

housing-price-prediction

August 3, 2023

```
[1]: from google.colab import files
      upload=files.upload()
```

<IPython.core.display.HTML object>

Saving Housing.csv to Housing.csv

```
[2]: #import Library

import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
[3]: #import Data_Sets
      df=pd.read_csv('Housing.csv')
```

```
[4]: df.head()
```

```
[4]:
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	13300000	7420	4	2	3	yes	no	no	
1	12250000	8960	4	4	4	yes	no	no	
2	12250000	9960	3	2	2	yes	no	yes	
3	12215000	7500	4	2	2	yes	no	yes	
4	11410000	7420	4	1	2	yes	yes	yes	

	hotwaterheating	airconditioning	parking	prefarea	furnishingstatus
0	no	yes	2	yes	furnished
1	no	yes	3	no	furnished
2	no	no	2	yes	semi-furnished
3	no	yes	3	yes	furnished
4	no	yes	2	no	furnished

```
[5]: #Housing Prices Data set info
      df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544

Data columns (total 13 columns):

#	Column	Non-Null Count	Dtype
0	price	545 non-null	int64
1	area	545 non-null	int64
2	bedrooms	545 non-null	int64
3	bathrooms	545 non-null	int64
4	stories	545 non-null	int64
5	mainroad	545 non-null	object
6	guestroom	545 non-null	object
7	basement	545 non-null	object
8	hotwaterheating	545 non-null	object
9	airconditioning	545 non-null	object
10	parking	545 non-null	int64
11	prefarea	545 non-null	object
12	furnishingstatus	545 non-null	object

dtypes: int64(6), object(7)

memory usage: 55.5+ KB

```
[6]: #Data Describes
df.describe(include=object)
```

```
[6]:      mainroad  guestroom  basement  hotwaterheating  airconditioning  prefarea  \
count      545         545         545              545              545         545
unique         2          2          2                2                2          2
top          yes         no         no                no                no         no
freq         468         448         354              520              373         417

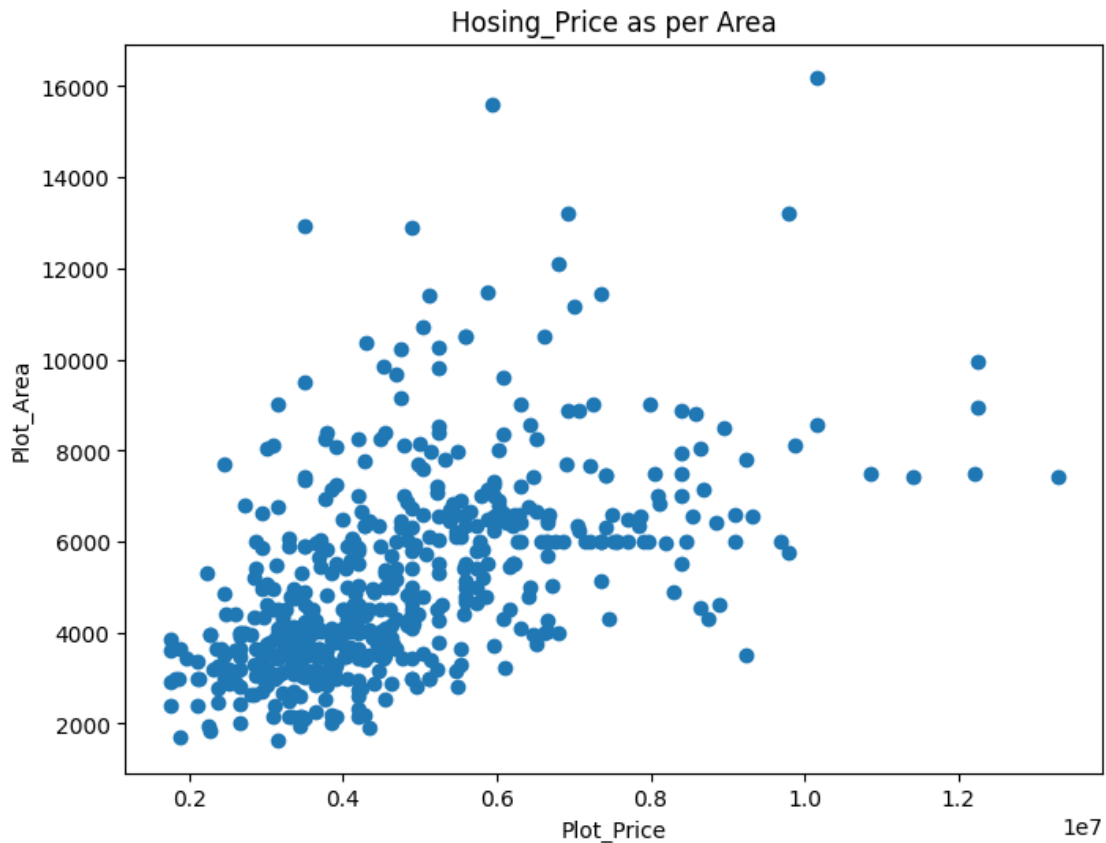
      furnishingstatus
count              545
unique              3
top      semi-furnished
freq              227
```

```
[7]: #Checking Missing Values
df.isna().sum()
```

```
[7]: price      0
area        0
bedrooms    0
bathrooms   0
stories     0
mainroad    0
guestroom   0
basement    0
hotwaterheating  0
airconditioning  0
```

```
parking          0
prefarea         0
furnishingstatus 0
dtype: int64
```

```
[8]: plt.figure(figsize=(8,6))
plt.scatter(x='price',y='area',data=df)
plt.xlabel("Plot_Price")
plt.ylabel("Plot_Area")
plt.title("Hosing_Price as per Area")
plt.show()
```



```
[9]: df.columns
```

```
[9]: Index(['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'mainroad',
          'guestroom', 'basement', 'hotwaterheating', 'airconditioning',
          'parking', 'prefarea', 'furnishingstatus'],
          dtype='object')
```

```
[10]: Categorical_Col = []
      Numerical_Col = []
```

```
[11]: for col in df.columns:
      if df[col].dtype=='object':
          Categorical_Col.append(col)
      else:
          Numerical_Col.append(col)
```

```
[12]: print(Categorical_Col)
      print(Numerical_Col)
```

```
['mainroad', 'guestroom', 'basement', 'hotwaterheating', 'airconditioning',
'prefarea', 'furnishingstatus']
['price', 'area', 'bedrooms', 'bathrooms', 'stories', 'parking']
```

```
[13]: #Encoding
      from sklearn.preprocessing import LabelEncoder
```

```
[14]: le=LabelEncoder()
      df['furnishingstatus']=le.fit_transform(df['furnishingstatus'])
```

```
[15]: #OneHotEncoder
      df=pd.get_dummies(df)
```

```
[16]: df.head()
```

```
[16]:
```

	price	area	bedrooms	bathrooms	stories	parking	furnishingstatus	\
0	13300000	7420	4	2	3	2	0	
1	12250000	8960	4	4	4	3	0	
2	12250000	9960	3	2	2	2	1	
3	12215000	7500	4	2	2	3	0	
4	11410000	7420	4	1	2	2	0	

	mainroad_no	mainroad_yes	guestroom_no	guestroom_yes	basement_no	\
0	0	1	1	0	1	
1	0	1	1	0	1	
2	0	1	1	0	0	
3	0	1	1	0	0	
4	0	1	0	1	0	

	basement_yes	hotwaterheating_no	hotwaterheating_yes	airconditioning_no	\
0	0	1	0	0	
1	0	1	0	0	
2	1	1	0	1	
3	1	1	0	0	
4	1	1	0	0	

	airconditioning_yes	prefarea_no	prefarea_yes
0	1	0	1
1	1	1	0
2	0	0	1
3	1	0	1
4	1	1	0

```
[17]: df.shape
```

```
[17]: (545, 19)
```

```
"""Column Name: mainroad Unique Values: ['yes' 'no']
```

```
Column Name: guestroom Unique Values: ['no' 'yes']
```

```
Column Name: basement Unique Values: ['no' 'yes']
```

```
Column Name: hotwaterheating Unique Values: ['no' 'yes']
```

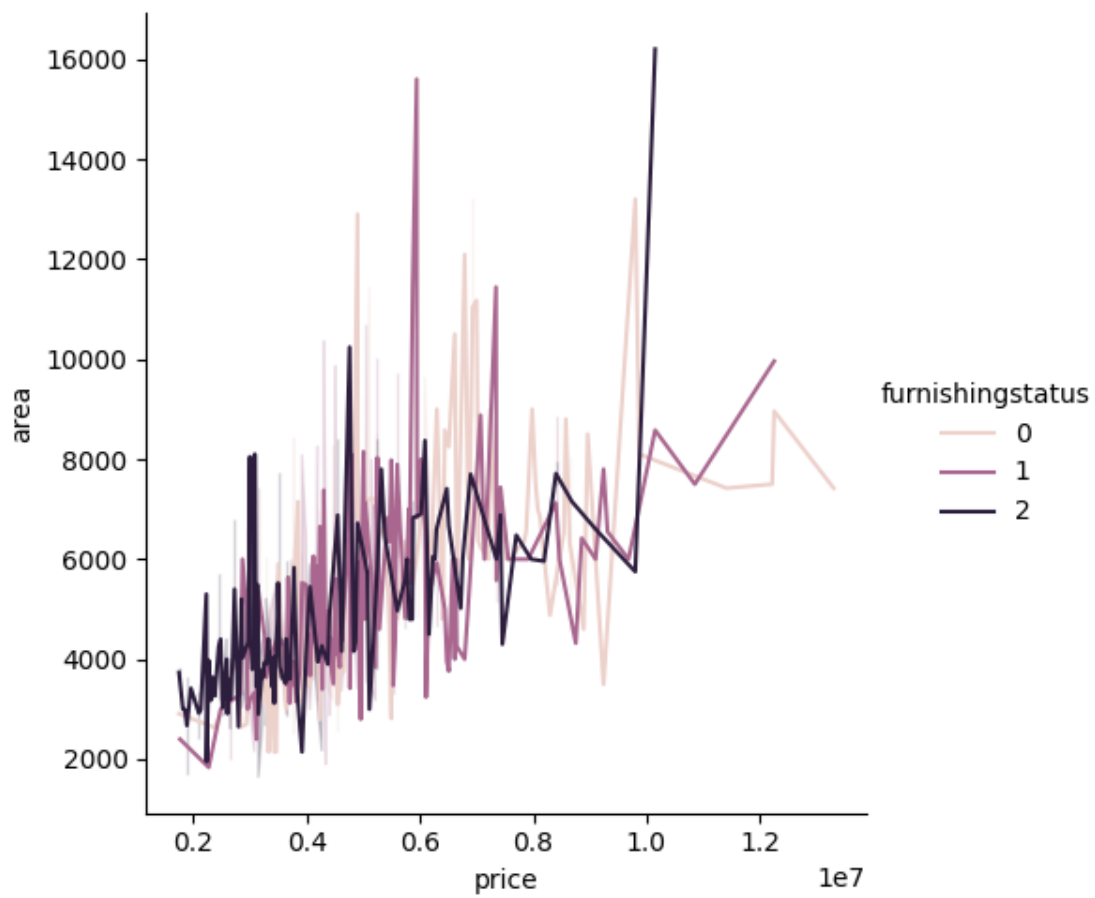
```
Column Name: airconditioning Unique Values: ['yes' 'no']
```

```
Column Name: prefarea Unique Values: ['yes' 'no']
```

```
Column Name: furnishingstatus Unique Values: ['furnished' 'semi-furnished' 'unfurnished']"""
```

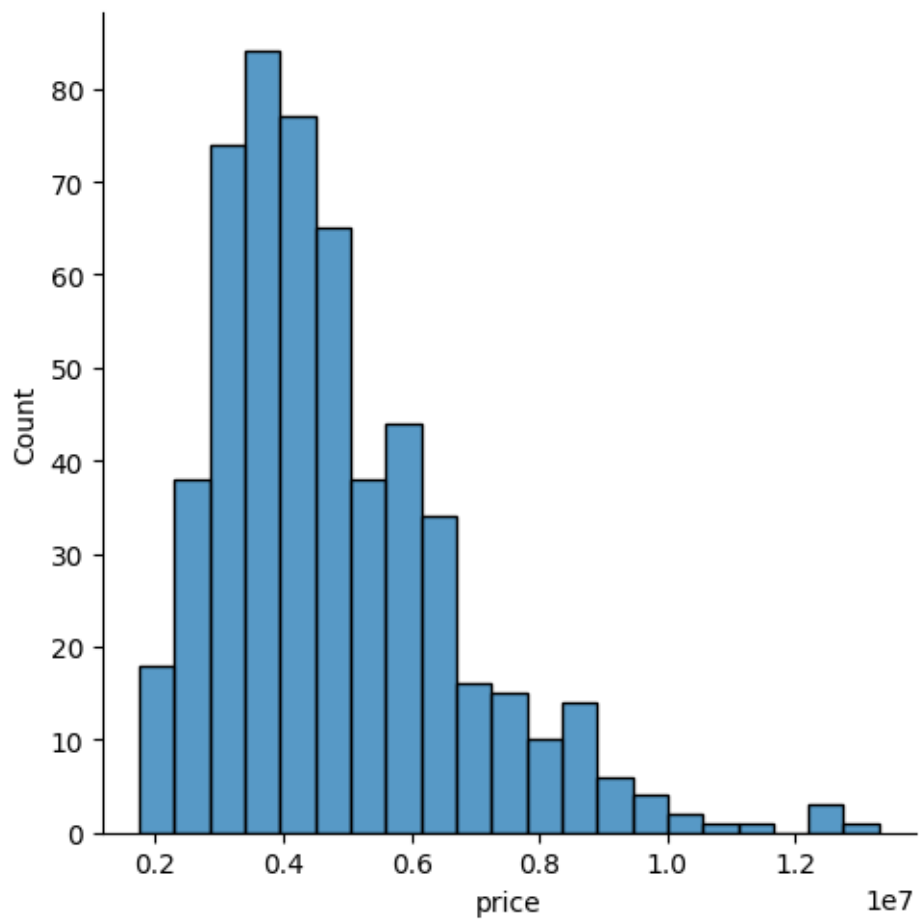
```
[18]: sns.relplot(
      data=df, kind="line",
      x="price", y="area",
      hue="furnishingstatus")
```

```
[18]: <seaborn.axisgrid.FacetGrid at 0x7ed8e8254ac0>
```



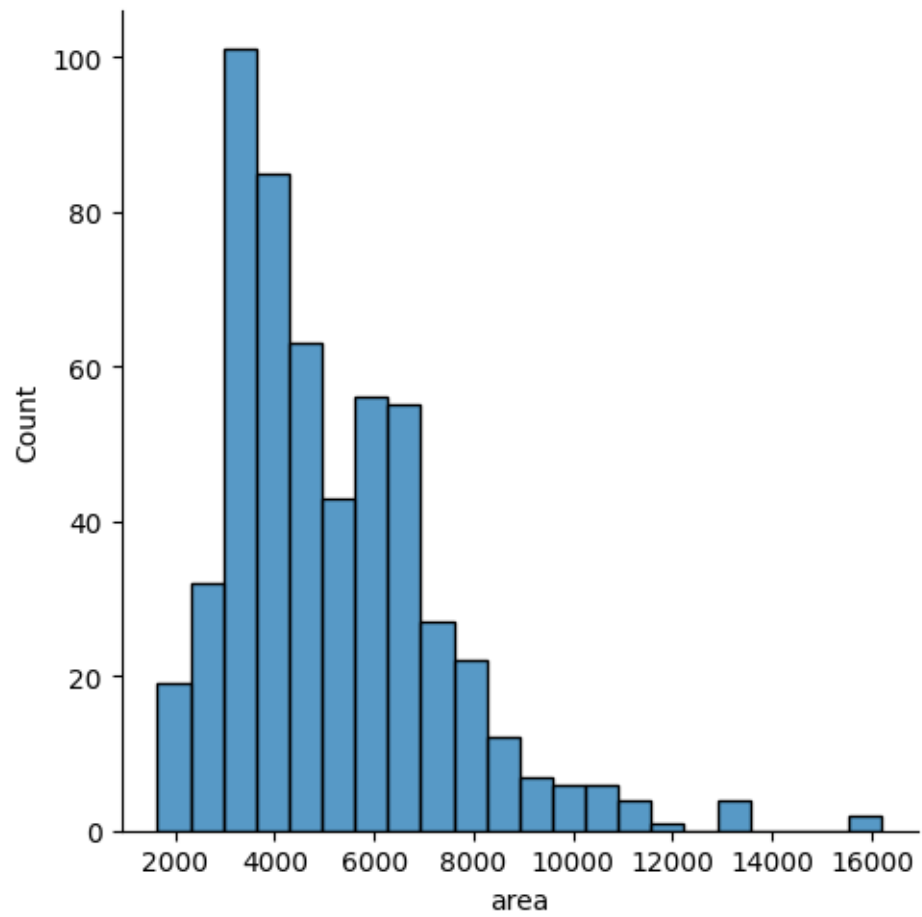
```
[21]: sns.displot(df['price'])
```

```
[21]: <seaborn.axisgrid.FacetGrid at 0x7ed8e5deba00>
```



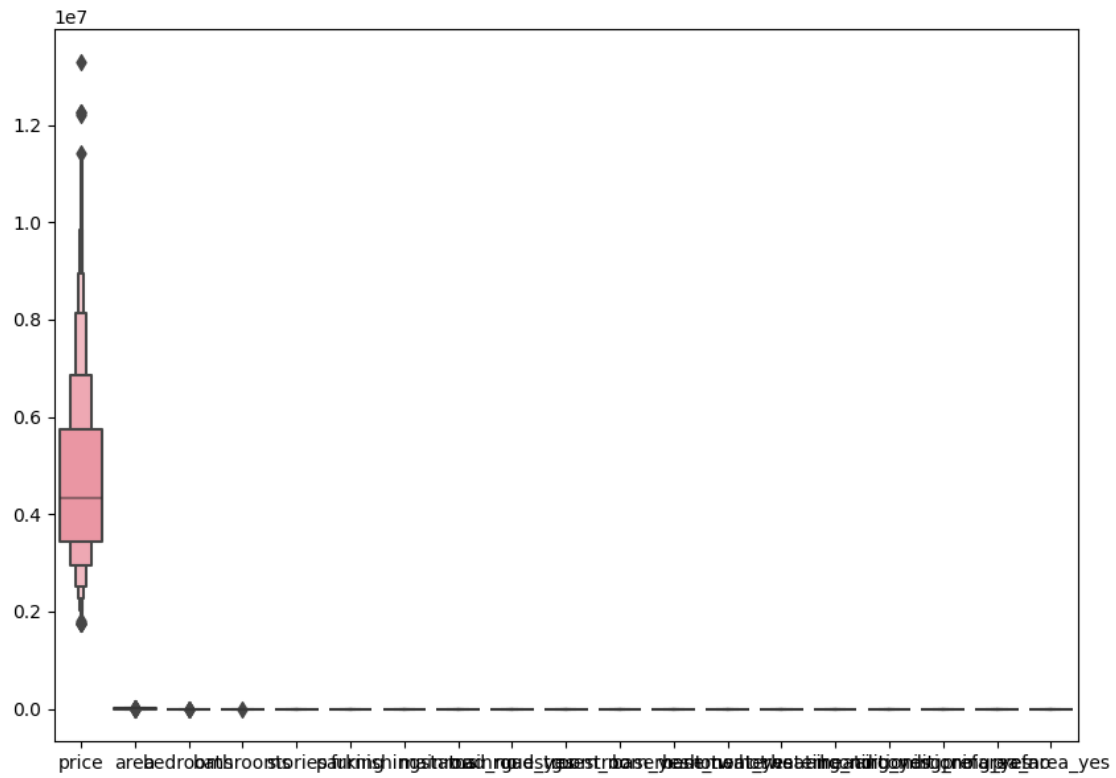
```
[23]: sns.displot(df['area'])
```

```
[23]: <seaborn.axisgrid.FacetGrid at 0x7ed8e80d5db0>
```



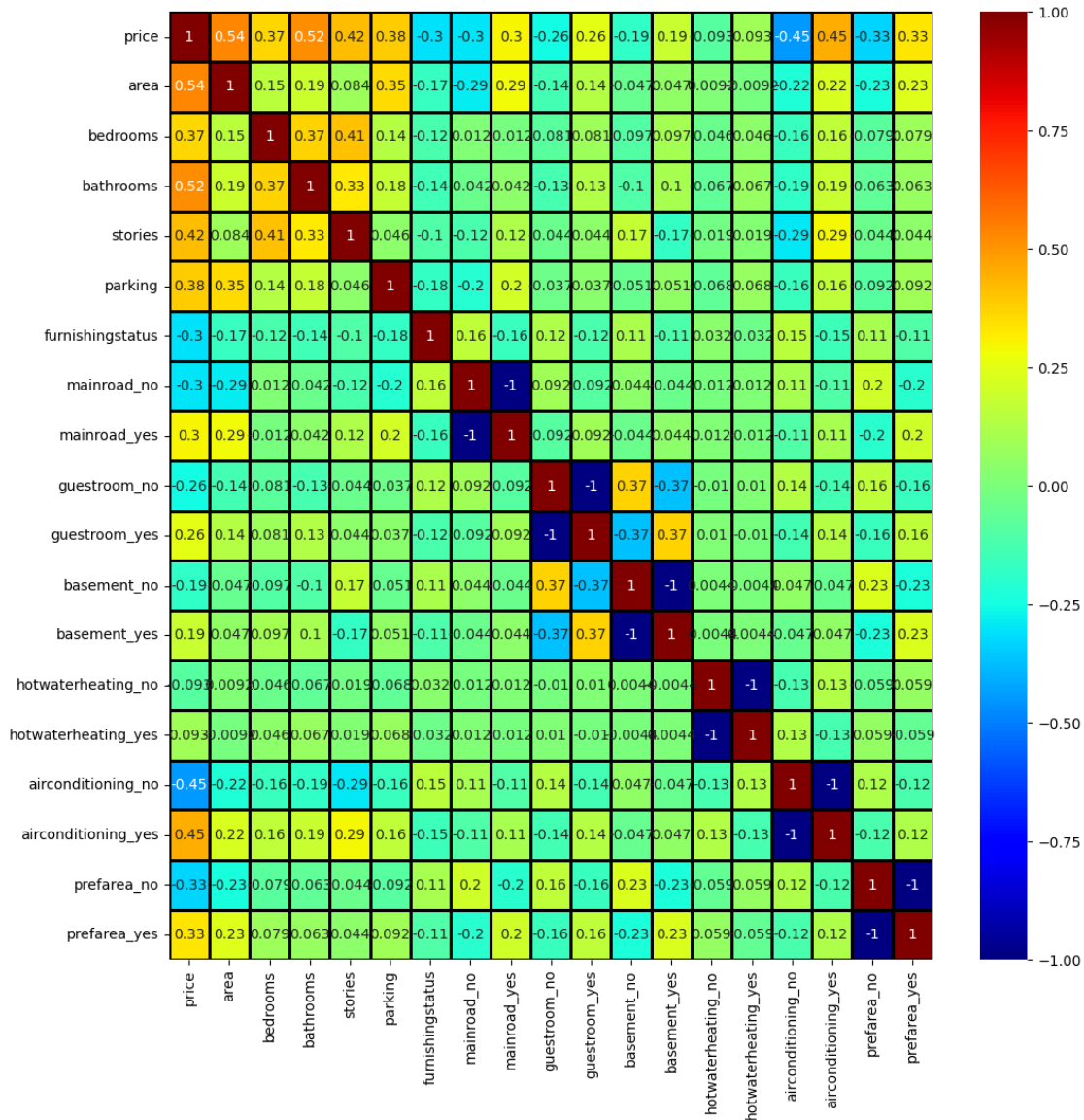
```
[24]: plt.figure(figsize=(10,7))  
sns.boxenplot(df)
```

```
[24]: <Axes: >
```

```
[25]: sns.pairplot(data=df, hue='price')
plt.show()
```

```
[26]: plt.figure(figsize=(12,12))
sns.heatmap(df.corr(), annot=True, linecolor='black', linewidths=1, cmap='jet')
plt.show()
```



```
[27]: #Train&Test Split
```

```
from sklearn.model_selection import train_test_split
```

```
[28]: features =df.drop(['price'],axis=1)
y=df['price']
```

```
[29]: x_train,x_test,y_train,y_test=train_test_split(features,y,train_size=0.8,random_state=889)
```

```
[30]: x_train.shape,y_train.shape,x_test.shape,y_test.shape
```

```
[30]: ((436, 18), (436,), (109, 18), (109,))
```

*Feature Scaling¶

```
[32]: #Scaling
      from sklearn.preprocessing import StandardScaler
```

```
[33]: sc=StandardScaler()
```

```
[34]: x_train=sc.fit_transform(x_train)
      x_test=sc.fit(x_test)
```

```
[37]: from sklearn.linear_model import LinearRegression
      model_lr=LinearRegression()
```

```
[38]: model_lr=LinearRegression()
      model_lr.fit(x_train,y_train)
```

```
[38]: LinearRegression()
```

```
[39]: pred_lr=model_lr.predict(x_train)
```

```
[40]: #Model Evaluation
      from sklearn.metrics import mean_absolute_error,mean_squared_error
```

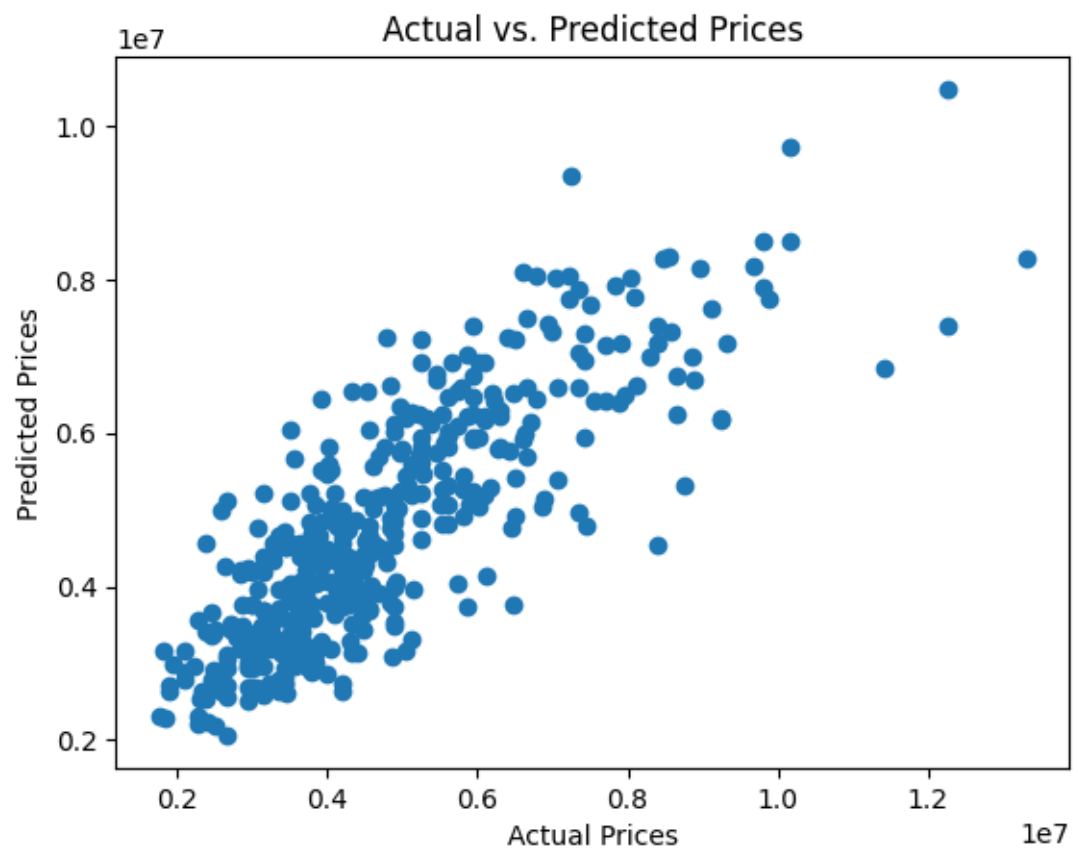
```
[41]: mean_absolute_error(pred_lr,y_train)
```

```
[41]: 774272.115083333
```

```
[42]: mean_squared_error(pred_lr,y_train)
```

```
[42]: 1110157649242.0427
```

```
[46]: plt.scatter(y_train,pred_lr)
      plt.xlabel('Actual Prices')
      plt.ylabel('Predicted Prices')
      plt.title('Actual vs. Predicted Prices')
      plt.show()
```



[]: