

Predicting Home Sales in the NYC Suburbs

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Why

- For most families, home equity is the largest component of household wealth
- US total real estate market is more than \$35 Trillion dollars
- Post-pandemic desire for more space (demand for suburban versus urban homes)



- Time on market
- Square Feet
- Baths
- Address/location
- Distance to midtown





BeautifulSoup

Per Zipcode

- Median household
- Average gross household income
- Public school test performance
- Average commute time



Pandas, geopy, sklearn, matplotlib, seaborn

Data

Challenges:

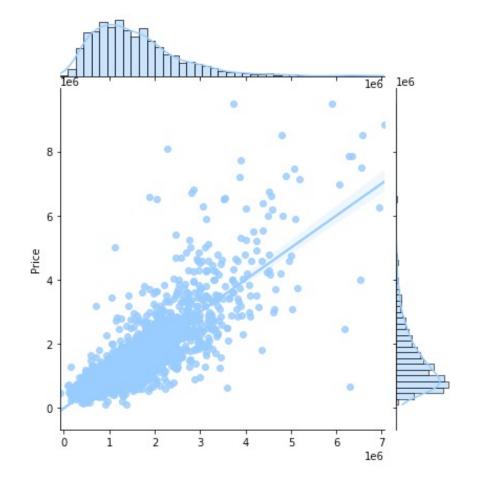
- Zillow's advanced web scrapping detection
 - Added request headers to mask HTML scrape from search page
 - Focused on features from aggregate search page as opposed to individual home pages
- Outliers
 - Dropped homes sold at greater than \$10mm to reduce variance in model
 - Dropped towns with less than 10 homes

Data focus:

- Sold homes (as opposed to currently listed homes)
 - ~3.5k homes over the past 6 months across 10 towns

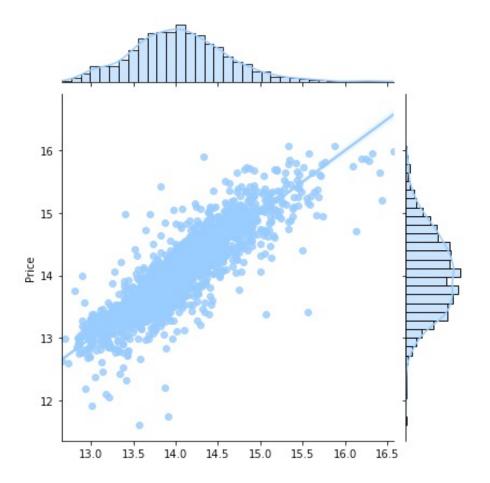
Step One: Basic OLS Regression

- Categorical features such as "Town" and "Zipcode" not taken into account
- Skewed distribution of prices
- Cross validation average score R² of .66



Step Two: Feature engineering and transformation

- Feature engineering:
 - Polynomial features: Beds²
 - Multiplicative features: School*Space
 - Formulaic features:
 - Average of commute time and distance to midtown
 - Total sum of square feet, beds, and baths
- Dummy variable creation for categorical features:
 - Zipcode dummies
 - Town dummies
- Transformation of target
 - Log transform on Y (Price) to adjust for skew
- Cross validation average R² score of .75



Step 3: Regularized models

- Standardize numeric features
- Solve for optimal alphas for lasso and ridge regularization
- Observe largest absolute value coefficients:
 - Ridge
 - Beds (+0.249)
 - Baths (+0.191)
 - Sqft (+.103)
 - Lasso
 - Average Gross Income (+5.602)
 - Zipcode 10573 (-2.617)
 - Sqft (+0.264)

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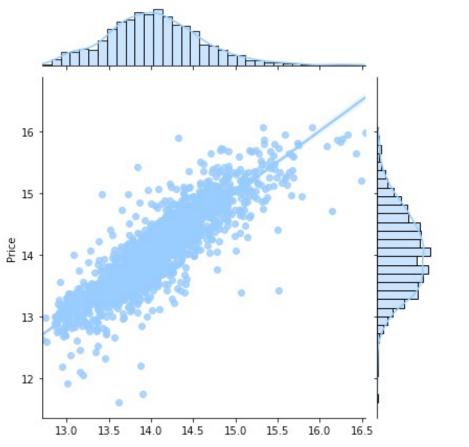
Lasso v Ridge training data

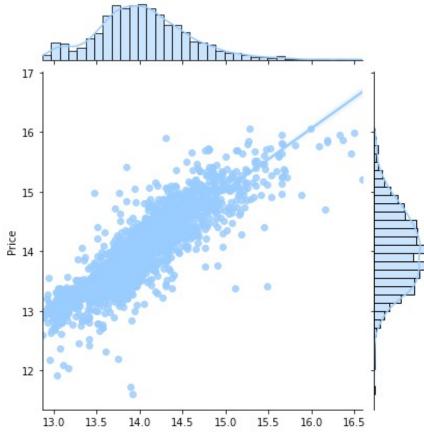
<u>Ridge</u>

CV average R^2 .754

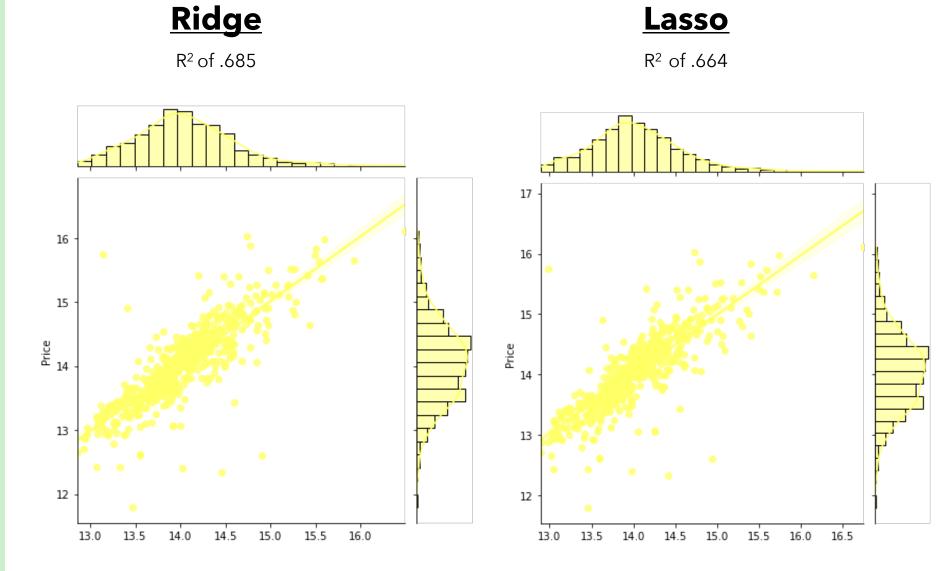
Lasso

CV average R².749



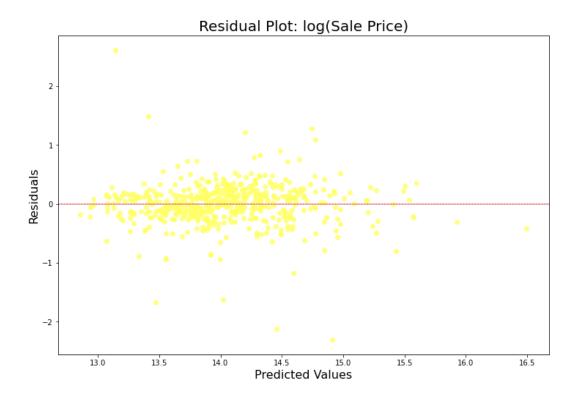


Lasso v Ridge test data

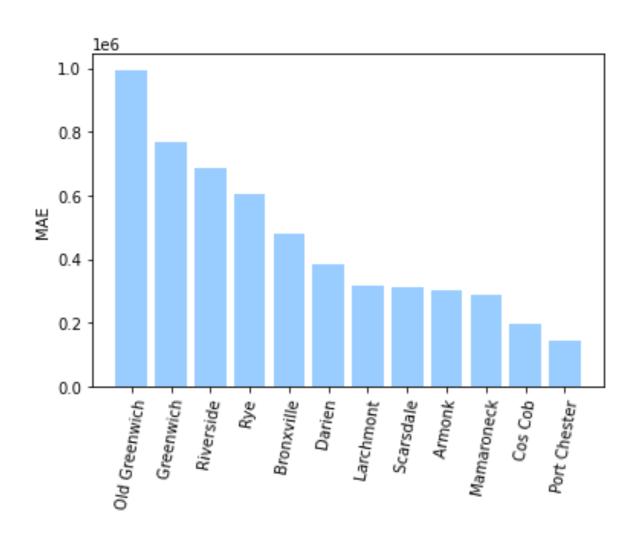


Are we overfitting...?

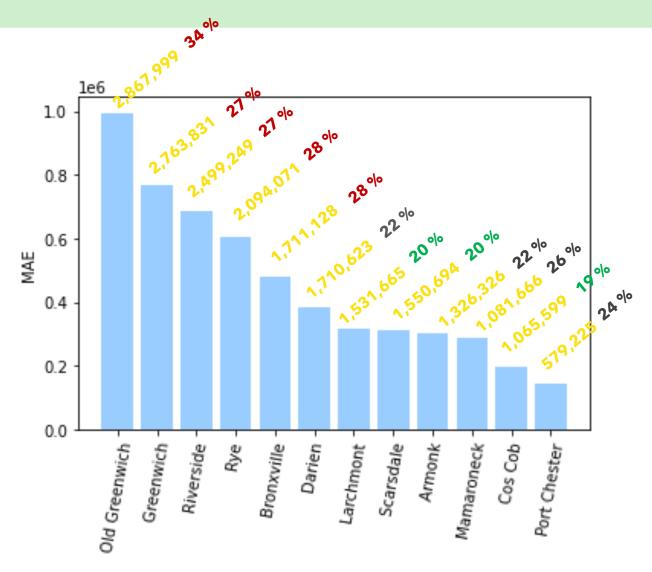


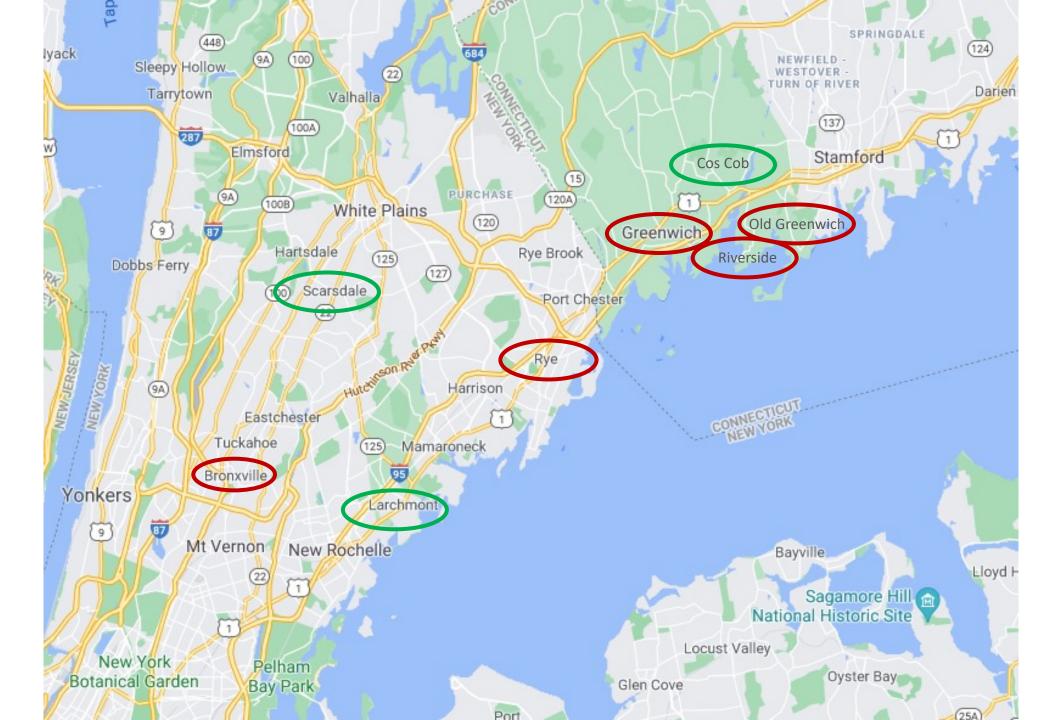


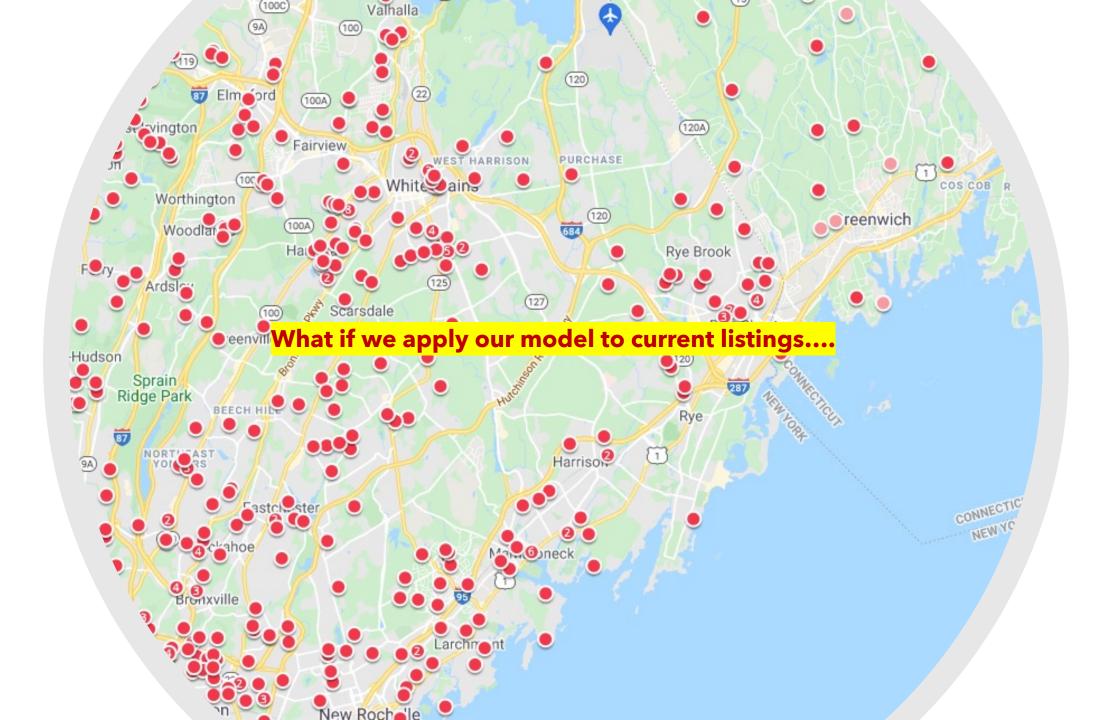
Mean Absolute Error: Town by Town Comparison



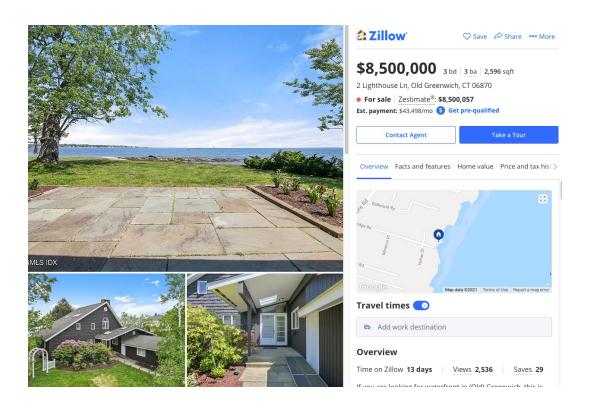
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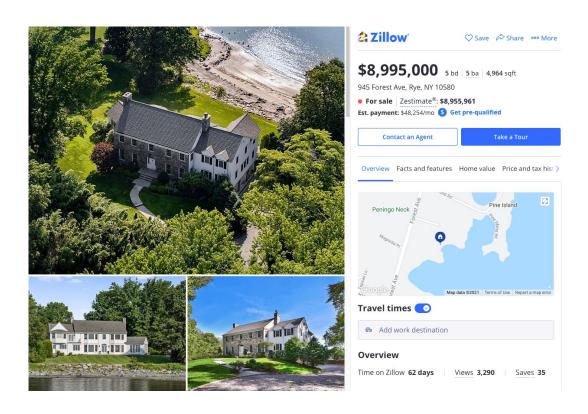




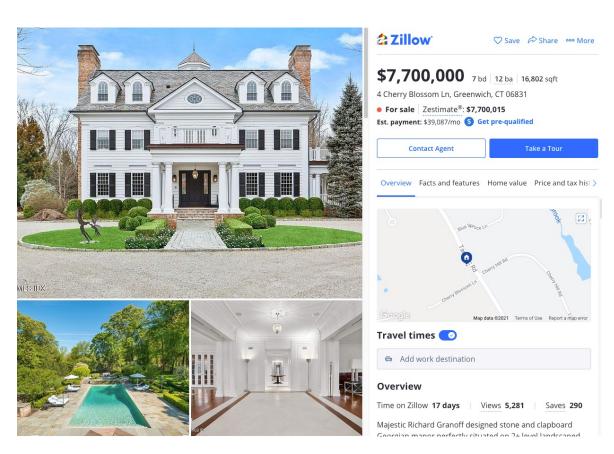


Most "over-priced" homes

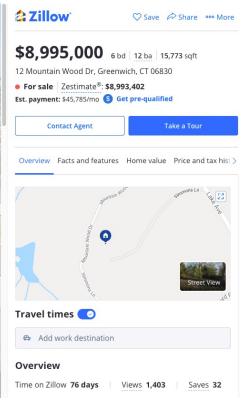




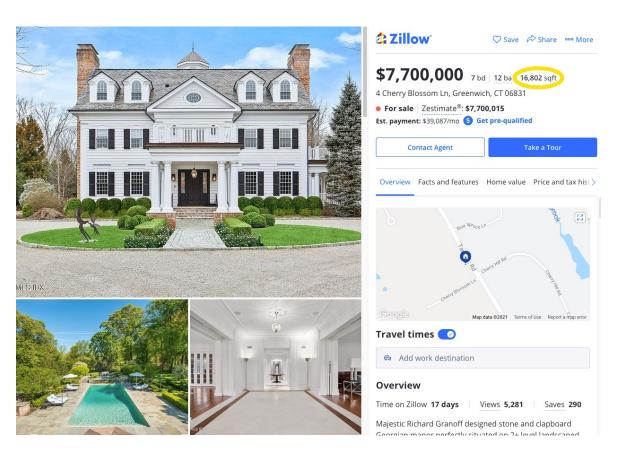
"Value" homes



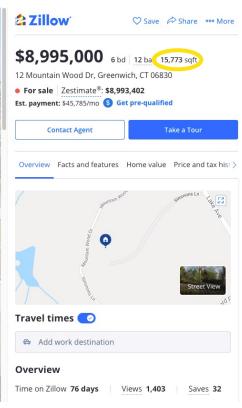




"Value" homes







Future Work

- Add more towns to my model
- Overlay tax conditions
- Build out selenium pipeline for more specific house data