

## Car 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/Car
%20Price.csv')
```

df

	Brand	Model	Year
Selling_Price \			
0	Maruti	Maruti 800 AC	2007
60000			
1	Maruti	Maruti Wagon R LXI Minor	2007
135000			
2	Hyundai	Hyundai Verna 1.6 SX	2012
600000			
3	Datsun	Datsun RediGO T Option	2017
250000			
4	Honda	Honda Amaze VX i-DTEC	2014
450000			
...	...	...	...
.			
4335	Hyundai	Hyundai i20 Magna 1.4 CRDi (Diesel)	2014
409999			
4336	Hyundai	Hyundai i20 Magna 1.4 CRDi	2014
409999			
4337	Maruti	Maruti 800 AC BSIII	2009
110000			
4338	Hyundai	Hyundai Creta 1.6 CRDi SX Option	2016
865000			
4339	Renault	Renault KWID RXT	2016
225000			

	KM_Driven	Fuel	Seller_Type	Transmission	Owner
0	70000	Petrol	Individual	Manual	First Owner
1	50000	Petrol	Individual	Manual	First Owner
2	100000	Diesel	Individual	Manual	First Owner
3	46000	Petrol	Individual	Manual	First Owner
4	141000	Diesel	Individual	Manual	Second Owner

4335	80000	Diesel	Individual	Manual	Second	Owner
4336	80000	Diesel	Individual	Manual	Second	Owner
4337	83000	Petrol	Individual	Manual	Second	Owner
4338	90000	Diesel	Individual	Manual	First	Owner
4339	40000	Petrol	Individual	Manual	First	Owner

[4340 rows x 9 columns]

## 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: 4340 entries, 0 to 4339
Data columns (total 9 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   Brand           4340 non-null   object
 1   Model           4340 non-null   object
 2   Year            4340 non-null   int64
 3   Selling_Price   4340 non-null   int64
 4   KM_Driven       4340 non-null   int64
 5   Fuel            4340 non-null   object
 6   Seller_Type     4340 non-null   object
 7   Transmission    4340 non-null   object
 8   Owner           4340 non-null   object
dtypes: int64(3), object(6)
memory usage: 339.1+ KB
```

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

	Year	Selling_Price	KM_Driven
count	4340.000000	4.340000e+03	4340.000000
mean	2013.090783	5.041273e+05	66215.777419
std	4.215344	5.785487e+05	46644.102194
min	1992.000000	2.000000e+04	1.000000
25%	2011.000000	2.087498e+05	35000.000000
50%	2014.000000	3.500000e+05	60000.000000
75%	2016.000000	6.000000e+05	90000.000000
max	2020.000000	8.900000e+06	806599.000000

```
df.head()
```

	Brand	Model	Year	Selling_Price	KM_Driven
Fuel \					
0	Maruti	Maruti 800 AC	2007	60000	70000
Petrol					
1	Maruti	Maruti Wagon R LXI Minor	2007	135000	50000
Petrol					
2	Hyundai	Hyundai Verna 1.6 SX	2012	600000	100000
Diesel					
3	Datsun	Datsun RediGO T Option	2017	250000	46000
Petrol					
4	Honda	Honda Amaze VX i-DTEC	2014	450000	141000
Diesel					

	Seller_Type	Transmission	Owner
0	Individual	Manual	First Owner
1	Individual	Manual	First Owner
2	Individual	Manual	First Owner
3	Individual	Manual	First Owner
4	Individual	Manual	Second Owner

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

```
Brand      0
Model      0
Year       0
Selling_Price  0
KM_Driven  0
Fuel       0
Seller_Type  0
Transmission  0
Owner      0
dtype: int64
```

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

```
Brand      29
Model     1491
Year       27
Selling_Price  445
KM_Driven  770
Fuel       5
Seller_Type  3
Transmission  2
Owner      5
dtype: int64
```

```
df.shape
```

```
(4340, 9)
```

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

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

```
Brand
Maruti      1280
Hyundai     821
Mahindra    365
Tata        361
Honda       252
Ford        238
Toyota      206
Chevrolet   188
Renault     146
Volkswagen  107
Skoda       68
Nissan       64
Audi        60
BMW         39
Fiat        37
Datsun       37
Mercedes-Benz 35
Mitsubishi   6
Jaguar       6
Land        5
Ambassador   4
Volvo        4
Jeep         3
OpelCorsa    2
MG           2
Isuzu        1
Force        1
Daewoo       1
Kia          1
dtype: int64
```

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

```
Transmission
Manual      3892
Automatic   448
dtype: int64
```

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

```
Model
Maruti Swift Dzire VDI      69
Maruti Alto 800 LXI         59
Maruti Alto LXi             47
Hyundai EON Era Plus        35
Maruti Alto LX              35
```

```

Mahindra KUV 100 G80 K4 Plus      1
Mahindra KUV 100 mFALCON D75 K8   1
Mahindra KUV 100 mFALCON D75 K8 AW 1
Mahindra KUV 100 mFALCON G80 K2 Plus 1
Volvo XC60 D5 Inscription         1
Length: 1491, dtype: int64

```

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

```

Seller_Type
Individual      3244
Dealer          994
Trustmark Dealer 102
dtype: int64

```

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

```

Owner
First Owner      2832
Second Owner     1106
Third Owner       304
Fourth & Above Owner 81
Test Drive Car    17
dtype: int64

```

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

```

Fuel
Diesel      2153
Petrol      2123
CNG         40
LPG         23
Electric     1
dtype: int64

```

## Get Column Names

```
df.columns
```

```

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

```

## Get Encoding of Categorical Features

```
df.replace({'Seller_Type':{'Individual':0,'Dealer':1,'Trustmark Dealer':2}},inplace=True)
```

```
df.replace({'Owner':{'First Owner':0,'Second Owner':1,'Third Owner':2,'Fourth & Above Owner':3,'Test Drive Car':4}},inplace=True)
```

```
df.replace({'Fuel':
{'Diesel':0,'Petrol':1,'CNG':2,'LPG':3,'Electric':4}},inplace=True)

df.replace({'Transmission':{'Automatic':0,'Manual':1}},inplace=True)

#df=pd.get_dummies(df,columns=['Fuel','Transmission'])
```

```
df
```

	Brand	Model	Year
Selling_Price \			
0	Maruti	Maruti 800 AC	2007
60000			
1	Maruti	Maruti Wagon R LXI Minor	2007
135000			
2	Hyundai	Hyundai Verna 1.6 SX	2012
600000			
3	Datsun	Datsun RediGO T Option	2017
250000			
4	Honda	Honda Amaze VX i-DTEC	2014
450000			
...	...	...	...
.			
4335	Hyundai	Hyundai i20 Magna 1.4 CRDi (Diesel)	2014
409999			
4336	Hyundai	Hyundai i20 Magna 1.4 CRDi	2014
409999			
4337	Maruti	Maruti 800 AC BSIII	2009
110000			
4338	Hyundai	Hyundai Creta 1.6 CRDi SX Option	2016
865000			
4339	Renault	Renault KWID RXT	2016
225000			

	KM_Driven	Fuel	Seller_Type	Transmission	Owner
0	70000	1	0	1	0
1	50000	1	0	1	0
2	100000	0	0	1	0
3	46000	1	0	1	0
4	141000	0	0	1	1
...	...	...	...	...	...
4335	80000	0	0	1	1
4336	80000	0	0	1	1
4337	83000	1	0	1	1
4338	90000	0	0	1	0
4339	40000	1	0	1	0

```
[4340 rows x 9 columns]
```

## Define y

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

```
y.shape
```

```
(4340,)
```

```
y
```

```
0      60000
1     135000
2     600000
3     250000
4     450000
```

```
...
4335    409999
4336    409999
4337    110000
4338    865000
4339    225000
```

```
Name: Selling_Price, Length: 4340, dtype: int64
```

```
df.columns
```

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

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

```
X.shape
```

```
(4340, 6)
```

```
X
```

	Year	KM_Driven	Fuel	Seller_Type	Transmission	Owner
0	2007	70000	1	0	1	0
1	2007	50000	1	0	1	0
2	2012	100000	0	0	1	0
3	2017	46000	1	0	1	0
4	2014	141000	0	0	1	1
...	...	...	...	...	...	...
4335	2014	80000	0	0	1	1
4336	2014	80000	0	0	1	1
4337	2009	83000	1	0	1	1
4338	2016	90000	0	0	1	0
4339	2016	40000	1	0	1	0

```
[4340 rows x 6 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

((3038, 6), (1302, 6), (3038,), (1302,))
```

## Get Model Test

```
from sklearn.linear_model import LinearRegression

lr= LinearRegression()
```

y\_train

```
385      600000
1440     200000
2265     239000
338      490000
4290      37500
```

...

```
2004     458000
2868     260000
3503      80000
2280     300000
2120     220000
```

Name: Selling\_Price, Length: 3038, dtype: int64

X\_train

	Year	KM_Driven	Fuel	Seller_Type	Transmission	Owner
385	2019	10000	1	0	1	0
1440	2006	73756	0	0	1	0
2265	2016	41000	1	1	1	0
338	2015	60000	0	0	1	1
4290	2004	90000	1	0	1	1
...	...	...	...	...	...	...
2004	2014	80000	0	0	1	0
2868	2010	50000	1	0	1	0
3503	2002	40000	1	0	1	3
2280	2016	30000	1	0	1	0
2120	2011	58000	1	1	1	1

[3038 rows x 6 columns]

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



```
LinearRegression()
```

### Get Model Prediction

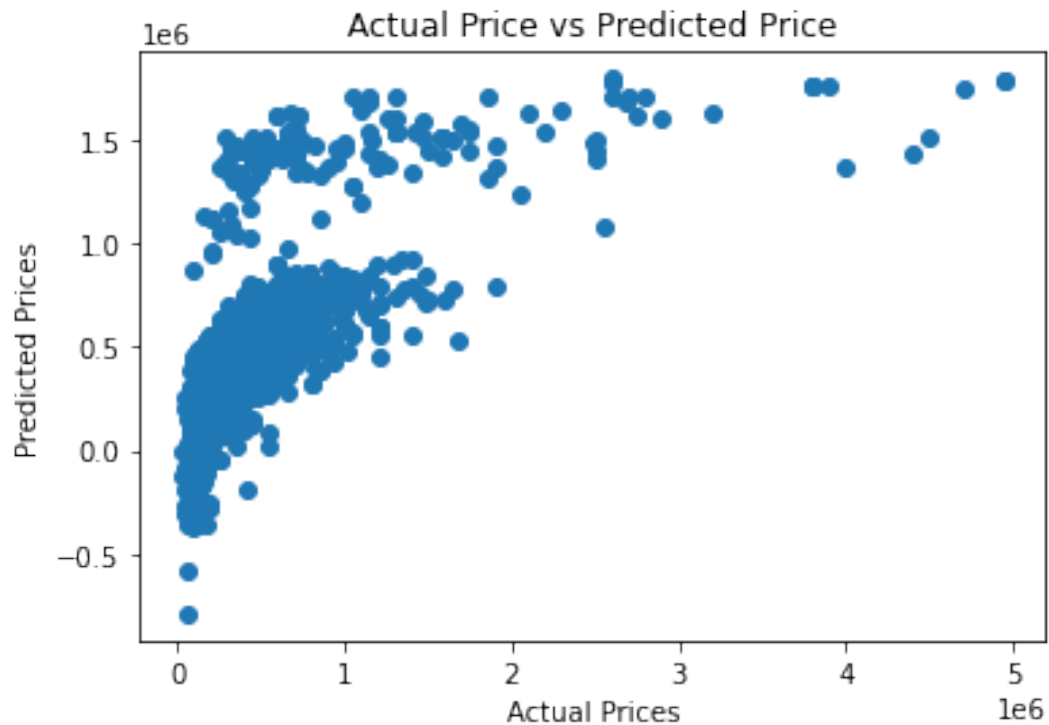
```
y_pred= lr.predict(X_test)
y_pred.shape
(1302,)
y_pred
array([ 429912.75948648,  532127.30528067, 1507458.24765234, ...,
        194865.88167524, 1508997.29560916,  559837.08726987])
```

### 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)
152798998815.43036
mean_absolute_percentage_error(y_test,y_pred)
0.7562522664663961
r2_score(y_test,y_pred)
0.4523964518624757
```

#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	Year	Selling_Price	KM_Driven
Fuel \	Maruti	Maruti Wagon R LXI Minor	2009	150000	53000
697					
1					

	Seller_Type	Transmission	Owner
697	0	1	0

```
X_new.shape
```

```
(1, 9)
```

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

```
X_new
```

	Year	KM_Driven	Fuel	Seller_Type	Transmission	Owner
697	2009	53000	1	0	1	0

```
X_new.shape
```

```
(1, 6)
```

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

```
y_pred_new
```

```
array([125601.80485147])
```