

Bike Price Prediction using Linear Regression

▼ Import Library

```
import pandas as pd
import numpy as np
```

▼ Import CSV file as Dataframe

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

▼ Get the first five rows of dataframe

```
df.head()
```

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▼ Get Information of Dataframe

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1061 entries, 0 to 1060
Data columns (total 8 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Brand                  1061 non-null   object
```

```
1  Model          1061 non-null  object
2  Selling_Price  1061 non-null  int64
3  Year           1061 non-null  int64
4  Seller_Type    1061 non-null  object
5  Owner          1061 non-null  object
6  KM_Driven      1061 non-null  int64
7  Ex_Showroom_Price 626 non-null  float64
dtypes: float64(1), int64(3), object(4)
memory usage: 66.4+ KB
```

▼ Get Missing values drop

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

▼ Get Summary Statistics

```
df.describe()
```

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▼ 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

```

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```

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 Shape of Dataframe

```
df.shape
```

```
(626, 8)
```

▼ 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)
```

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

▼ Define target variable (y) and features (X)

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

```
y.shape
```

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```
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
```

▼ Get Train Test Split

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```
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=2529)
```

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

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

▼ Get Model Train

```
from sklearn.linear_model import LinearRegression
```

```
model = LinearRegression()
```

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

```
LinearRegression()
```

► Get Model Prediction

```
y_pred = model.predict(X_test)
```

```
y_pred.shape
```

```
(188,)
```

```
y_pred
```

```
array([ 27210.52271465,  56340.08335163,  63471.94671996,  53627.63844785,
        55612.75744268,  53888.92259719,  33751.35275102,  60311.4950183 ,
       113713.05684467,  76639.49332954,  27826.7399381 ,  49919.83255841,
        65886.64311457,  26755.12664064,  48277.75426038, 127646.56079335,
        70047.10661635,  39350.67963653,  36081.03597878,  45360.79436339,
        48079.89470577,  44803.02464799,  55161.44026111,  71041.51821318,
        91689.22699159,  49301.53594645,  55988.19326252, 108171.54600296,
        32771.06897901,  25468.20072996,  17128.61806164, 179271.41130746,
        45698.99857622,  31371.09285079,  67886.52106737,  41492.49575815,
        56855.22238602,  47820.47003468,  74682.14053958,  24984.21822736,
        55374.00513699,  41412.36775222,  67991.60287764,  26553.59421844,
        89788.69870689,  45764.83633686, 133888.03770389, 106988.113825 ,
        25485946,  79512.43778826,  63914.38088173,
        13623937, -5396.37132904,  70377.44571174,
        92478411,  67509.85836352,  59735.05378847,
        22199.83644217, 15374.18984158,  44510.76819427,  30279.52476752,
       108243.77037514, 19291.8895874 ,  53614.312976 ,  59230.23269131,
        60174.2108109 ,  45924.63468736,  25770.81883496,  63471.36257814,
       242123.45729792,  61387.72544548,  56510.98127074,  48123.28087213,
        51668.27442011,  90279.76190495,  14827.76533556, 112437.70820504,
        35066.88027405,  30902.41069172,  31441.48921433, 125593.75847157,
        27705.38813164, -11590.29205553,  15582.17108685,  75113.64511232,
       504085.44522282, 123545.42050116,  74770.89327697,  50747.47663245,
        44174.3618212 ,  25426.7156106 ,  30298.3052462 ,  47625.67836414,
        27850.37544807,  28845.23330928,  31580.38624692,  32309.63375635,
        47979.16788554,  65955.46375944,  13432.28218017,  15368.80064986,
        31973.23052409, 110353.92870546,  68181.49509136,  23143.49139797,
        53194.65732076,  34603.36376989,  56002.50967868,  62432.66994305,
```

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```

391470.77533201, 3558.29480891, 36019.18494305, 70876.34866549,
72890.00667025, 137596.01384364, 27620.36308877, 135789.30486854,
39674.40366791, 58367.0924453 , 42401.21202624, 61864.4379567 ,
42688.89652842, 63710.34571021, 10604.39360071, 38458.82820943,
112251.84744225, 115403.00577536, 13658.41734785, 36196.83359584,
54146.22998932, 97297.85724851, 55029.68137265, 22923.26533437,
104569.97029689, 41965.75852017, 38759.68546491, 28930.61369011,
45231.66612551, 48475.43422775, 26739.7225731 , 53598.65972203,
32558.54954524, 32212.22834942, 68172.98738422, 71839.47716461,
32003.46692215, 40652.69995971, 39935.92211843, 63444.41846202,
44545.5818771 , 120873.38389616, 60926.58683174, 62641.82167496,
60816.47379994, 27098.95433573, 26803.64749618, 48956.00468627,
62032.88118713, 26471.97495723, 104937.23068766, 132903.3578847 ,
37469.2040942 , 57579.12080094, 40371.00915736, -7039.40662503,
26485.40030077, 90782.42554145, 52153.21149321, 56453.74542453,
80440.59426003, 31890.46870273, 49505.97985573, 24288.36959514,
25540.47481573, 117708.26333955, 23399.66596746, 63678.40865459,
70144.29372668, 33434.89010059, 60885.29444481, 58389.55370878,
35118.7040348 , 58729.4540196 , 34627.9532246 , 38583.4623973 ]

```

▼ Get Model Evaluation

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

```
mean_squared_error(y_test, y_pred)
```

```
554715615.5043668
```

```
mean_absolute_error(y_test, y_pred)
```

```
12225.7370104107
```

```
r2_score(y_test, y_pred)
```

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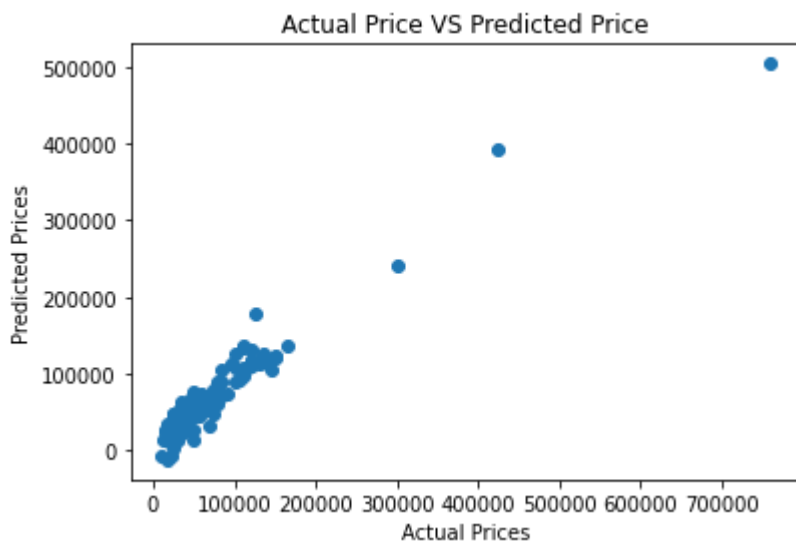


▼ Get Visualization of Actual VS Predicted Results

```

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()

```



▼ Get Future Predictions

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

```
df_new
```

| Brand | Model | Selling_Price | Year | Seller_Type | Owner | KM_Driven | Ex_Showroom_Price |
|-------|-------|---------------|------|-------------|-------|-----------|-------------------|
| Honda | | | | | | | |

```
df_new.shape
```

```
(1, 8)
```

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

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```
y_pred_new
```

```
array([22923.26533437])
```


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