

Model For used car price prediction

```
In [2]: import pandas as pd  
import numpy as np
```

Stage 1

1.1 Reading the file

1.2 Exploratory data analysis (to check correlation)

1.3 Identifying Input and Output

1.1 Reading the file

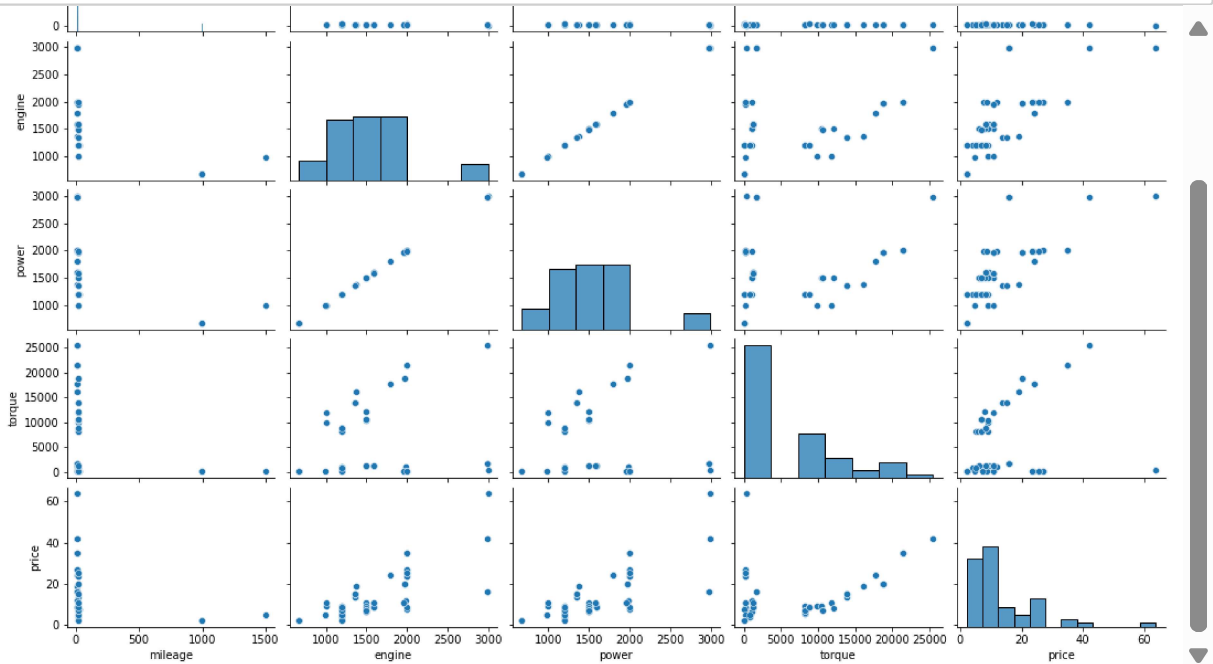
```
In [27]: data=pd.read_csv(r"C:\Users\Param\Desktop\new used cars.csv")  
print(data)
```

	mileage	engine	power	torque	price
0	7.81	2996	2996	333	63.75
1	17.40	999	999	9863	8.99
2	20.68	1995	1995	188	23.75
3	16.50	1353	1353	13808	13.56
4	14.67	1798	1798	17746	24.00
5	18.70	1199	1199	887	5.45
6	18.90	1197	1197	8186	5.12
7	15.80	1591	1591	1213	9.30
8	13.50	2987	2987	25479	42.00
9	17.00	1198	1198	1085	8.02
10	17.40	1497	1497	1176	10.95
11	16.42	1498	1498	10455	8.99
12	18.88	1995	1995	184	7.45
13	18.15	998	998	11835	10.95
14	18.90	1197	1197	8186	5.78
15	21.00	1197	1197	8180	8.95
16	18.88	1995	1995	184	8.50
17	22.69	1995	1995	190	23.50
18	17.00	1497	1497	12136	7.95
19	14.10	1368	1368	16077	18.95
20	12.40	1996	1996	21501	35.00
21	16.10	1197	1197	85	2.09
22	18.00	1497	1497	1173	6.25
23	998.00	671	671	90	2.12
24	18.53	1968	1968	18774	20.00
25	15.29	1591	1591	1213	8.35
26	21.40	1197	1197	831	5.03
27	17.10	1496	1496	10594	6.95
28	12.55	2982	2982	1685	16.00
29	16.80	1353	1353	13808	15.22
30	13.50	1999	1999	177	26.95
31	18.60	1197	1197	8186	6.70
32	17.10	1956	1956	170	10.75
33	20.68	1995	1995	188	25.50
34	22.56	1197	1197	8850	8.38
35	17.60	1582	1582	1262	10.90
36	17.80	1198	1198	867	3.95
37	26.49	1199	1199	72	7.25
38	1498.00	986	986	200	4.75
39	10.98	1984	1984	1144	11.95
40	12.40	1996	1996	21501	35.00
41	16.10	1197	1197	85	2.09
42	18.00	1497	1497	1173	6.25
43	998.00	671	671	90	2.12
44	18.53	1968	1968	18774	20.00
45	15.29	1591	1591	1213	8.35
46	21.40	1197	1197	831	5.03
47	17.10	1496	1496	10594	6.95
48	12.55	2982	2982	1685	16.00
49	16.80	1353	1353	13808	15.22
50	13.50	1999	1999	177	26.95
51	18.60	1197	1197	8186	6.70
52	17.10	1956	1956	170	10.75
53	20.68	1995	1995	188	25.50

```
54    22.56    1197    1197    8850    8.38
55    17.60    1582    1582    1262   10.90
```

1.2 EDA (Exploratory Data analysis)

```
In [28]: import seaborn as sns
sns.pairplot(data,height=2,aspect=1.5)
```



```
In [29]: sns.heatmap(data.corr(), annot=True, linewidth=0.6)
```

```
Out[29]: <AxesSubplot:>
```



Stage 2: Splitting the Data

```
In [30]: X=data.drop(['price','mileage'],axis='columns')
print(X)
Y=data.drop(['mileage','engine','power','torque'],axis='columns')
print(Y)
```

	engine	power	torque
0	2996	2996	333
1	999	999	9863
2	1995	1995	188
3	1353	1353	13808
4	1798	1798	17746
5	1199	1199	887
6	1197	1197	8186
7	1591	1591	1213
8	2987	2987	25479
9	1198	1198	1085
10	1497	1497	1176
11	1498	1498	10455
12	1995	1995	184
13	998	998	11835
14	1197	1197	8186
15	1197	1197	8180
16	1995	1995	184
17	1995	1995	190
18	1197	1197	13136

Stage 3:Splitting

```
In [31]: 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=42)
print(X_train)
print(X_test)
print(Y_train)
print(Y_test)
```

	engine	power	torque
46	1197	1197	831
17	1995	1995	190
54	1197	1197	8850
41	1197	1197	85
49	1353	1353	13808
12	1995	1995	184
30	1999	1999	177
39	1984	1984	1144
16	1995	1995	184
2	1995	1995	188
25	1591	1591	1213
19	1368	1368	16077
6	1197	1197	8186
14	1197	1197	8186
36	1198	1198	867
24	1968	1968	18774
5	1199	1199	887
38	986	986	200
33	1005	1005	100

Stage 3: Fitting of model or Equation

```
In [32]: from sklearn.linear_model import LinearRegression  
equation=LinearRegression()
```

```
In [33]: equation.fit(X_train,Y_train)
```

```
Out[33]: LinearRegression()
```

```
In [34]: equation.intercept_
```

```
Out[34]: array([-10.35620098])
```

```
In [35]: equation.coef_
```

```
Out[35]: array([[0.0061953 , 0.0061953 , 0.00057644]])
```

```
In [36]: Y_test_predicted=equation.predict(X_test)  
Y_test_predicted
```

```
Out[36]: array([[ 9.97319504],  
                [14.51464047],  
                [14.47141888],  
                [13.97780955],  
                [ 9.19407068],  
                [14.28692331],  
                [26.95799426],  
                [14.36773926],  
                [ 7.70741826],  
                [ 8.86869095],  
                [27.5638701 ],  
                [ 5.11317462],  
                [ 8.87042026],  
                [ 8.83176352],  
                [14.28692331],  
                [13.97780955],  
                [-1.99022811]])
```

Stage 4: Evaluation of errors and R2

```
In [37]: from sklearn import metrics  
MAE=metrics.mean_absolute_error(Y_test, Y_test_predicted)  
print(MAE)
```

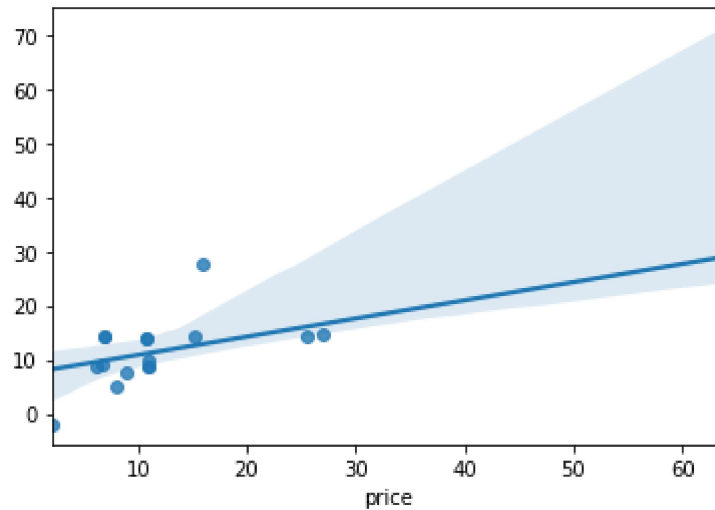
```
6.60815064654816
```

```
In [38]: R2=metrics.r2_score(Y_test, Y_test_predicted)
R2
```

```
Out[38]: 0.4002569463316845
```

```
In [39]: sns.regplot(x=Y_test, y=Y_test_predicted)
```

```
Out[39]: <AxesSubplot:xlabel='price'>
```



```
In [ ]:
```