7/20/24, 5:43 PM task2.ipynb - Colab

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## TrackCode:ML

## Task2:Predict The Prices Of House

1 import pandas as pd

 $1\ df=pd.read\_csv("\underline{https://github.com/YBI-Foundation/Dataset/raw/main/House%20Prices.csv")$ 

<sup>2</sup> df

0       1       20140916T000000       280000.0       6       3.00       2400       9373         1       2       20150422T000000       300000.0       6       3.00       2400       9373         2       3       20140508T000000       647500.0       4       1.75       2060       26036         3       4       20140811T000000       235000.0       3       1.00       1460       43000         4       5       20150401T000000       235000.0       3       1.00       1430       7599                   21608       21609       20140725T000000       365000.0       5       2.00       1600       4168         21609       21610       20150311T000000       380000.0       2       1.00       1040       7372         21610       21611       20140624T00000       339000.0       3       1.00       1100       4128         21612       21613       20141030T000000       268950.0       3       1.00       1320       8100         21613 rows × 21 columns	<b>→</b>		ID	Date	Price	Bedrooms	Bathrooms	Sqft_living	Sqft_lot	F				
2 3 20140508T000000 647500.0 4 1.75 2060 26036 3 4 20140811T000000 400000.0 3 1.00 1460 43000 4 5 20150401T000000 235000.0 3 1.00 1430 7599		0	1	20140916T000000	280000.0	6	3.00	2400	9373					
3 4 20140811T000000 400000.0 3 1.00 1460 43000 4 5 20150401T000000 235000.0 3 1.00 1430 7599		1	2	20150422T000000	300000.0	6	3.00	2400	9373					
4 5 20150401T000000 235000.0 3 1.00 1430 7599		2	3	20140508T000000	647500.0	4	1.75	2060	26036					
21608 21609 20140725T000000 365000.0 5 2.00 1600 4168 21609 21610 20150311T000000 380000.0 2 1.00 1040 7372 21610 21611 20140624T000000 339000.0 3 1.00 1100 4128 21611 21612 20140703T000000 399900.0 2 1.75 1410 1005 21612 21613 20141030T000000 268950.0 3 1.00 1320 8100 21613 rows × 21 columns   Distributions		3	4	20140811T000000	400000.0	3	1.00	1460	43000					
21608 21609 20140725T000000 365000.0 5 2.00 1600 4168 21609 21610 20150311T000000 380000.0 2 1.00 1040 7372 21610 21611 20140624T000000 339000.0 3 1.00 1100 4128 21611 21612 20140703T000000 39990.0 2 1.75 1410 1005 21612 21613 20141030T000000 268950.0 3 1.00 1320 8100 21613 rows × 21 columns   Distributions		4	5	20150401T000000	235000.0	3	1.00	1430	7599					
21610 21611 20140624T000000 339000.0 2 1.00 1040 7372  21610 21611 20140624T000000 339000.0 3 1.00 1100 4128  21611 21612 20140703T000000 268950.0 3 1.00 1320 8100  21613 rows × 21 columns  Distributions														
21610 21611 20140624T000000 339000.0 3 1.00 1100 4128  21611 21612 20140703T000000 399900.0 2 1.75 1410 1005  21612 21613 20141030T000000 268950.0 3 1.00 1320 8100  21613 rows × 21 columns  Distributions		21608	21609	20140725T000000	365000.0	5	2.00	1600	4168					
21611 21612 20140703T000000 399900.0 2 1.75 1410 1005 21612 21613 20141030T000000 268950.0 3 1.00 1320 8100  21613 rows × 21 columns  Distributions		21609	21610	20150311T000000	380000.0	2	1.00	1040	7372					
21612 21613 20141030T000000 268950.0 3 1.00 1320 8100  21613 rows × 21 columns  Distributions		21610	21611	20140624T000000	339000.0	3	1.00	1100	4128					
21613 rows × 21 columns  Distributions  4		21611	21612	20140703T000000	399900.0	2	1.75	1410	1005					
Distributions  2-d distributions		21612	21613	20141030T000000	268950.0	3	1.00	1320	8100					
2-d distributions		21613 rows × 21 columns												
20 20 20 20 20 20 20 20 20 20 20 20 20 2		1000 -	tions  o  see see see see see see see see see se	1000- 000- 000- 000- 000- 000- 200- 10093 2000 i	Pice	12000 000 000 000 000 000 000 000 000 00	Bedroo	2005 27 23 28						
2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2		5 6 7 8												
2000-		7 - 6 - 5 - 6 - 6 - 6 - 6 - 6 - 6 - 6 - 6	ob 1986	22 - 22 - 22 - 22 - 22 - 22 - 22 - 22		to the state of th		, 35 / 35 - 35 -						

## 1 df.head()

₹		ID	Date	Price	Bedrooms	Bathrooms	Sqft_living	Sqft_lot	Floors	Waterfront	View	 Grade	Sqft_above	Sqft_base
	0	1	20140916T000000	280000.0	6	3.00	2400	9373	2.0	0	0	 7	2400	
	1	2	20150422T000000	300000.0	6	3.00	2400	9373	2.0	0	0	 7	2400	

<sup>2</sup> import numpy as np

**2** 3 20140508T000000 647500.0 1.75 2060 26036 1.0 0 1160 0 8 1 df.info() <<class 'pandas.core.frame.DataFrame'> RangeIndex: 21613 entries, 0 to 21612 Data columns (total 21 columns): # Column Non-Null Count Dtype --------0 ID 21613 non-null int64 Date 21613 non-null object 1 2 Price 21613 non-null float64 3 Bedrooms 21613 non-null Bathrooms 21613 non-null float64 5 Sqft\_living 21613 non-null int64 6 Sqft\_lot 21613 non-null int64 21613 non-null Floors float64 8 Waterfront 21613 non-null int64 9 View 21613 non-null int64 10 Condition 21613 non-null int64 Grade 21613 non-null 11 int64 Sqft\_above 21613 non-null int64 12 13 Sqft\_basement 21613 non-null int64 14 Yr\_built 21613 non-null int64 15 Yr\_renovated 21613 non-null int64 16 zipcode 21613 non-null int64 17 21613 non-null float64 Lat 18 Long 21613 non-null float64 19 Sqft\_living15 21613 non-null int64 20 Sqft\_lot15 21613 non-null int64 dtypes: float64(5), int64(15), object(1) memory usage: 3.5+ MB 1 df.describe() ₹ ID Price Bedrooms Bathrooms Sqft\_living Sqft\_lot Floors Waterfront View Condi **count** 21613.00000 2.161300e+04 21613.000000 21613.000000 21613.000000 2.161300e+04 21613.000000 21613.000000 21613.000000 21613.00 mean 10807.00000 5.401822e+05 3.370842 2.114757 2079.899736 1.510697e+04 1.494309 0.007542 0.234303 3.40 std 6239.28002 3.673622e+05 0.930062 0.770163 918.440897 4.142051e+04 0.539989 0.086517 0.766318 0.65 min 1.00000 7.500000e+04 0.000000 0.000000 290.000000 5.200000e+02 1.000000 0.000000 0.000000 1.00 0.000000 0.000000 25% 5404.00000 3.219500e+05 3.000000 1.750000 1427.000000 5.040000e+03 1.000000 3.00 50% 10807.00000 4.500000e+05 3.000000 2.250000 1910.000000 7.618000e+03 1.500000 0.000000 0.000000 3.00 75% 16210.00000 6.450000e+05 4.000000 2.500000 2550.000000 1.068800e+04 2.000000 0.000000 0.000000 4.00 21613.00000 7.700000e+06 33.000000 8.000000 13540.000000 1.651359e+06 3.500000 1.000000 4.000000 max 5.00 1 df[['Bedrooms']].value\_counts() 2 **→** Bedrooms 3 9824 6882 4 2 2760 5 1601 272 6 1 199 7 38 0 13 8 13 9 6 10 3 11 1 33 1 Name: count, dtype: int64 1 df[['Bathrooms']].value\_counts()  $\rightarrow$ Bathrooms 2.50 5380 1.00 3852

1.75

2.25

3048 2047

1930

```
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        1.50
        2.75
        3.00
        3.50
        3.25
        3.75
        4.00
        4.50
```

4.25

0.75

4.75

5.00

5.25

5.50

0.00

1.25

6.00

0.50

5.75

6.25 6.50 6.75

8.00 7.50

7.75

→ Sqft\_living 1300

1400

1440

1660

1010

2456

2473

2478

2481

1 df.columns

1 df.shape → (21613, 21)

1 y.shape **→** (21613,)

2

3 4

21608

21609

21610

21611

21612

1 y 0

1 y=df['Price']

280000.0 300000.0

647500.0 400000.0

235000.0

365000.0

380000.0

339000.0

399900.0

268950.0

Name: Price, Length: 21613, dtype: float64

13540

1446

1185

753

731

589

155

136

100

79

72

23

21

13

10

10

9

6

4

4

2

1

1 Name: count, dtype: int64

138

135

133

129

129

1

1

1

1

1 Name: count, Length: 1038, dtype: int64

1 df[['Sqft\_living']].value\_counts()

```
1 x=df[['Bedrooms', 'Bathrooms', 'Sqft_living',
2 'Sqft_lot','Sqft_living15', 'Sqft_lot15']]
```

1 x.shape

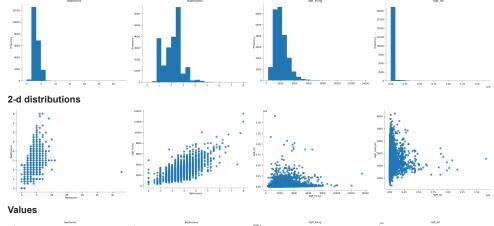
→ (21613, 6)

1 x

	Bedrooms	Bathrooms	Sqft_living	Sqft_lot	Sqft_living15	Sqft_lot15
0	6	3.00	2400	9373	2060	7316
1	6	3.00	2400	9373	2060	7316
2	4	1.75	2060	26036	2590	21891
3	3	1.00	1460	43000	2250	20023
4	3	1.00	1430	7599	1290	10320
21608	5	2.00	1600	4168	1190	4168
21609	2	1.00	1040	7372	1930	5150
21610	3	1.00	1100	4128	1510	4538
21611	2	1.75	1410	1005	1440	1188
21612	3	1.00	1320	8100	1000	8100

21613 rows × 6 columns





1 from sklearn.model\_selection import train\_test\_split

1 x\_train,x\_test,y\_train,y\_test=train\_test\_split(x,y, test\_size=0.3,random\_state=222529)

1 x\_train.shape,x\_test.shape,y\_train.shape,y\_test.shape

((15129, 6), (6484, 6), (15129,), (6484,))

1 from sklearn.linear\_model import LinearRegression

1 lr=LinearRegression()

1 lr.fit(x\_train,y\_train)

```
LinearRegression
```

1 y\_pred=lr.predict(x\_test)

1 y\_pred.shape

→ (6484,)

1 y\_pred

array([497844.51474271, 307193.5024687, 440275.88332779, ..., 501885.22737581, 281005.21986967, 742647.79487191])

1 from sklearn.metrics import mean\_squared\_error,mean\_absolute\_error,r2\_score

1 mean\_squared\_error(y\_test,y\_pred)

→ 69374029792.9394

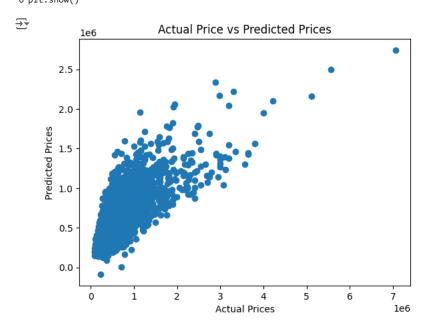
1 mean\_absolute\_error(y\_test,y\_pred)

**170909.97763844364** 

1 r2\_score(y\_test,y\_pred)

→ 0.5214172896943516

```
1 import matplotlib.pyplot as plt
2 plt.scatter(y_test,y_pred)
3 plt.xlabel("Actual Prices")
4 plt.ylabel("Predicted Prices")
5 plt.title("Actual Price vs Predicted Prices")
6 plt.show()
```



1 df\_new=df.sample(1)

1 df\_new

TID Date Price Bedrooms Bathrooms Sqft\_living Sqft\_lot Floors Waterfront View ... Grade Sqft\_above Sqft\_1 7671 7672 20141119T000000 1010000.0 4 3.5 3350 3752 2.0 0 0 0 ... 9 2550 1 rows × 21 columns

1 df\_new.shape

```
→ (1, 21)
```