

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

	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)
```

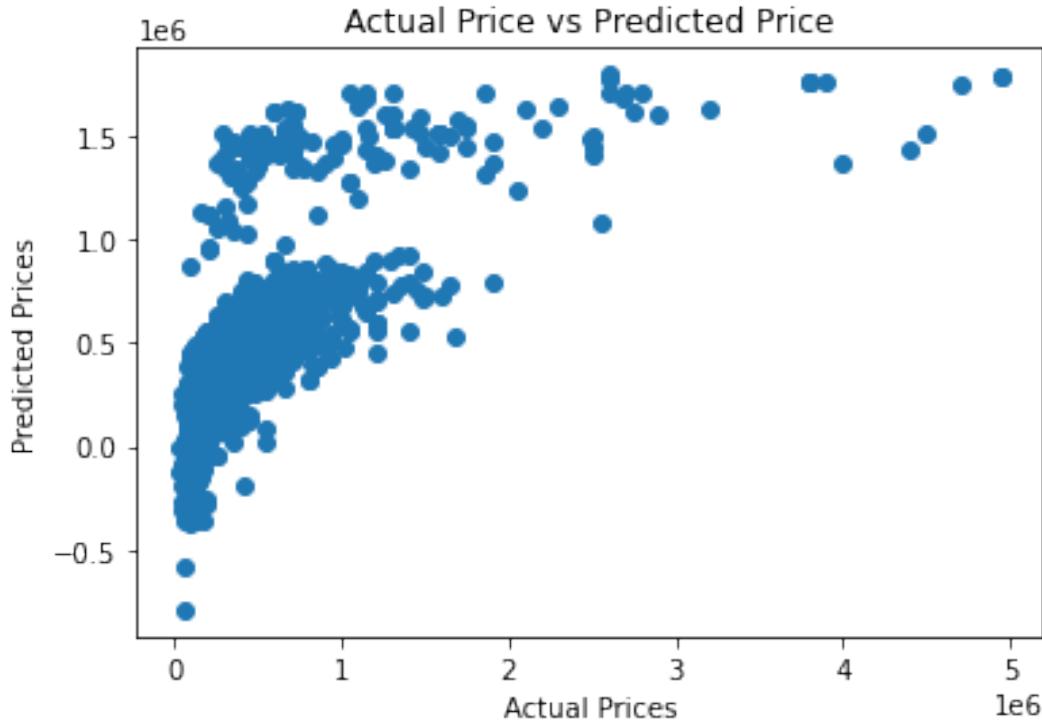
```
LinearRegression()
```

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  
  
#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()
```



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

```
X_new
```

	Brand	Model	Year	Selling_Price	KM_Driven
Fuel \	Maruti	Maruti Wagon R LXI Minor	2009	150000	53000
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])
```