

```
In [1]: import pandas as pd
```

```
In [2]: df = pd.read_csv('USA_Housing.csv')
print(df)
```

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	\
0	79545.458574	5.682861	7.009188	
1	79248.642455	6.002900	6.730821	
2	61287.067179	5.865890	8.512727	
3	63345.240046	7.188236	5.586729	
4	59982.197226	5.040555	7.839388	
...	...	...	...	
4995	60567.944140	7.830362	6.137356	
4996	78491.275435	6.999135	6.576763	
4997	63390.686886	7.250591	4.805081	
4998	68001.331235	5.534388	7.130144	
4999	65510.581804	5.992305	6.792336	

	Avg. Area Number of Bedrooms	Area Population	Price	\
0	4.09	23086.800503	1.059034e+06	
1	3.09	40173.072174	1.505891e+06	
2	5.13	36882.159400	1.058988e+06	
3	3.26	34310.242831	1.260617e+06	
4	4.23	26354.109472	6.309435e+05	
...	...	...	...	
4995	3.46	22837.361035	1.060194e+06	
4996	4.02	25616.115489	1.482618e+06	
4997	2.13	33266.145490	1.030730e+06	
4998	5.44	42625.620156	1.198657e+06	
4999	4.07	46501.283803	1.298950e+06	

	Address
0	208 Michael Ferry Apt. 674\nLaurabury, NE 3701...
1	188 Johnson Views Suite 079\nLake Kathleen, CA...
2	9127 Elizabeth Stravenue\nDanielstown, WI 06482...
3	USS Barnett\nFPO AP 44820
4	USNS Raymond\nFPO AE 09386
...	...
4995	USNS Williams\nFPO AP 30153-7653
4996	PSC 9258, Box 8489\nAPO AA 42991-3352
4997	4215 Tracy Garden Suite 076\nJoshualand, VA 01...
4998	USS Wallace\nFPO AE 73316
4999	37778 George Ridges Apt. 509\nEast Holly, NV 2...

[5000 rows x 7 columns]

```
In [3]: print(df.info())
print(df.describe())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5000 entries, 0 to 4999
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
None
```

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms \
count	5000.000000	5000.000000	5000.000000
mean	68583.108984	5.977222	6.987792
std	10657.991214	0.991456	1.005833
min	17796.631190	2.644304	3.236194
25%	61480.562388	5.322283	6.299250
50%	68804.286404	5.970429	7.002902
75%	75783.338666	6.650808	7.665871
max	107701.748378	9.519088	10.759588

	Avg. Area Number of Bedrooms	Area Population	Price
count	5000.000000	5000.000000	5.000000e+03
mean	3.981330	36163.516039	1.232073e+06
std	1.234137	9925.650114	3.531176e+05
min	2.000000	172.610686	1.593866e+04
25%	3.140000	29403.928702	9.975771e+05
50%	4.050000	36199.406689	1.232669e+06
75%	4.490000	42861.290769	1.471210e+06
max	6.500000	69621.713378	2.469066e+06

```
In [4]: X = df.drop(['Price', 'Address'], axis=1)
y = df['Price']
```

```
In [5]: from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
In [6]: from sklearn.linear_model import LinearRegression
```

```
model = LinearRegression()
model.fit(X_train, y_train)
```

Out[6]:

```
LinearRegression()
LinearRegression()
```

```
In [7]: from sklearn.metrics import mean_squared_error, r2_score
import numpy as np
```

```
y_pred = model.predict(X_test)

mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)

absolute_errors = abs(y_pred - y_test)
percentage_errors = (absolute_errors / y_test) * 100

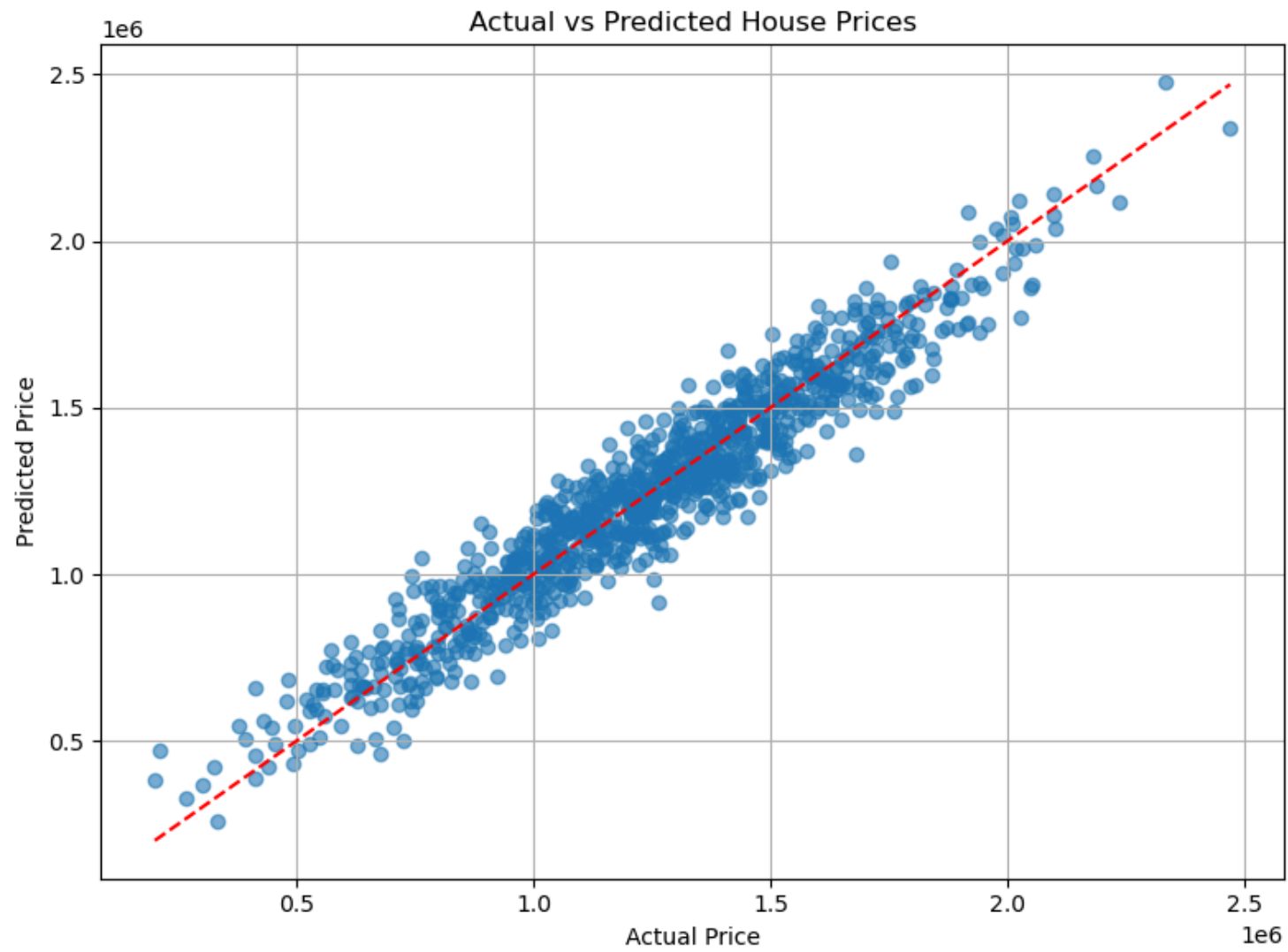
mean_absolute_error = np.mean(absolute_errors)
mean_percentage_error = np.mean(percentage_errors)

print(f"Mean Squared Error      : {mse:.4f}")
print(f"Root Mean Squared Error  : {rmse:.4f}")
print(f"Mean Absolute Error        : {mean_absolute_error:.2f}")
print(f"Mean Percentage Error       : {mean_percentage_error:.2f}%")
print(f"R2 Score (Overall Fit)     : {r2_score(y_test, y_pred):.4f}")
```

```
Mean Squared Error      : 10089009300.8929
Root Mean Squared Error  : 100444.0606
Mean Absolute Error      : 80879.10
Mean Percentage Error     : 7.39%
R2 Score (Overall Fit)   : 0.9180
```

```
In [11]: import matplotlib.pyplot as plt
```

```
plt.figure(figsize=(8, 6))
plt.scatter(y_test, y_pred, alpha=0.6)
plt.plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--')
plt.xlabel("Actual Price")
plt.ylabel("Predicted Price")
plt.title("Actual vs Predicted House Prices")
plt.grid(True)
plt.tight_layout()
plt.show()
```



```
In [12]: index = 0

sample_input = X_test.iloc[[index]]
actual_price = y_test.iloc[index]
predicted_price = model.predict(sample_input)[0]

print(f"Actual Price      : ${actual_price:,.2f}")
print(f"Predicted Price    : ${predicted_price:,.2f}")
print(f"Absolute Error     : ${abs(predicted_price - actual_price):,.2f}")
print(f"Percentage Error      : {abs(predicted_price - actual_price) / actual_price * 100:,.2f}%")
```

Actual Price : \$1,339,096.08  
Predicted Price : \$1,308,587.93  
Absolute Error : \$30,508.15  
Percentage Error : 2.28%

In [ ]: