Predicting & Comparing House Prices in Istanbul and London DSA210

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Project Overview

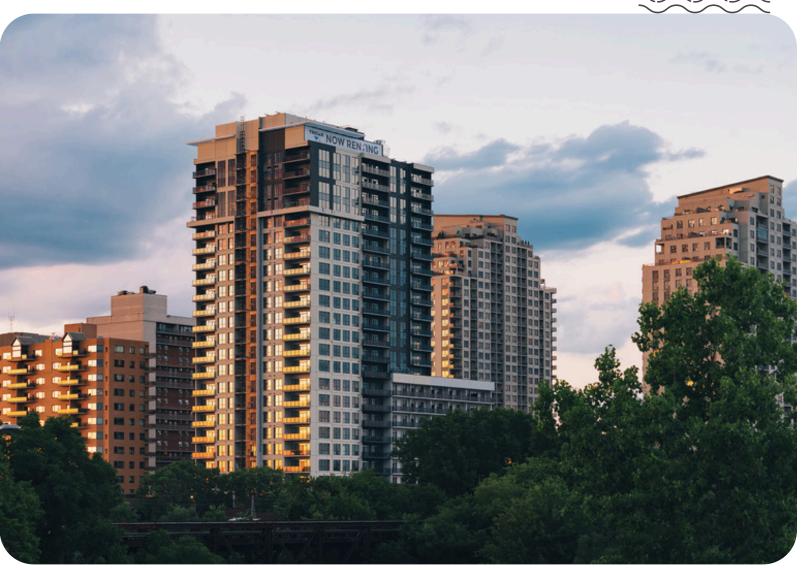
Objective:

This project aims to compare residential property prices in London and Istanbul by analyzing key influencing factors such as size and number of bathrooms.

Approach:

A combination of data preprocessing, exploratory data analysis (EDA), and machine learning was used to uncover trends and build predictive models.





Data Sources & Preprocessing

Datasets:

- London Dataset Kaggle
- <u>Istanbul Dataset Kaggle</u>

Preprocessing:

- Standardized column names
- Converted Turkish Lira to GBP (1 GBP = 20.38 TL)
- Cleaned missing or invalid entries
- Unified units (e.g., square meters)



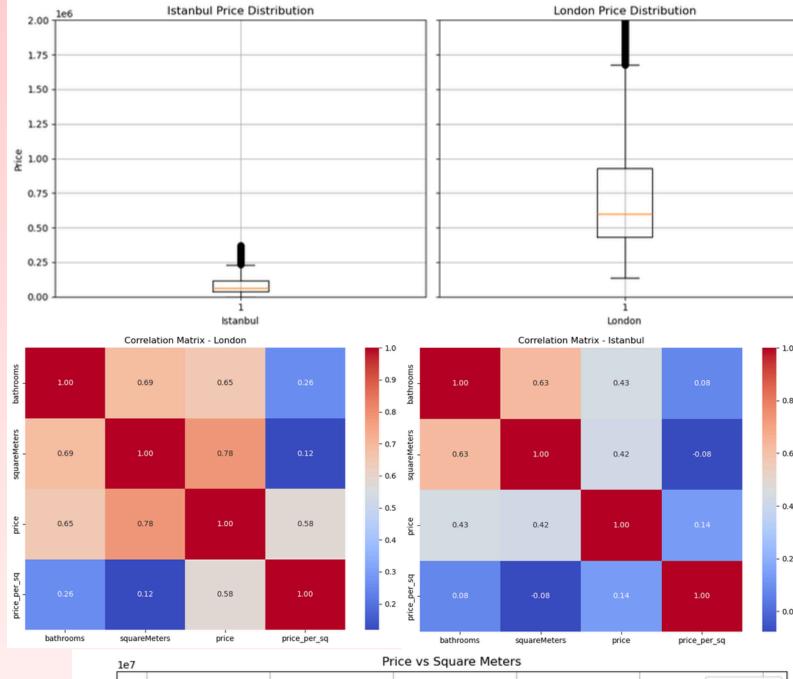
Exploratory Data Analysis (EDA)

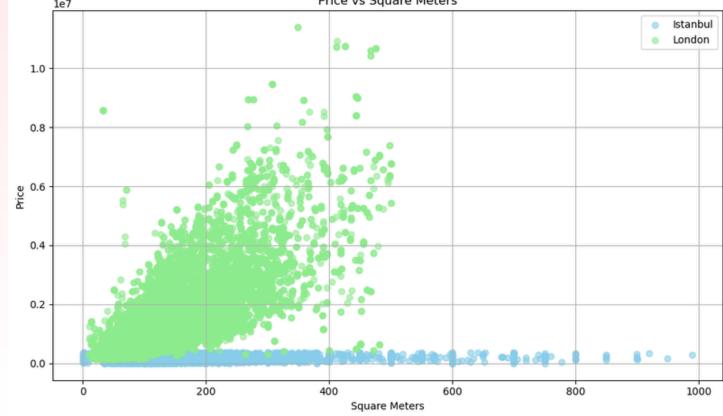
Data Visualizations:

- Multiple plot types were used (histograms, boxplots, scatterplots)
- A correlation matrix was created to explore feature relationships

Preprocessing:

- London house prices are significantly higher than Istanbul's (in GBP). And price is correlated with square meters(0.78) and bathrooms (0.69)
- Istanbul homes tend to be larger in square meters, but pricing is less predictable. This suggests location-based and unobserved factors may dominate in Istanbul





T-statistic: -139.741

P-value: 0.0000

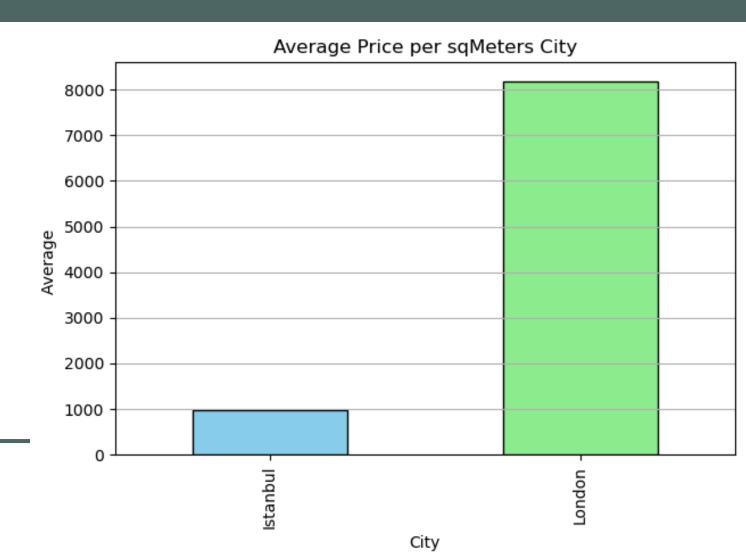
HO rejected: The average price per square meter in Istanbul and London is statistically different .

Null Hypothesis H0:

There is no statistically significant difference in the average price per square meter between Istanbul and London

Alternative Hypothesis H1:

The average price per square meter in Istanbul and London is statistically different .



T-statistic: 67.223

One-sided p-value: 0.0000

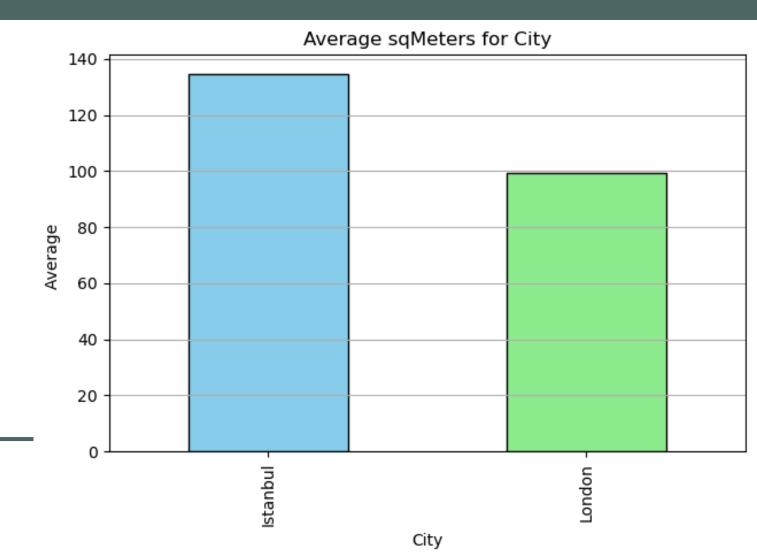
HO rejected: Houses in Istanbul are significantly larger then London.

Null Hypothesis H0:

We cannot conclude that houses in Istanbul are larger then London.

Alternative Hypothesis H1:

Houses in Istanbul are significantly larger then London.



Pearson Correlation Coefficient: 0.777

P-value: 0.0000

HO rejected: There is a statistically significant relationship

between price and square meters in London.

Null Hypothesis H0:

There is no statistically significant relationship between price and square meters in London.

Alternative Hypothesis H1:

There is a statistically significant relationship between price and square meters in London.



Pearson Correlation Coefficient: 0.416

P-value: 0.0000

HO rejected: There is a statistically significant relationship between price and square meters in iStanbul.

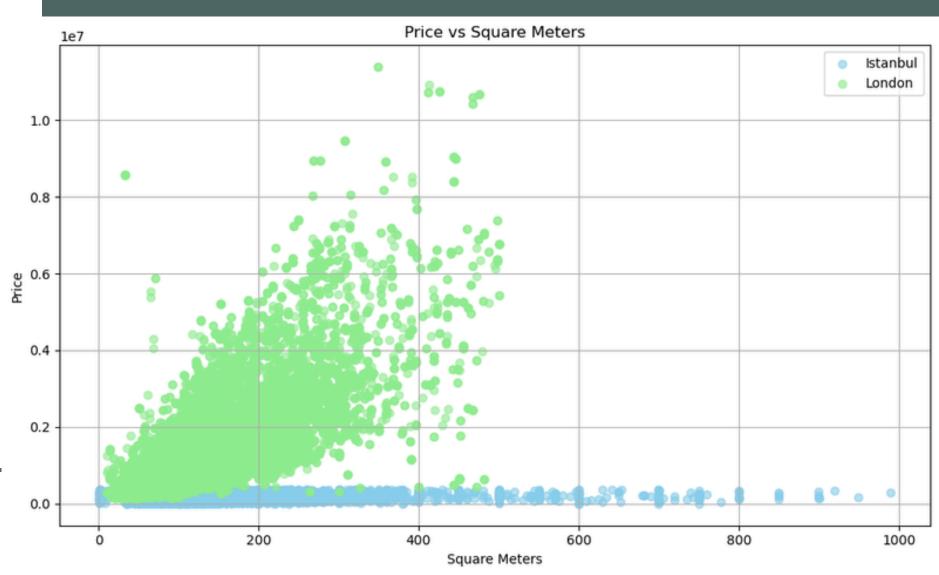
However with checking graph and low pearson core. coeffficient we cannot sa there is a significant relationship.

Null Hypothesis H0:

There is no statistically significant relationship between price and square meters in İstanbul.

Alternative Hypothesis H1:

There is a statistically significant relationship between price and square meters in İstanbul.



Machine Learning Part



Goal:

To predict house prices in Istanbul and London using square meters and number of bathrooms.

Models Used

- Linear Regression
- Random Forest Regressor
- XGBoost Regressor

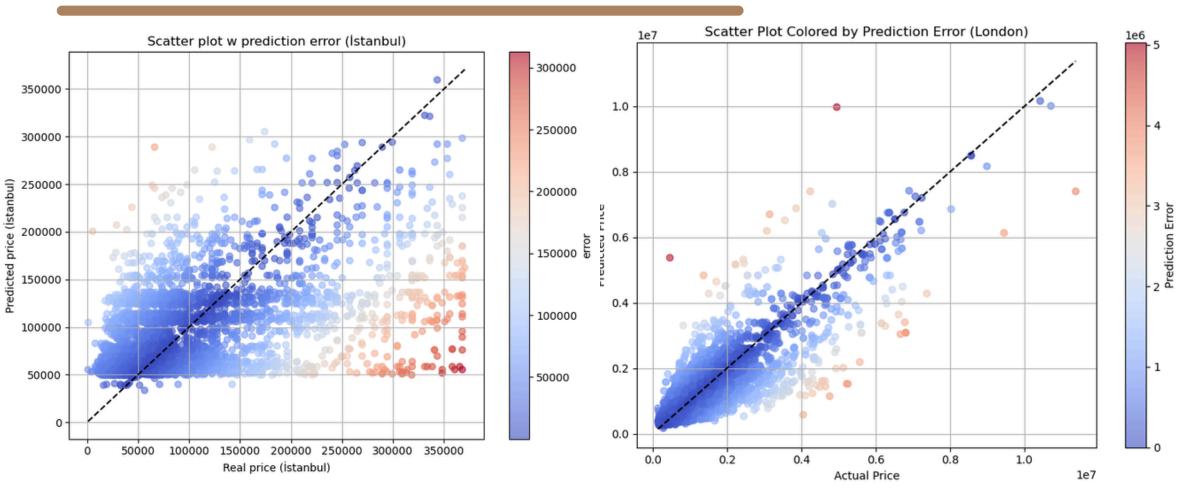
Evaluation Metrics

- R² Score
- MAE (Mean Absolute Error)
- RMSE (Root Mean Squared Error)
- MAPE (Mean Absolute Percentage Error)

Machine Learning Part







- Istanbul Feature Importances: squareMeters: 0.8467 bathrooms: 0.1533
 London Feature Importances: squareMeters: 0.9121 bathrooms: 0.0879
- Models perform well in London due to strong correlation and cleaner data. Istanbul shows weak results, which may be resulted from location-based and unobserved factors

Conclusion

Key Takeaways:

- London housing prices are significantly higher than Istanbul's.
- Price per square meter is statistically different between the two cities (H₀ rejected).
- Houses in Istanbul are larger, but have weaker correlation with price.
- Square meters is the most important feature in both cities.

Machine Learning Summary:

- Models predict London house prices not bad (R² > 0.70).
- Istanbul models perform poorly may due to missing location-specific data and market irregularities.
- Tree-based models (Random Forest, XGBoost) showed better performance than linear models.

