

PHASE 2 PRESENTATION

Linear regression project with King County house data

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Stakeholder

- BlackSock is a large real estate company
- They want to diversify by buying houses in King County, WA
- Plan: **buy** homes, **rent** them until their value goes up a sufficient amount, then **sell**
- Problem: BlackSock needs insight into what factors influence a home's price

Data

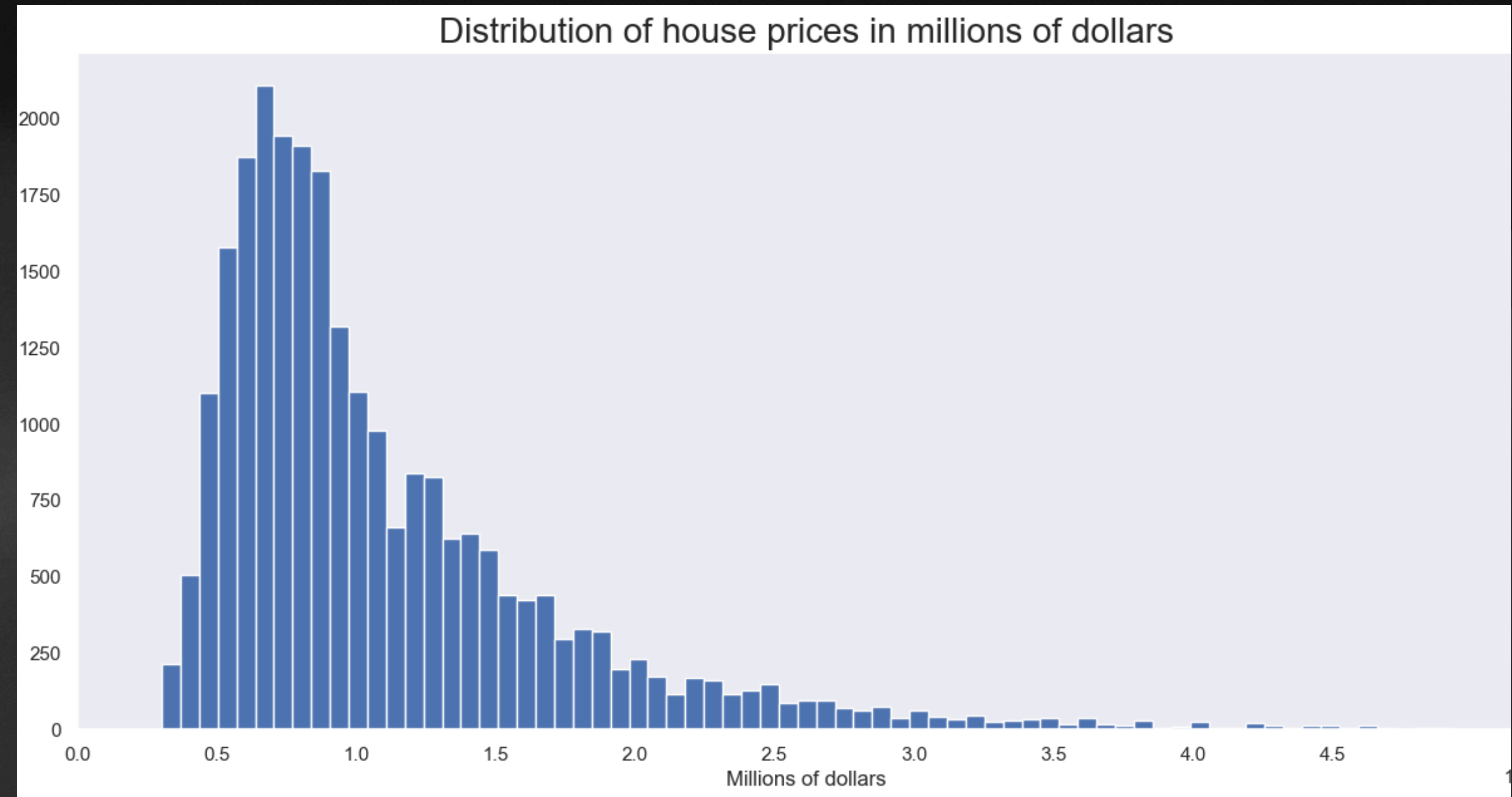
- Obtained from King County government website
 - <https://info.kingcounty.gov/>
- 30,000 records of home sales
- June 10th 2021 – June 9th 2022
- Has all the basic home information

Data – Important Variables

- Date built & sold
- Price sold for
- Square footage
- Bedrooms & bathrooms
- House condition
- Address
- Latitude/longitude

House price distribution

- This is a histogram of house prices in the dataset.
- Most house prices are between \$300,000 and \$2 million.



Modeling

- Data was used in a **linear regression model**
- This type of model is well suited for the data and problem, and it allows us to determine relationships among data as well as make predictions
- This has the potential to yield useful insights for our stakeholder

Data cleaning

- **Started** with roughly 30,000 values, **ended** with 26,000
- Data eliminated
 - Outliers
 - Null values
 - Data incompatible with linear regression
- **All visualizations** in this presentation were created with the cleaned dataset.

Data used

- Three variables
 - Average price of surrounding homes (“vicinity price”)
 - Home quality (“grade”)
 - Square footage of living space
- In the upcoming slides, I will show **scatterplots** of each of these variables vs. house price
- These scatterplots show that each variable has a **linear relationship** with price – as one variable goes up, price also goes up.
 - This made each of these variables good candidates for use in our model.

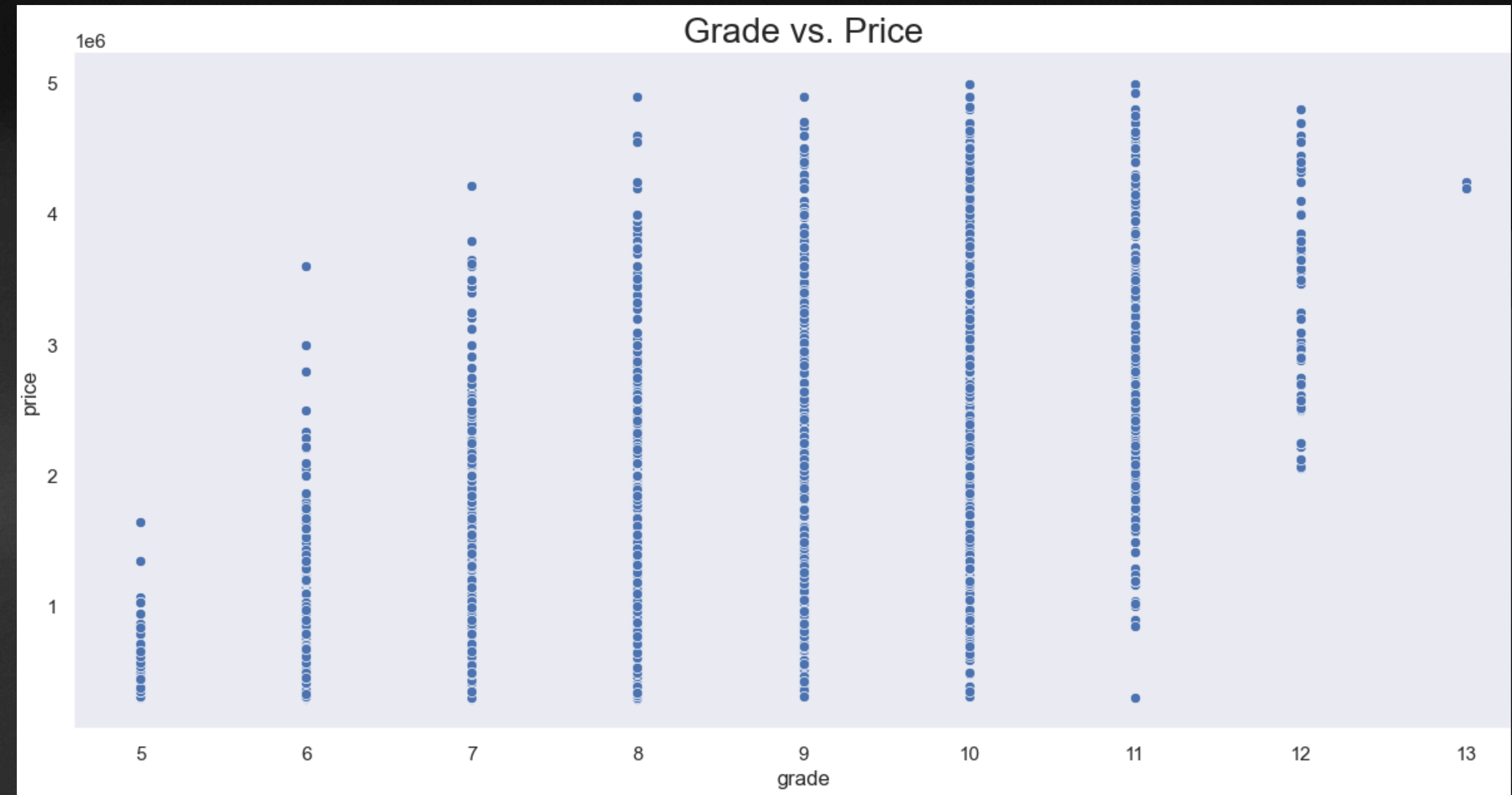
“Sqft_living” variable

- This variable measures the square footage of living space in a home.



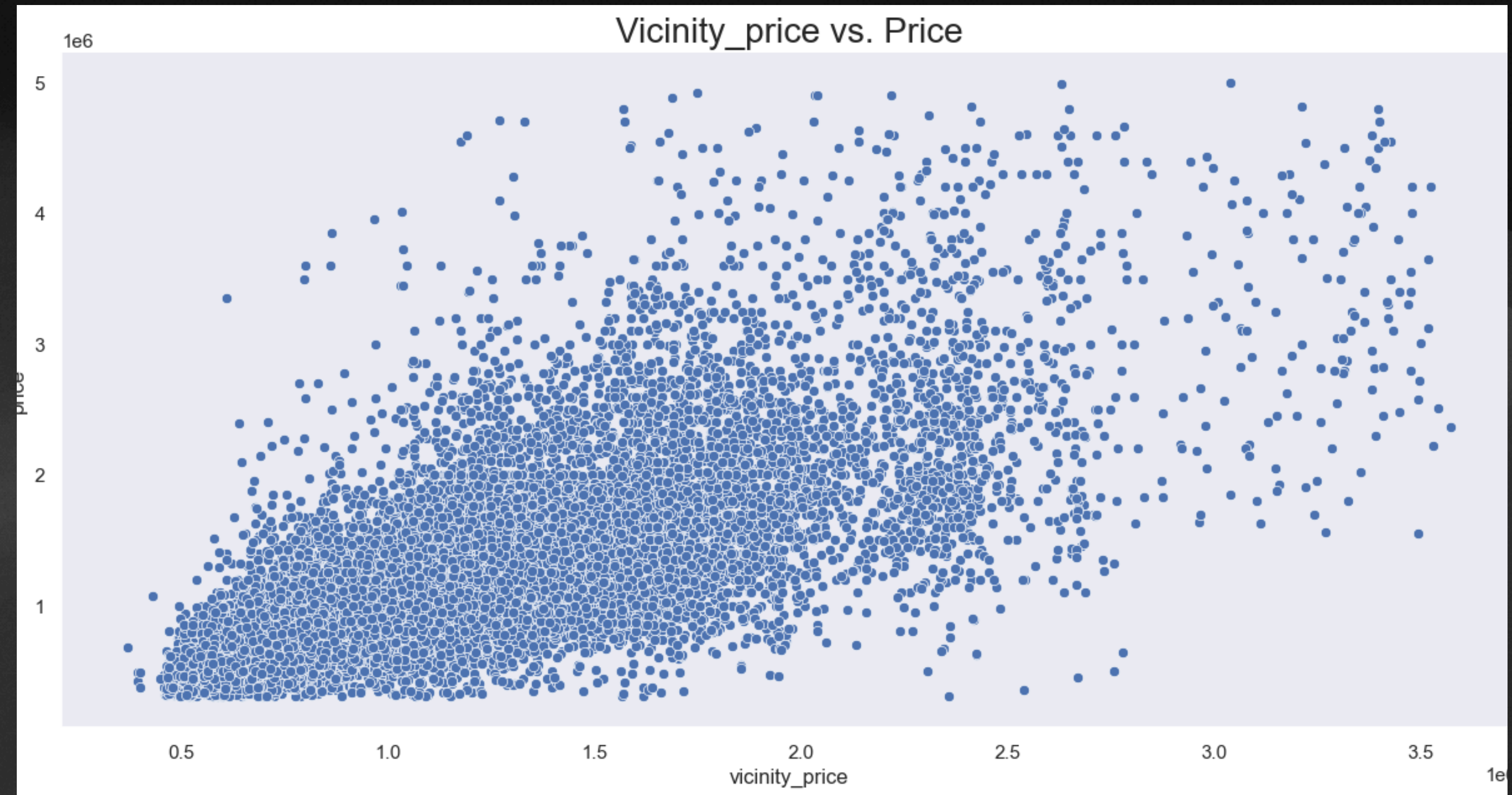
“Grade” variable

- Measures the quality of a home on a **13-point scale** after being sold.
- Houses ranging from 1-4 on this scale were considered outliers and eliminated from the dataset.
- **Full scale:**
 - 1 (Cabin), 2 (Substandard), 3 (Poor), 4 (Low), 5 (Fair), 6 (Low Average), 7 (Average), 8 (Good), 9 (Better), 10 (Very Good), 11 (Excellent), 12 (Luxury), 13 (Mansion)



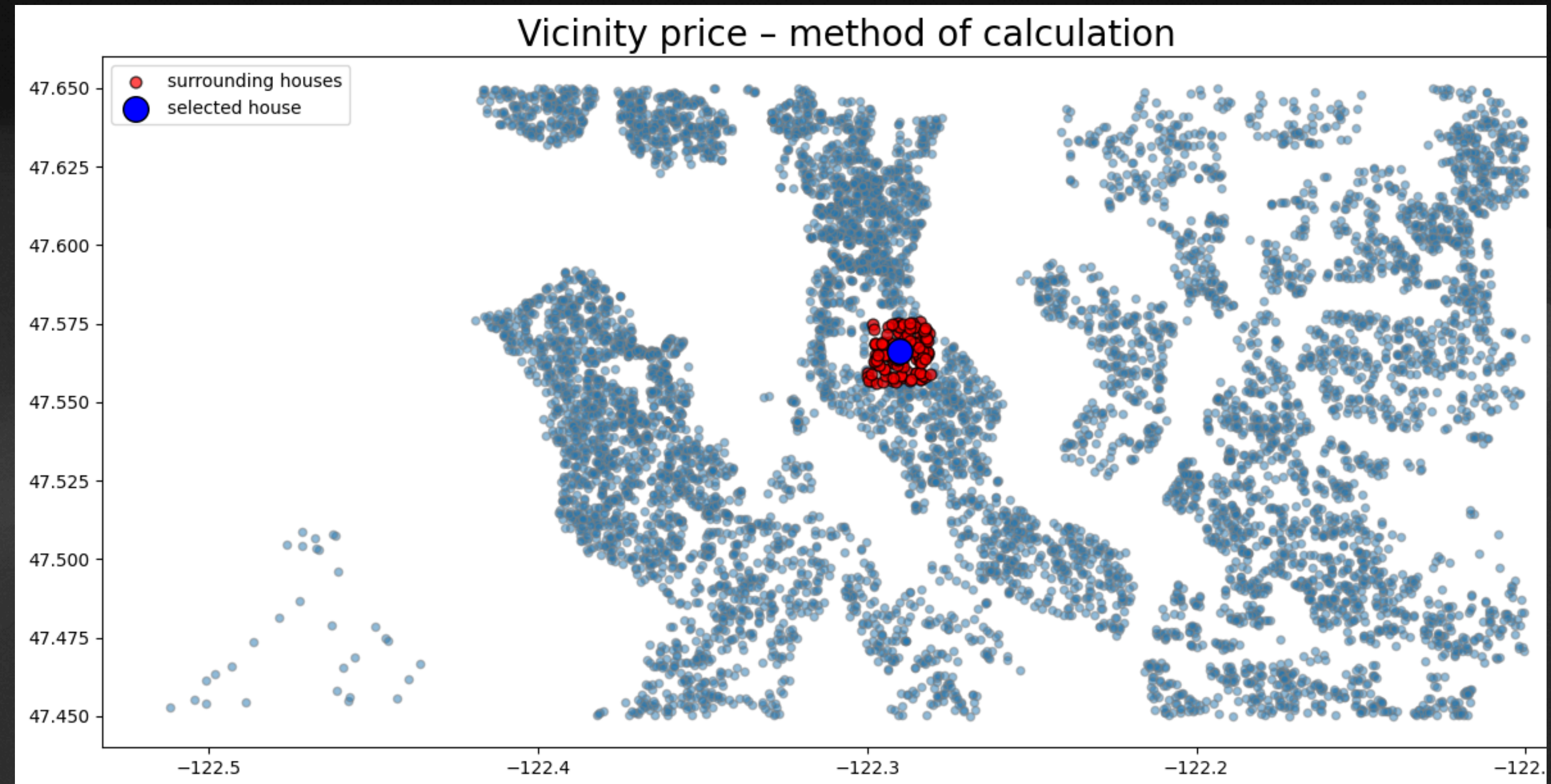
“Vicinity price” variable

- Measures the average price of nearby homes for each house in the dataset.
- This “average” price does not include the original house for which it was calculated, to avoid the house predicting its own price in the model.



Vicinity_price visual demonstration

- This is a visual demonstration of how we calculated the vicinity_price variable.
- For each house, we used its precise location in latitude/longitude to draw a small square-shaped area around the house.
- In this area, we averaged the price of all the surrounding houses (the red dots).
- We eliminated the original house (the blue dot) from the average to avoid houses predicting their own price in the model.



Model results

- We used an **iterative approach** to modeling; that is, we started with a basic model and made adjustments from there.
- A house with average living space, an average grade, and average-priced surrounding homes will cost 1.1 million dollars.
- For every dollar increase in surrounding homes, a house's price goes up by about 76 cents.
- Every increase on the home quality scale leads to a \$85,000 increase in home price
- Every additional square foot of living space increases house price by \$260.

Model performance

- The adjusted **R-Squared** of the final model was 0.735
- The **MAE** (mean absolute error) of the final model was \$222,000
- These metrics show that the model is a **decent fit** for the data

Recommendations

- Our data was limited, and as a result, our model can only give **preliminary insights** at best.
- Although we recommend the stakeholder take note of our model's results, we also recommend the stakeholder consult other models in addition to our own before making any significant decisions.
- We recommend buying houses whose surrounding properties are likely to go up in value.
- We recommend increasing home quality if doing so costs less than \$85,000.

Thank you!

- All questions are welcome
- Contact me at: angelo.turri@gmail.com