First Estimator



House prices prediction model by Maria | Apostolis | Nikos

WHY?- Why did we develop the ML model?

Empowering a real estate company to make better, faster pricing decisions.

 The real estate market is competitive and accurate pricing is crucial for success.

 Brokers need a reliable starting point to estimate house prices quickly and effectively.

• A machine learning model would be a solution that **saves a lot of time**, since it can predict a **first estimation** of each house's price.





HOW?- How Did We Approach This?

We built a model based on the features brokers consider important.

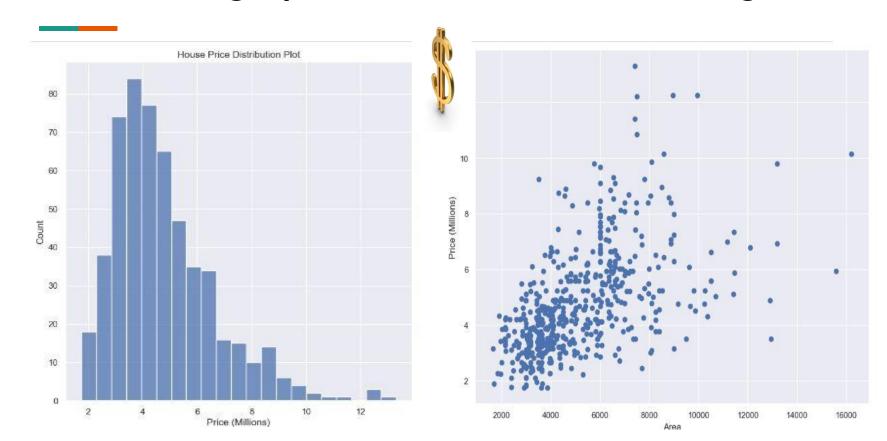
• **Understanding the Data**: Analyzed the dataset provided by brokers, focusing on key features influencing house prices (e.g., number of bedrooms, size).

Model Development:

- We analyzed relationships between these features and historical house prices.
- Built a predictive system to estimate prices based on **new inputs**.

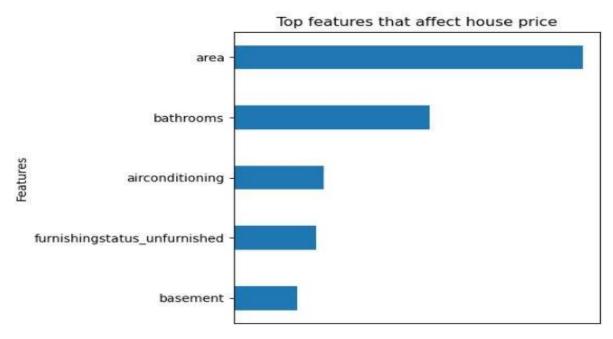
• **Evaluation**: Tested the model to ensure the estimates **align** closely with real market prices.

HOW?-Some graphs for better understanding



HOW - What Features Are Considered in the Model?

Features brokers should prioritize in pricing decisions.





HOW - How Does the Model Work?

Turning raw data into actionable insights.

- 1. The **broker inputs a house's features** into the model.
- 2. The model analyzes these features based on patterns it learned from historical data.
- 3. The model provides a **price estimate**, acting as a **starting point** for brokers.



Features Model analyzing.. Price Estimation

WHAT?- What did we achieve?

A tool that delivers first price estimations, with an R² of 0.66 and MAPE of 0.18.

- The model can predict house prices within a **reasonable range** of market values.
- It **complements brokers' expertise** by offering a consistent, data-backed starting price.

• **Time Efficiency**: Provides quick estimates, reducing manual analysis.

Improving the Model

To make the model even better, we need more detailed data.

Additional data could improve accuracy:

- Neighborhood trends, such as upcoming developments
- Market conditions, like seasonality or demand shifts
- Crime rate
- Kitchen, garden, pool, fireplace, etc.
- Year built



Call to Action! Collaboration with the company to collect more features that will enhance our model.

Closing and Next Steps

Next Steps:

 Collaborate with the company to gather more feature information and improve the performance of the model.

• **Pilot** the tool in a selected region, gather **feedback** and improve.

Explore how this tool can integrate into brokers' workflows seamlessly.

We're excited to hear your feedback and work together to bring this tool to life!

