Project Proposal

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Dataset: Kaggle Competition: House Price Prediction

Data source:

According to Kaggle, <u>Ames Housing dataset</u> was compiled by Dean De Cock. The House Prices dataset is from the Kaggle platform. It has two CSV files, "train.csv" and "test.csv" and has <u>79 explanatory variables</u> describing (almost) every aspect of residential homes and contains information on various features of houses sold in Ames, Iowa from 2006 to 2010.

Motivation and goals:

Our group's motivation is to build a model that accurately predicts the sale price of a house based on various features. This model can be used by various stakeholders such as real estate agents, homeowners, and property investors to make informed decisions about buying, selling, or investing in a property. The model should be able to provide accurate predictions of house prices, which will enable homeowners to price their homes competitively and real estate agents to provide informed advice to their clients. Additionally, the model can be used by property investors to identify undervalued properties and make informed decisions about buying or selling them.

The ultimate aim of our project is to create a model that performs well on new and unseen data, is interpretable, and provides insights into the most important factors that drive house prices. This will enable stakeholders to gain a better understanding of the housing market and make informed decisions based on the insights provided by the model. We expect our model to have the potential to provide valuable insights and assist stakeholders in making informed decisions related to the housing market.

Preliminary ideas on technique:

In pursuing our goal, our team will apply ensemble leaning to attempt for better predictive performance through the combination of predictions from multiple models. To build the models, we plan to first use ridge regression and random forest regression. We will test the models through cross-validation and find the optimal ridge and RF results by looking at the CV scores. Then, we plan on applying bagging method and boosting method to find and compare the optimal number of regressors. When we arrive at our first-round model and its predictions, we may try to apply the logic of model stacking here. We will try to use the predicted housing price as our new input to remodel again for a new set of predictions and compare our results.

Exploratory data analysis:

You may find our exploratory data analysis <u>here</u>.