Statistical Inferences into the Property Development Industry

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Outline

- Business Problem
- Data and Methods
- Results
- Conclusions
- Next Steps

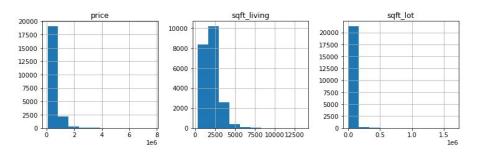
Business Problem

- Property development -- rewarding yet risky business to engage in
 - Can hard to make decisions when starting out as a new developer
 - Goal is to maximize profits while making informed decisions
- Aim to gain foundational understanding of best choices to make by asking questions:
 - What kinds of factors contribute to a house's base value?
 - When building a house, what factors should we look out for before beginning the construction of property?
 - How does the condition of a house at the point of sale influence the value of a house?
 - Are there specific features that add significant value to houses (e.g. rooms, floors)?

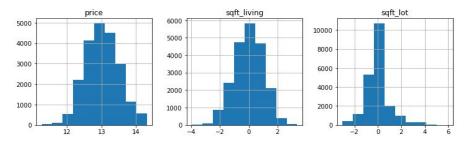
Data and Methods

- Dataset containing information about house sales in Kings County, California
- Use of feature scaling and transformation tools to normalize distribution of data and improve reliability of data

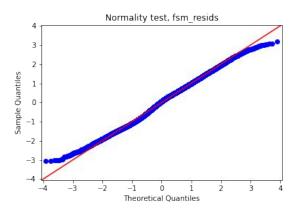
Before transformation:

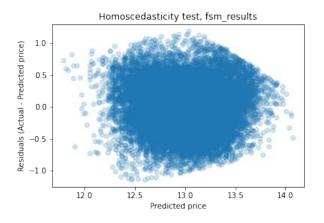


After transformation:

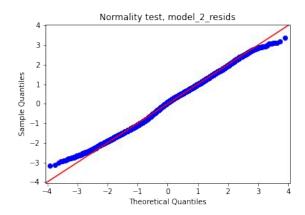


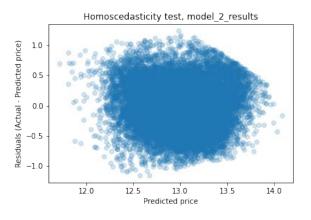
- Target variable = price
- Features:
 - sqft_living
- Regression statistic = 0.412
- Assumptions of linear regression:
 - √ Linearity
 - X Normality
 - X Homoscedasticity
 - ✓ Independence



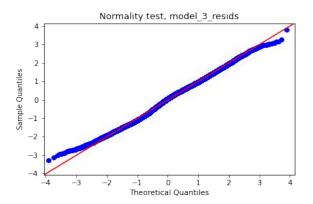


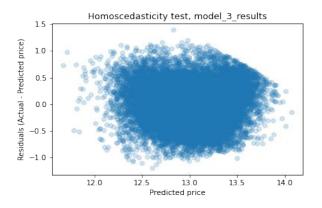
- Target variable = price
- Features:
 - sqft_living
 - o sqft_lot
- Regression statistic = 0.422
- Assumptions of linear regression:
 - √ Linearity
 - X Normality
 - X Homoscedasticity
 - ✓ Independence



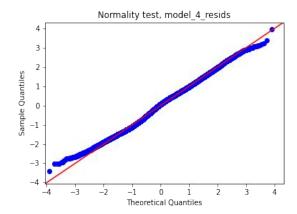


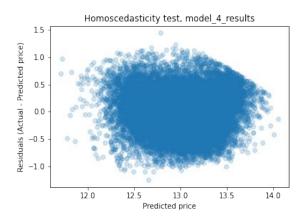
- Target variable = price
- Features:
 - sqft_living
 - sqft_lot
 - condition
- Regression statistic = 0.429
- Assumptions of linear regression:
 - √ Linearity
 - X Normality
 - X Homoscedasticity
 - ✓ Independence





- Target variable = price
- Features:
 - sqft_living
 - sqft_lot
 - condition
 - floors
- Regression statistic = 0.432
- Assumptions of linear regression:
 - √ Linearity
 - X Normality
 - X Homoscedasticity
 - X Independence





Results and Conclusion

- Third model is our best model -- gives us best regression statistic while meeting most assumptions of linear regression
 - A house's base value with no features would be about 12.670
 - For 1 unit increase in square footage of the house, price increases by 0.313
 - For 1 unit increase in square footage of the lot, price decreases by 0.041
 - For 1 unit increase in house's condition, price increases by 0.061
- Important factor: proportion between square footage of lot and house
- Condition and number of floors do have significant impact, but smaller
 - Rooms were not observed because of low correlation with price and multicollinearity with other variables

Next Steps

- Finding other methods to scale and transform data to gain better regression statistics and parameters
- Additional analysis on housing data in other counties
- Analyzing data about how housing prices change over time in response to economic factors (i.e. inflation)
 - May help us be proactive in decisions as prospective property developers

Thank you!

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