10	1.00
17.592696	BR-BA	11	

#### **Model Selection and Results**

- Used kfold cross-validation on training set to get the R<sup>2</sup> score, mean absolute error (MAE) and root mean squared error (RMSE)
- Applied Ridge and Lasso regularization, compared scores
- Lasso regularization improved scores slightly

#### TRAINING DATA

- R^2 value: 0.526
- Mean Absolute Error: \$648.65
- Root Mean Squared Error: \$1,027.45



#### **TEST DATA**

- R^2 value: 0.519
- Mean Absolute Error: \$671.15
- Root Mean Squared Error: \$1,127.72

#### Conclusions

- The model can only explain about 50% of an apartment's rental price
- Other possible factors: neighborhood, amenities, utilities
- External factors: housing policies, market forces

Real Estate

### New York Renters Face 70% Increases as Pandemic Discounts Expire

The era of widespread Covid concessions for apartment hur





**NEW YORK** 

New York City, Income Inequality Capital Of America, Now Also Facing Soaring Rent Prices: Report

#### **Future Steps**



- Get more data, use information from other sites (e.g. StreetEasy, apartments.com)
- Look into more features
- Figure out more accurate interaction terms
- Categorize by neighborhood instead of borough, get more information about location
- Try log transformation or use weighted least squares regression to deal with heteroskedasticity