

Bike Price Prediction using Linear Regression

Importing the libraries

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

Importing the dataset

```
df =
pd.read_csv('https://github.com/YBI-Foundation/Dataset/raw/main/Bike
%20Prices.csv')
```

Get Missing Values Drop

```
df=df.dropna()
```

Get Information of Dataframe

```
df.info()    #gives column name, count, not null category, D-type(data
type)
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 626 entries, 0 to 625
Data columns (total 8 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   Brand                 626 non-null   object
 1   Model                 626 non-null   object
 2   Selling_Price         626 non-null   int64
 3   Year                  626 non-null   int64
 4   Seller_Type           626 non-null   object
 5   Owner                 626 non-null   object
 6   KM_Driven             626 non-null   int64
 7   Ex_Showroom_Price     626 non-null   float64
dtypes: float64(1), int64(3), object(4)
memory usage: 44.0+ KB
```

```
df.describe()    #gives the linear relation of each column with another
column
```

	Selling_Price	Year	KM_Driven	Ex_Showroom_Price
count	626.000000	626.000000	626.000000	6.260000e+02
mean	59445.164537	2014.800319	32671.576677	8.795871e+04
std	59904.350888	3.018885	45479.661039	7.749659e+04

```

min      6000.000000  2001.000000    380.000000    3.049000e+04
25%     30000.000000  2013.000000   13031.250000    5.485200e+04
50%     45000.000000  2015.000000   25000.000000    7.275250e+04
75%     65000.000000  2017.000000   40000.000000    8.703150e+04
max     760000.000000  2020.000000  585659.000000   1.278000e+06

```

```
df.head()
```

```

      Brand      Model  Selling_Price  Year  Seller_Type
Owner \
0  TVS      TVS XL 100      30000    2017  Individual  1st
owner
1  Bajaj    Bajaj ct 100      18000    2017  Individual  1st
owner
2  Yo      Yo Style      20000    2011  Individual  1st
owner
3  Bajaj    Bajaj Discover 100      25000    2010  Individual  1st
owner
4  Bajaj    Bajaj Discover 100      24999    2012  Individual  2nd
owner

```

```

      KM_Driven  Ex_Showroom_Price
0         8000         30490.0
1        35000         32000.0
2        10000         37675.0
3        43000         42859.0
4        35000         42859.0

```

```

df.isnull().sum()  #(df.isna()).sum() gives same result)
#gives the sum of all null values columns-wise


```

```

Brand      0
Model      0
Selling_Price  0
Year      0
Seller_Type  0
Owner      0
KM_Driven  0
Ex_Showroom_Price  0
dtype: int64

```

```
df.nunique()  #gives total no. of unique entries
```

```

Brand      18
Model     183
Selling_Price  99
Year      18
Seller_Type   2
Owner       4
KM_Driven  219

```

```
Ex_Showroom_Price    230
dtype: int64
```

```
df.shape
```

```
(626, 8)
```

```
#Get Categories and Counts of Categorical Variables
```

```
df[['Brand']].value_counts()
```

```
Brand
Honda      170
Bajaj      143
Hero       108
Yamaha      94
Royal       40
TVS         23
Suzuki      18
KTM         6
Mahindra    6
Kawasaki    4
UM          3
Activa      3
Harley      2
Vespa       2
BMW         1
Hyosung     1
Benelli     1
Yo          1
dtype: int64
```

```
df[['Model']].value_counts()
```

```
Model
Honda Activa [2000-2015]      23
Honda CB Hornet 160R         22
Bajaj Pulsar 180             20
Yamaha FZ S V 2.0            16
Bajaj Discover 125           16
..
Royal Enfield Thunderbird 500 1
Royal Enfield Continental GT [2013 - 2018] 1
Royal Enfield Classic Stealth Black 1
Royal Enfield Classic Squadron Blue 1
Yo Style                     1
Length: 183, dtype: int64
```

```
df[['Seller_Type']].value_counts()
```

```
Seller_Type
Individual    623
```

```

Dealer          3
dtype: int64

df[['Owner']].value_counts()

Owner
1st owner    556
2nd owner     66
3rd owner      3
4th owner      1
dtype: int64

```

Get Column Names

```

df.columns

Index(['Brand', 'Model', 'Selling_Price', 'Year', 'Seller_Type',
      'Owner',
      'KM_Driven', 'Ex_Showroom_Price'],
      dtype='object')

```

Get Encoding of Categorical Features

```

df.replace({'Seller_Type':{'Individual':0,'Dealer':1}},inplace=True)

df.replace({'Owner':{'1st owner':0,'2nd owner':1,'3rd owner':2,'4th
owner':3}},inplace=True)

#X=pd.get_dummies(X,columns=['Seller_Type','Owner'],drop_first=True)

```

```

df

   Brand      Model  Selling_Price  Year \
0    TVS  TVS XL 100         30000  2017
1  Bajaj  Bajaj ct 100         18000  2017
2    Yo  Yo Style         20000  2011
3  Bajaj  Bajaj Discover 100         25000  2010
4  Bajaj  Bajaj Discover 100         24999  2012
..    ...
621  Harley  Harley-Davidson Street 750      330000  2014
622  Kawasaki  Kawasaki Ninja 650 [2018-2019]  300000  2011
623  Kawasaki  Kawasaki Ninja 650 [2018-2019]  425000  2017
624  Suzuki  Suzuki GSX S750      760000  2019
625  Harley  Harley-Davidson Street Bob      750000  2013

   Seller_Type  Owner  KM_Driven  Ex_Showroom_Price
0             0      0      8000         30490.0
1             0      0     35000         32000.0
2             0      0     10000         37675.0
3             0      0     43000         42859.0
4             0      1     35000         42859.0

```

621
	0	3	6500	534000.0
622	0	0	12000	589000.0
623	0	1	13600	599000.0
624	0	0	2800	752020.0
625	0	1	12000	1278000.0

[626 rows x 8 columns]

Define y

```
y=df['Selling_Price']
```

```
y.shape
```

(626,)

```
y
```

0	30000
1	18000
2	20000
3	25000
4	24999

	...
621	330000
622	300000
623	425000
624	760000
625	750000

Name: Selling_Price, Length: 626, dtype: int64

```
X=df[['Year','Seller_Type','Owner','KM_Driven','Ex_Showroom_Price']]
```

```
X.shape
```

(626, 5)

```
X
```

	Year	Seller_Type	Owner	KM_Driven	Ex_Showroom_Price
0	2017	0	0	8000	30490.0
1	2017	0	0	35000	32000.0
2	2011	0	0	10000	37675.0
3	2010	0	0	43000	42859.0
4	2012	0	1	35000	42859.0
...
621	2014	0	3	6500	534000.0
622	2011	0	0	12000	589000.0
623	2017	0	1	13600	599000.0
624	2019	0	0	2800	752020.0
625	2013	0	1	12000	1278000.0

[626 rows x 5 columns]

Train Test Split

```
from sklearn.model_selection import train_test_split

from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=
train_test_split(X,y,test_size=0.3,random_state=2323)

X_train.shape, X_test.shape, y_train.shape, y_test.shape

((438, 5), (188, 5), (438,), (188,))
```

Get Model Test

```
from sklearn.linear_model import LinearRegression

lr= LinearRegression()

lr.fit(X_train, y_train)

LinearRegression()
```

Get Model Prediction

```
y_pred= lr.predict(X_test)

y_pred.shape

(188,)

y_pred

array([ 35870.79961642,  65547.96126415, 12371.09961996,
 64148.8895839 ,
        7062.69388653,  54402.63506136, -1912.44035159,
29775.20161928,
        14243.08480753,  22080.83178243,  84025.43006156,
38292.33063928,
        29354.69337304,  21525.66151869,  51718.07386067,
133108.79218061,
        36717.94936617,  61355.37093641,  28387.67135344,
43699.33828063,
        73829.7721122 , 121281.37404139, 257025.1574888 , -
13165.16881032,
        46855.75420218,  52525.81083583,  21528.77516102,
58963.76859935,
        36453.23827565,  59092.23873301,  52537.87619985,
31275.42528094,
```

50620.8218115 , 67631.38902032, 56630.86444395,
117293.32518539,
34037.48760149, 49778.37056618, 111705.80584799,
21536.11351636,
44867.13135661, 74909.17128765, 421105.83467059,
73670.58645457,
51287.25308003, 14423.62915765, 99846.44490246,
54487.16853425,
101946.69838917, 79576.38985325, 117362.89863505,
37565.98215828,
68125.63747921, 35121.70517691, 52265.51803945,
54517.4438797 ,
46366.66325235, 192872.63669555, 50035.40444538,
95226.6560928 ,
50547.81489465, 25268.83224858, 62635.88716991,
34669.44918167,
117369.04807865, 79570.16256859, 29637.7731142 ,
29989.47887016,
7756.2225401 , 29835.95697668, 67271.25283088,
44340.23851247,
91684.46261394, 63810.8340198 , 80449.25240847,
35102.22261493,
43725.02582983, 61355.87846011, 32852.18571348,
47285.70027847,
81095.64161965, 53340.80472393, 35507.12858801,
127940.00023025,
54748.27265082, 33764.27590934, 21555.24112081,
47793.88009026,
99600.47368742, 26427.56110399, 50843.0311352 ,
2393.36675538,
44245.38463449, 30557.61972929, 64331.44738662,
46052.30174221,
127948.19300162, 25875.99311672, 54013.85916036,
47862.33637501,
54548.65103582, 45496.28368746, 31788.72526496,
25415.45959194,
31554.54862708, 41029.82132688, 46789.06447165,
68798.88433921,
118678.1466251 , 38280.26527526, 51618.90192728,
60607.30748903,
40486.80042414, 23634.458532 , 48728.72961516,
133108.01377003,
49844.92467094, 14273.22715688, 65610.31437702,
111888.54121412,
71709.81448763, 140861.10089833, 29964.08104634,
49711.03130947,
95879.34214946, 58311.43819235, 21723.74978983,
134163.14808106,
132056.05882463, 55999.53310058, 40556.27936137,
71719.15541461,

```

        61303.17569047, 69364.60182646, 51074.45481062,
28320.44222635,
        51773.58311705, 239835.53704415, 64318.1023156 ,
37399.68698658,
        38289.86152092, 59716.48021255, 21721.41455808,
14608.88730471,
        52467.53834686, 50611.39269332, 50026.06245928,
11788.30069698,
        59391.43797137, 51698.22439083, 59672.88921996,
38739.74209449,
        52091.16644551, 17201.62221796, 70016.38987149,
30835.15898424,
        101950.59044208, 46806.18950446, 60690.11143484,
126922.35408882,
        7089.9382569 , 84927.52216659, 109939.32315248,
60867.65085507,
        13311.86850318, 28477.46071707, 35870.79961642,
40632.64055195,
        130166.96427995, 68066.86002907, 34033.59554858,
54548.88455899,
        42884.61600956, 37735.67541855, 25863.9277527 ,
58185.3564429 ,
        138145.48035431, 112389.92337885, 60911.95193693,
50891.34804191,
        121286.7450744 , 30918.49434596, 51936.08269254,
12699.70125396,
        57749.07933698, 45996.1445545 , 23637.15540436,
35095.67384671])

```

Get Model Evaluation

```

from sklearn.metrics import mean_squared_error, mean_absolute_error,
mean_absolute_percentage_error, r2_score

```

```

mean_squared_error(y_test,y_pred)

```

```

293014701.66187763

```

```

mean_absolute_percentage_error(y_test,y_pred)

```

```

0.24947046843526693

```

```

r2_score(y_test,y_pred)

```

```

0.8738616789555578

```

```

#Geet Visualization of Actual Vs Predicted Results

```

```

import matplotlib.pyplot as plt

```

```

plt.scatter(y_test,y_pred)
plt.xlabel("Actual Prices")

```



```
plt.ylabel('Predicted Prices')
plt.title('Actual Price vs Predicted Price')
plt.show()
```



```
X_new= df.sample(1)
```

```
X_new
```

	Brand	Model	Selling_Price	Year	Seller_Type	Owner
4	Bajaj	Bajaj Discover 100	24999	2012	0	1

	KM_Driven	Ex_Showroom_Price
4	35000	42859.0

```
X_new.shape
```

```
(1, 8)
```

```
X_new=X_new.drop(['Selling_Price','Brand','Model'],axis=1)
```

```
X_new
```

	Year	Seller_Type	Owner	KM_Driven	Ex_Showroom_Price
4	2012	0	1	35000	42859.0

```
X_new.shape
```

(1, 5)

```
y_pred_new= lr.predict(X_new)
```

```
y_pred_new
```

```
array([11014.37068462])
```