

## USA\_Housing Dataset Description

USA\_Housing dataset contains 5000 rows. The dataset has 6 columns, and the prices column is the target value. There is non-null in USA\_Housing dataset. Also, there is no duplicated rows in USA\_Housing dataset as shown in Figure 1 below.

```
data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5000 entries, 0 to 4999
Data columns (total 7 columns):
# Column No
         Avg. Area Income
                                                             5000 non-null
        Avg. Area House Age
Avg. Area Number of Rooms
Avg. Area Number of Bedrooms
Area Population
                                                             5000
                                                                    non-null
                                                                                         float64
                                                            5000 non-null
                                                                                         float64
                                                             5000 non-null
                                                                                         float64
        Price
                                                                                         float64
                                                             5000
                                                                     non-null
         Address
                                                             5000 non-null
dtypes: float64(6), object(1)
memory usage: 273.6+ KB
```

Figure 1: USA\_Housing dataset info

## Exploratory Data Analysis (EDA)

We used Seaborn and Matplotlib for data visualization. We used the seaborn heatmap to represents the collinearity of the multiple features in the dataset. We used data.corr() to show the correlation between the features. We used these methods to find the answer to our questions

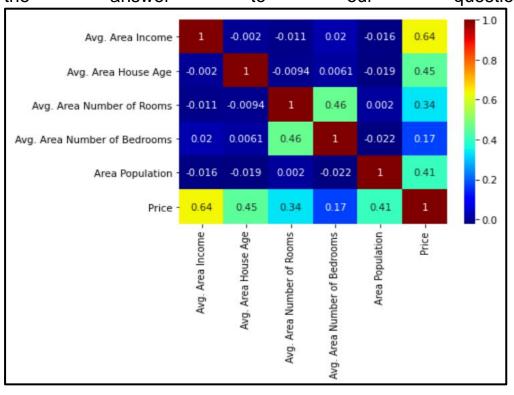


Figure 2: USA\_Housing dataset heatmap

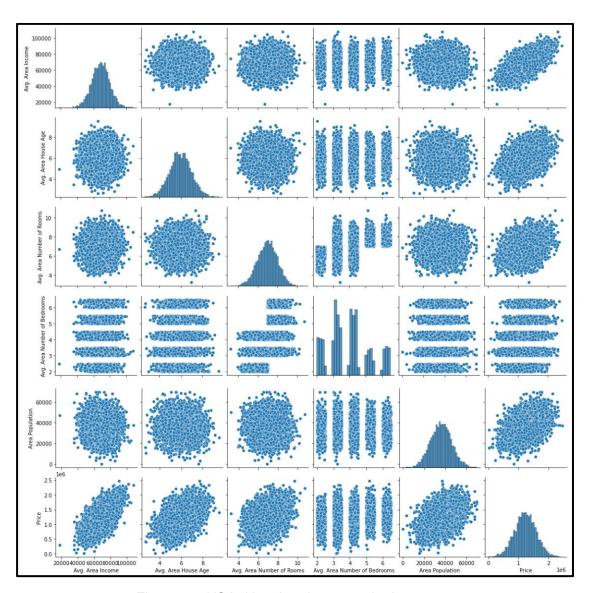


Figure 3: : USA\_Housing dataset pair plot

From In Figure 2 and 3, we can see there is no clear relation between 'Avg. Area Number of Bedrooms' and the target 'Price'. The highest correlation among all features is between 'Avg. Area Income' feature and the target 'Price'. Also, from the graph we find the three features 'Avg. Area House Age' 'Avg. Area Number of Rooms' and 'Area Population' had medium correlation with the target 'Price'.

Figure 4, illustrates a strong relationship between 'Avg. Area Income' and 'price'. It shows that the higher in come of individuals the higher the prices of houses are, and vice versa.

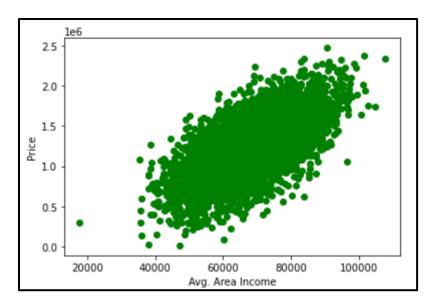


Figure 4: Scatter plots between 'Avg. Area Income' and price

The price has normal distribution as shown in Figure 5 below.

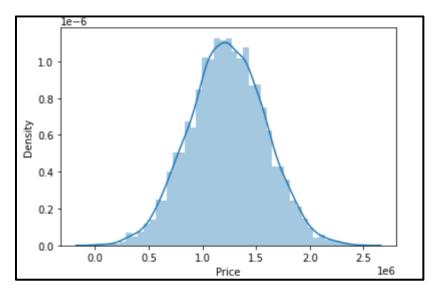


Figure 5: Price distribution plot

## ❖ Implementation:

We used Python and Jupyter notebooks for implement MVP - Predicting USA\_Housing prices. The link for Python notebook is

https://github.com/FatimahNainf/Project/blob/main/MVP%20-%20Predicting%20USA\_Housing%20prices.ipynb