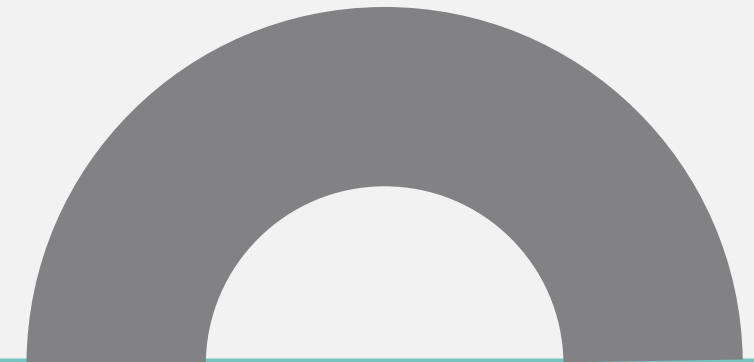


HOUSE PRICE PREDICTION USING MACHINE LEARNING

PHASE 3



INTRODUCTION

Welcome to the revolution of real estate! Discover how **AI-powered house prediction** is transforming the industry. With advanced algorithms and machine learning, we can accurately forecast property values, market trends, and investment opportunities. Join us on this journey to explore the future of real estate.





UNDERSTANDING AI IN REAL ESTATE

Artificial Intelligence (**AI**) is reshaping the real estate landscape. By analyzing vast amounts of data, AI algorithms can predict property prices, identify buyer preferences, and optimize investment decisions. With AI, real estate professionals gain powerful insights, enabling them to make informed choices and provide better services to clients.

program

```
import pandas as pd
```

```
import numpy as np import seaborn as sns
```

```
import matplotlib.pyplot as plt
```

```
from sklearn.model_selection
```

```
import train_test_split
```

```
from sklearn.preprocessing
```

```
import StandardScaler from sklearn.metrics
```

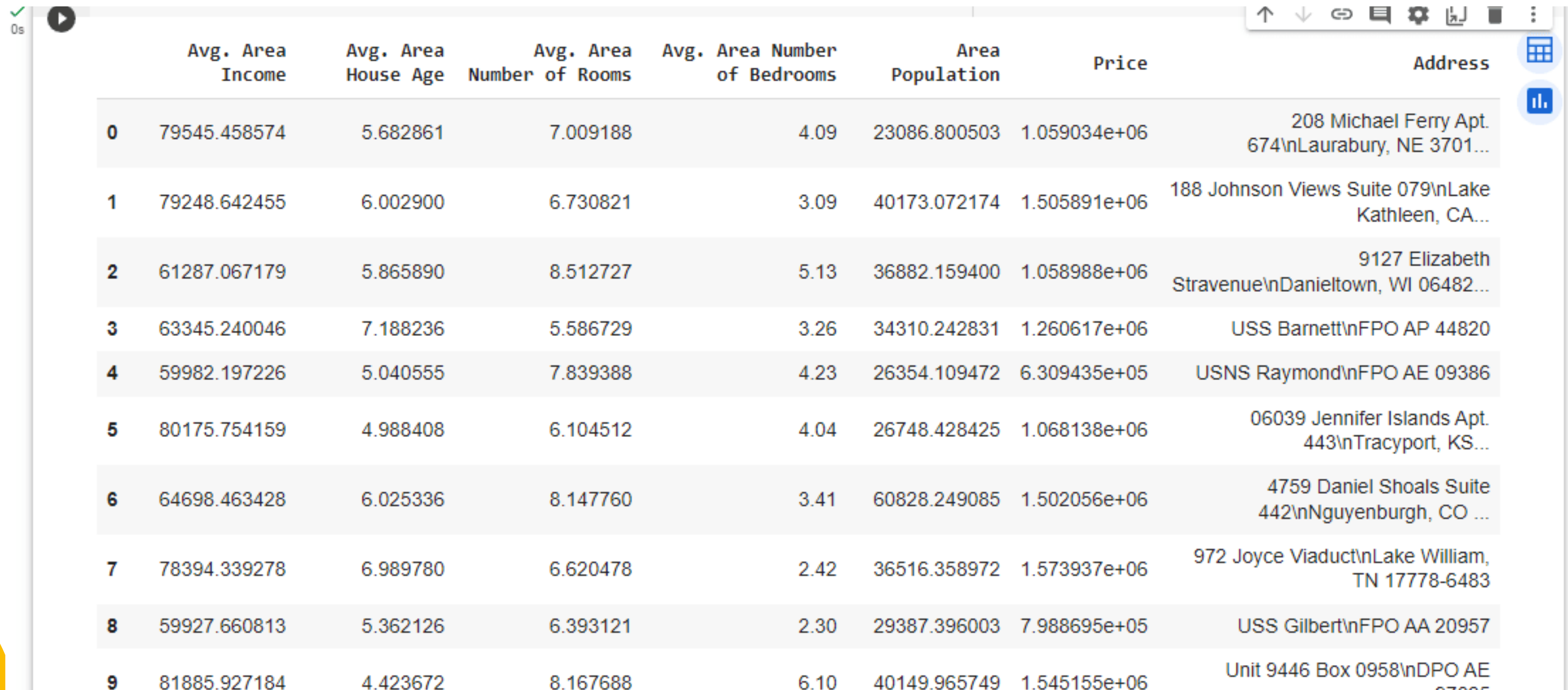
```
import r2_score, mean_absolute_error, mean_square
```

```
d_error
```

- from sklearn.linear_model
- import LinearRegression
- from sklearn.linear_model import Lasso
- from sklearn.ensemble
- import RandomForestRegressor
- from sklearn.svm import SVR import xgboost as

Data set

```
dataset = pd.read_csv('USA_Housing.csv')  
dataset.head(10)
```



	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
0	79545.458574	5.682861	7.009188	4.09	23086.800503	1.059034e+06	208 Michael Ferry Apt. 674\nLaurabury, NE 3701...
1	79248.642455	6.002900	6.730821	3.09	40173.072174	1.505891e+06	188 Johnson Views Suite 079\nLake Kathleen, CA...
2	61287.067179	5.865890	8.512727	5.13	36882.159400	1.058988e+06	9127 Elizabeth Stravenue\nDanieltown, WI 06482...
3	63345.240046	7.188236	5.586729	3.26	34310.242831	1.260617e+06	USS Barnett\nFPO AP 44820
4	59982.197226	5.040555	7.839388	4.23	26354.109472	6.309435e+05	USNS Raymond\nFPO AE 09386
5	80175.754159	4.988408	6.104512	4.04	26748.428425	1.068138e+06	06039 Jennifer Islands Apt. 443\nTracyport, KS...
6	64698.463428	6.025336	8.147760	3.41	60828.249085	1.502056e+06	4759 Daniel Shoals Suite 442\nNguyenburgh, CO ...
7	78394.339278	6.989780	6.620478	2.42	36516.358972	1.573937e+06	972 Joyce Viaduct\nLake William, TN 17778-6483
8	59927.660813	5.362126	6.393121	2.30	29387.396003	7.988695e+05	USS Gilbert\nFPO AA 20957
9	81885.927184	4.423672	8.167688	6.10	40149.965749	1.545155e+06	Unit 9446 Box 0958\nDPO AE 07005

USA_housing.shape

(5000, 7)

USA_housing.info()

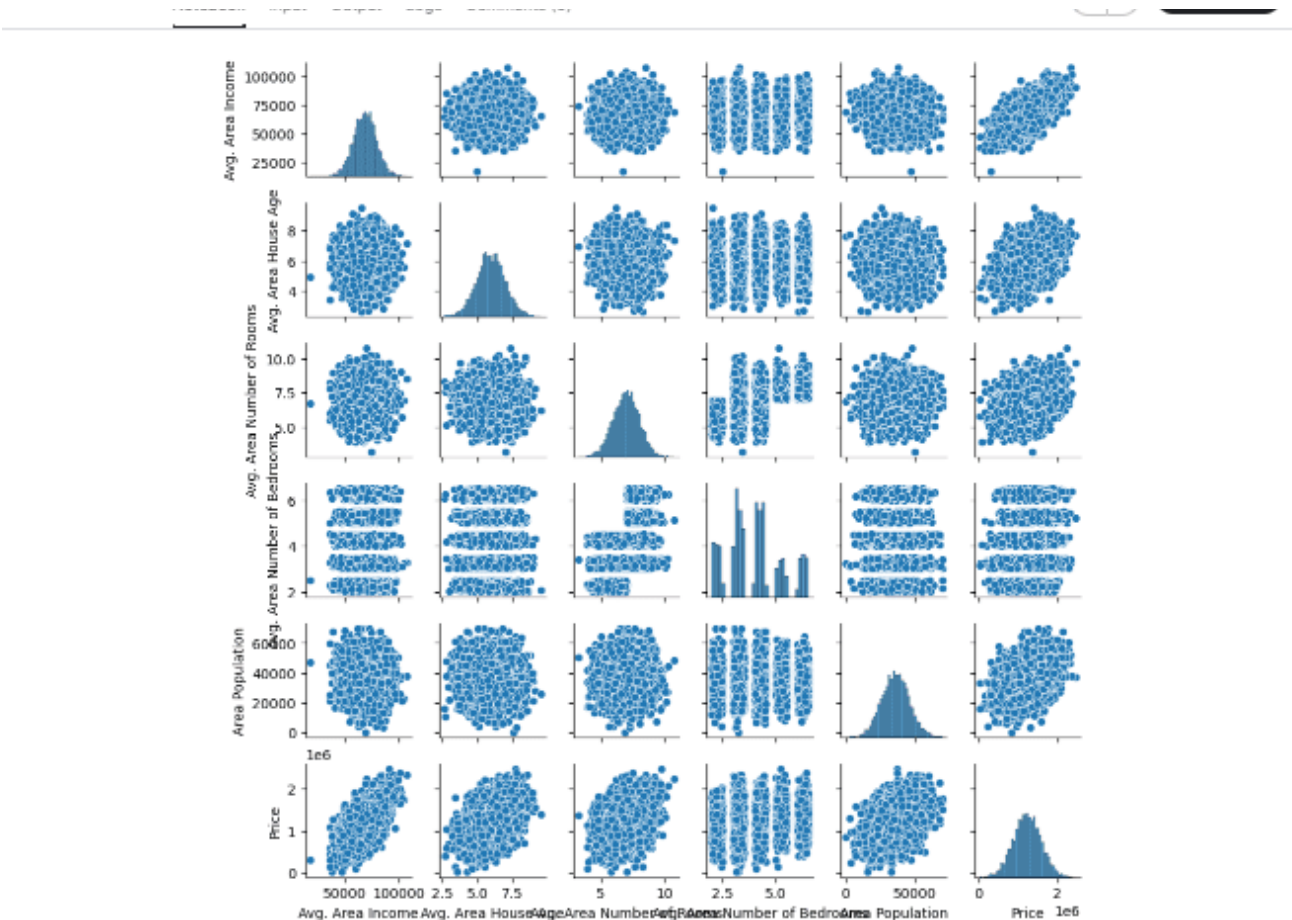
```
Data columns (total 7 columns):  
#   Column                                Non-Null Count  Dtype  
---  -  
0   Avg. Area Income                       5000 non-null  float64  
1   Avg. Area House Age                     5000 non-null  float64  
2   Avg. Area Number of Rooms                5000 non-null  float64  
3   Avg. Area Number of Bedrooms             5000 non-null  float64  
4   Area Population                          5000 non-null  float64  
5   Price                                    5000 non-null  float64  
6   Address                                  5000 non-null  object  
  
dtypes: float64(6), object(1)  
memory usage: 273.6+ KB  
  
In [5]:
```

USA_housing.describe()

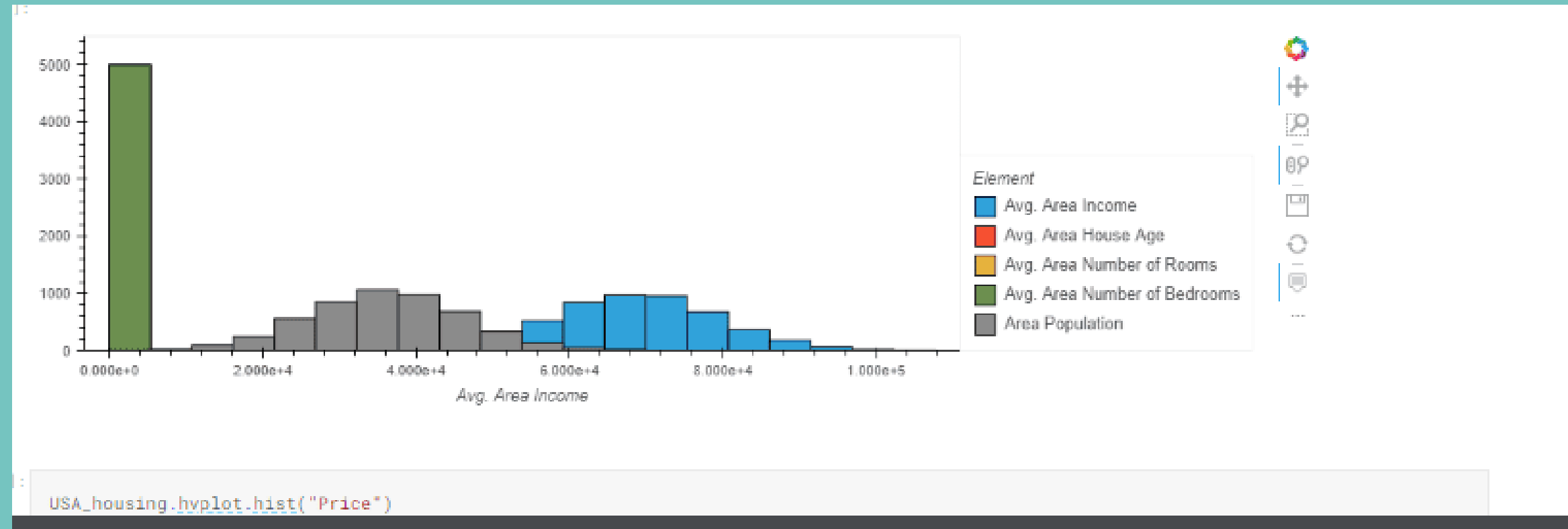
5]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price
count	5000.000000	5000.000000	5000.000000	5000.000000	5000.000000	5.000000e+03
mean	68583.108984	5.977222	6.987792	3.981330	36163.516039	1.232073e+06
std	10657.991214	0.991456	1.005833	1.234137	9925.650114	3.531176e+05
min	17796.631190	2.644304	3.236194	2.000000	172.610686	1.593866e+04
25%	61480.562388	5.322283	6.299250	3.140000	29403.928702	9.975771e+05
50%	68804.286404	5.970429	7.002902	4.050000	36199.406689	1.232669e+06
75%	75783.338666	6.650808	7.665871	4.490000	42861.290769	1.471210e+06
max	107701.748378	9.519088	10.759588	6.500000	69621.713378	2.469066e+06

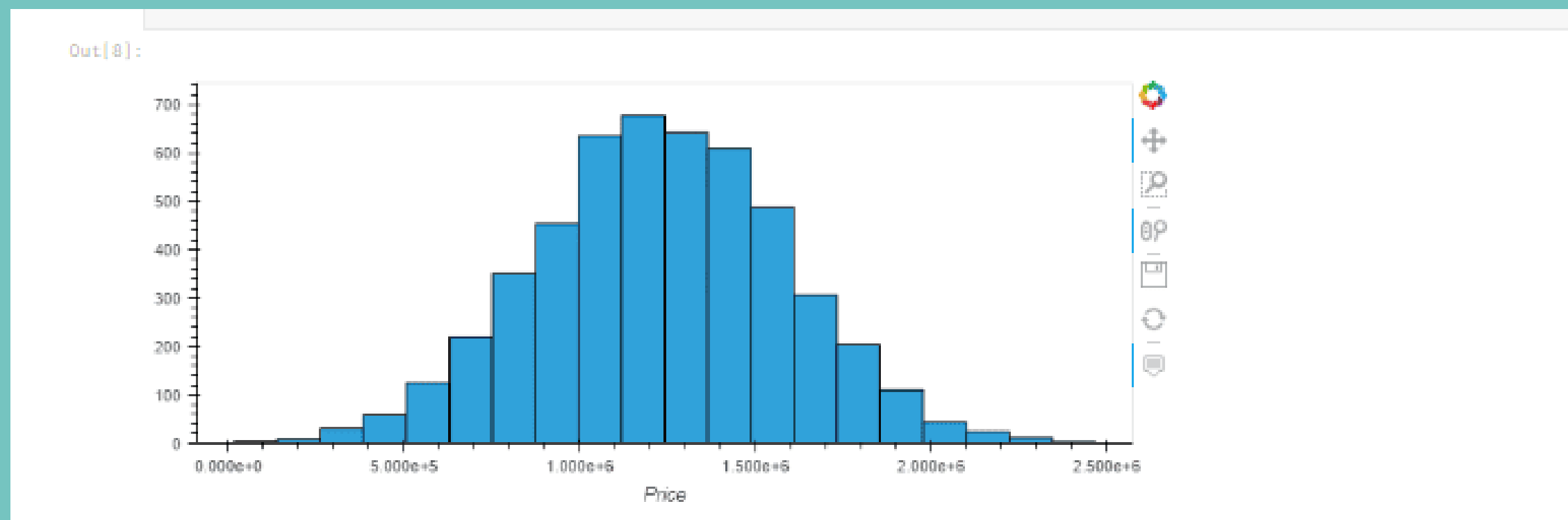
sns.pairplot(USA_housing, height = 1.5)



USA_housing.hvplot.hist(by='Price', subplots=False, width=1000)



USA_housing.hvplot.hist("Price")



CONCLUSION

AI-powered house prediction is transforming real estate by providing accurate insights, improving decision-making, and enhancing customer experiences. While challenges exist, embracing AI technology opens doors to endless possibilities in the industry. Stay ahead of the curve and revolutionize your real estate business with AI-powered solutions.