

Double-click (or enter) to edit

## Car Price Prediction using Linear Regression

### ▼ Importing libraries

```
import pandas as pd
```

```
import numpy as np
```

### ▼ Importing CSV as DataFrame

```
df=pd.read_csv('/content/sample_data/Car Price.csv')
```

### ▼ Getting the first five rows of the DataFrame

```
df.head()
```



	Brand	Model	Year	Selling_Price	KM_Driven	Fuel	Seller_Type	Transmission
0	Maruti	Maruti 800 AC	2007	60000	70000	Petrol	Individual	Manual
1	Maruti	Maruti Wagon R LXI Minor	2007	135000	50000	Petrol	Individual	Manual
2	Hyundai	Hyundai Verna 1.6 SX	2012	600000	100000	Diesel	Individual	Manual
		Datsun						
		Option						

## ▼ Getting information of DataFrame

DIG

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 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: 305.3+ KB
```

## ▼ Getting summary statistics

df.describe()

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

## ➤ Getting 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[['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[['Fuel']].value_counts()
```

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

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

```
Seller_Type
Individual      3244
Dealer          994
Trustmark Dealer 102
dtype: int64
```

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

```
Transmission
Manual      3892
Automatic   448
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
```

## ▼ Getting column names

```
df.columns
```

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

## ▼ Getting Shape of DataFrame

```
df.shape
```

```
(4340, 9)
```

## ▼ Getting encoding of categorical features

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

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

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

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

Defining y(dependent/label/target variable) and  
▼ X(independent/features/attribute variable)

```
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
```

```
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	0	0	0	0	
2	2012	100000	1	0	0	0	

## ▼ Getting 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=2529)
```

```
4337 2009 83000 0 0 0 1
```

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

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

```
4340 rows x 6 columns
```

## ▼ Getting Model Train

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

```
lr=LinearRegression()
```

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

```
LinearRegression()
```

## ▼ Getting model prediction

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

```
y_pred.shape
```

```
(1302,)
```

y\_pred

```
array([502458.82786413, 646333.17428704, 521962.74075836, ...,  
       620183.32683781, 315403.8278857 , 731862.54196037])
```

## ▼ Getting model evaluation

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

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

```
193242972302.19553
```

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

```
228808.95522977872
```

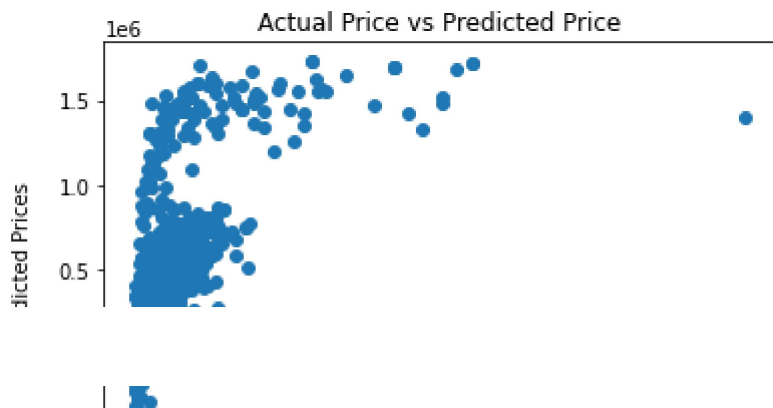
```
r2_score(y_test,y_pred)
```

```
0.4075563394370839
```

## ▼ Getting 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()
```





## ▼ Getting future predictions

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

```
df_new
```

	Brand	Model	Year	Selling_Price	KM_Driven	Fuel	Seller_Type	Transmission
3562	Maruti	Maruti Swift VDI Optional	2017	459999	50000	1	0	0



```
df_new.shape
```

```
(1, 9)
```

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

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

```
y_pred_new
```

```
array([659625.04339699])
```

---

✓

0s

completed at 1:11 PM

●

✕