

CAR PRICE PREDICTION _____

ABOUT DATASET

The dataset is used to predict the price of a car. It contains 205 rows and 26 columns such as car_ID, symboling, CarName, fueltype, aspiration, doornumber, carbody, drivewheel, enginelocation, wheelbase, carlength, carwidth, carheight, curbweight, enginetype, cylindernumber, enginesize, fuelsystem, boreratio, stroke, compressionratio, horsepower, peakrpm, citympg, highwaympg, price.

IMPORTING PYTHON LIBRARIES

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_percentage_error
from sklearn.metrics import mean_absolute_error
from sklearn.metrics import mean_squared_error
from sklearn.metrics import r2_score
```

LOADING DATASET

```
df=pd.read_csv('/home/anusha/Desktop/CarPrice_Assignment.csv')
df
```

	car_ID	symboling	CarName	fueltype	
aspiration \					
0	1	3	alfa-romero giulia	gas	std
1	2	3	alfa-romero stelvio	gas	std
2	3	1	alfa-romero Quadrifoglio	gas	std
3	4	2	audi 100 ls	gas	std
4	5	2	audi 100ls	gas	std
..
200	201	-1	volvo 145e (sw)	gas	std
201	202	-1	volvo 144ea	gas	turbo
202	203	-1	volvo 244dl	gas	std
203	204	-1	volvo 246	diesel	turbo

204	205	-1	volvo 264gl	gas	turbo	
	doornumber	carbody	drivewheel	engine	location	wheelbase ...
\						
0	two	convertible	rwd	front	88.6	...
1	two	convertible	rwd	front	88.6	...
2	two	hatchback	rwd	front	94.5	...
3	four	sedan	fwd	front	99.8	...
4	four	sedan	4wd	front	99.4	...
..
200	four	sedan	rwd	front	109.1	...
201	four	sedan	rwd	front	109.1	...
202	four	sedan	rwd	front	109.1	...
203	four	sedan	rwd	front	109.1	...
204	four	sedan	rwd	front	109.1	...
	enginesize	fuelsystem	boreratio	stroke	compressionratio	
horsepower \						
0	130	mpfi	3.47	2.68	9.0	
111						
1	130	mpfi	3.47	2.68	9.0	
111						
2	152	mpfi	2.68	3.47	9.0	
154						
3	109	mpfi	3.19	3.40	10.0	
102						
4	136	mpfi	3.19	3.40	8.0	
115						
..	
...						
200	141	mpfi	3.78	3.15	9.5	
114						
201	141	mpfi	3.78	3.15	8.7	
160						
202	173	mpfi	3.58	2.87	8.8	
134						
203	145	idi	3.01	3.40	23.0	
106						

204	141	mpfi	3.78	3.15	9.5
-----	-----	------	------	------	-----

	peakrpm	citympg	highwaympg	price
0	5000	21	27	13495.0
1	5000	21	27	16500.0
2	5000	19	26	16500.0
3	5500	24	30	13950.0
4	5500	18	22	17450.0
...
200	5400	23	28	16845.0
201	5300	19	25	19045.0
202	5500	18	23	21485.0
203	4800	26	27	22470.0
204	5400	19	25	22625.0

[205 rows x 26 columns]

DATA PREPROCESSING

df.head()

car_ID	symboling	CarName	fueltype	aspiration	
doornumber	\				
0	1	3	alfa-romero giulia	gas std	
two					
1	2	3	alfa-romero stelvio	gas std	
two					
2	3	1	alfa-romero Quadrifoglio	gas std	
two					
3	4	2	audi 100 ls	gas std	
four					
4	5	2	audi 100ls	gas std	
four					
carbody	drivewheel	engine	location	wheelbase	...
enginesize	\				
0	convertible	rwd	front	88.6	... 130
1	convertible	rwd	front	88.6	... 130
2	hatchback	rwd	front	94.5	... 152
3	sedan	fwd	front	99.8	... 109
4	sedan	4wd	front	99.4	... 136
fuelsystem	boreratio	stroke	compressionratio	horsepower	peakrpm
citympg	\				

```

0      mpfi      3.47      2.68      9.0      111      5000
21
1      mpfi      3.47      2.68      9.0      111      5000
21
2      mpfi      2.68      3.47      9.0      154      5000
19
3      mpfi      3.19      3.40      10.0     102      5500
24
4      mpfi      3.19      3.40      8.0      115      5500
18

```

```

      highwaympg      price
0          27  13495.0
1          27  16500.0
2          26  16500.0
3          30  13950.0
4          22  17450.0

```

```
[5 rows x 26 columns]
```

```
df.tail()
```

```

      car_ID  symboling      CarName  fueltype  aspiration  doornumber
\
200      201      -1  volvo 145e (sw)      gas      std      four
201      202      -1      volvo 144ea      gas      turbo      four
202      203      -1      volvo 244dl      gas      std      four
203      204      -1      volvo 246      diesel      turbo      four
204      205      -1      volvo 264gl      gas      turbo      four

```

```

      carbody drivewheel  enginelocation  wheelbase  ...  enginesize
fuelsystem \
200      sedan      rwd      front      109.1  ...      141
mpfi
201      sedan      rwd      front      109.1  ...      141
mpfi
202      sedan      rwd      front      109.1  ...      173
mpfi
203      sedan      rwd      front      109.1  ...      145
idi
204      sedan      rwd      front      109.1  ...      141
mpfi

```

```

      boreratio  stroke  compressionratio  horsepower  peakrpm  citympg  \
200      3.78      3.15      9.5      114      5400      23
201      3.78      3.15      8.7      160      5300      19

```

202	3.58	2.87	8.8	134	5500	18
203	3.01	3.40	23.0	106	4800	26
204	3.78	3.15	9.5	114	5400	19

	highwaympg	price
200	28	16845.0
201	25	19045.0
202	23	21485.0
203	27	22470.0
204	25	22625.0

[5 rows x 26 columns]

df.shape

(205, 26)

df.columns

```
Index(['car_ID', 'symboling', 'CarName', 'fueltype', 'aspiration',
      'doornumber', 'carbody', 'drivewheel', 'enginelocation',
      'wheelbase',
      'carlength', 'carwidth', 'carheight', 'curbweight',
      'enginetype',
      'cylindernumber', 'enginesize', 'fuelsystem', 'boreratio',
      'stroke',
      'compressionratio', 'horsepower', 'peakrpm', 'citympg',
      'highwaympg',
      'price'],
      dtype='object')
```

df.dtypes

car_ID	int64
symboling	int64
CarName	object
fueltype	object
aspiration	object
doornumber	object
carbody	object
drivewheel	object
enginelocation	object
wheelbase	float64
carlength	float64
carwidth	float64
carheight	float64
curbweight	int64
enginetype	object
cylindernumber	object
enginesize	int64
fuelsystem	object

```
boreratio      float64
stroke         float64
compressionratio float64
horsepower      int64
peakrpm        int64
citympg        int64
highwaympg     int64
price          float64
dtype: object
```

Here

CarName,fueltype,aspiration,doornumber,carbody,drivewheel,engine location,enginetype,cylindernumber,fuelsystem are objects.

```
df.isna().sum()
```

```
car_ID      0
symboling   0
CarName     0
fueltype    0
aspiration  0
doornumber  0
carbody     0
drivewheel  0
engine location 0
wheelbase   0
carlength   0
carwidth    0
carheight   0
curbweight  0
enginetype  0
cylindernumber 0
enginesize  0
fuelsystem  0
boreratio   0
stroke      0
compressionratio 0
horsepower  0
peakrpm     0
citympg     0
highwaympg  0
price       0
dtype: int64
```

No missing values are found

```
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 205 entries, 0 to 204
Data columns (total 26 columns):
#   Column                Non-Null Count  Dtype
---  -
0   car_ID                205 non-null    int64
1   symboling              205 non-null    int64
2   CarName               205 non-null    object
3   fueltype              205 non-null    object
4   aspiration            205 non-null    object
5   doornumber            205 non-null    object
6   carbody               205 non-null    object
7   drivewheel           205 non-null    object
8   enginelocation        205 non-null    object
9   wheelbase            205 non-null    float64
10  carlength             205 non-null    float64
11  carwidth              205 non-null    float64
12  carheight            205 non-null    float64
13  curbweight           205 non-null    int64
14  enginetype           205 non-null    object
15  cylindernumber       205 non-null    object
16  enginesize           205 non-null    int64
17  fuelsystem           205 non-null    object
18  boreratio            205 non-null    float64
19  stroke               205 non-null    float64
20  compressionratio     205 non-null    float64
21  horsepower           205 non-null    int64
22  peakrpm              205 non-null    int64
23  citympg              205 non-null    int64
24  highwaympg           205 non-null    int64
25  price                205 non-null    float64
dtypes: float64(8), int64(8), object(10)
memory usage: 41.8+ KB

```

DATA VISUALIZATION

```

df['CarName'].value_counts()

CarName
toyota corona      6
toyota corolla     6
peugeot 504        6
subaru dl          4
mitsubishi mirage g4  3
..
mazda glc 4        1
mazda rx2 coupe    1
maxda glc deluxe   1
maxda rx3          1

```

```

volvo 246      1
Name: count, Length: 147, dtype: int64

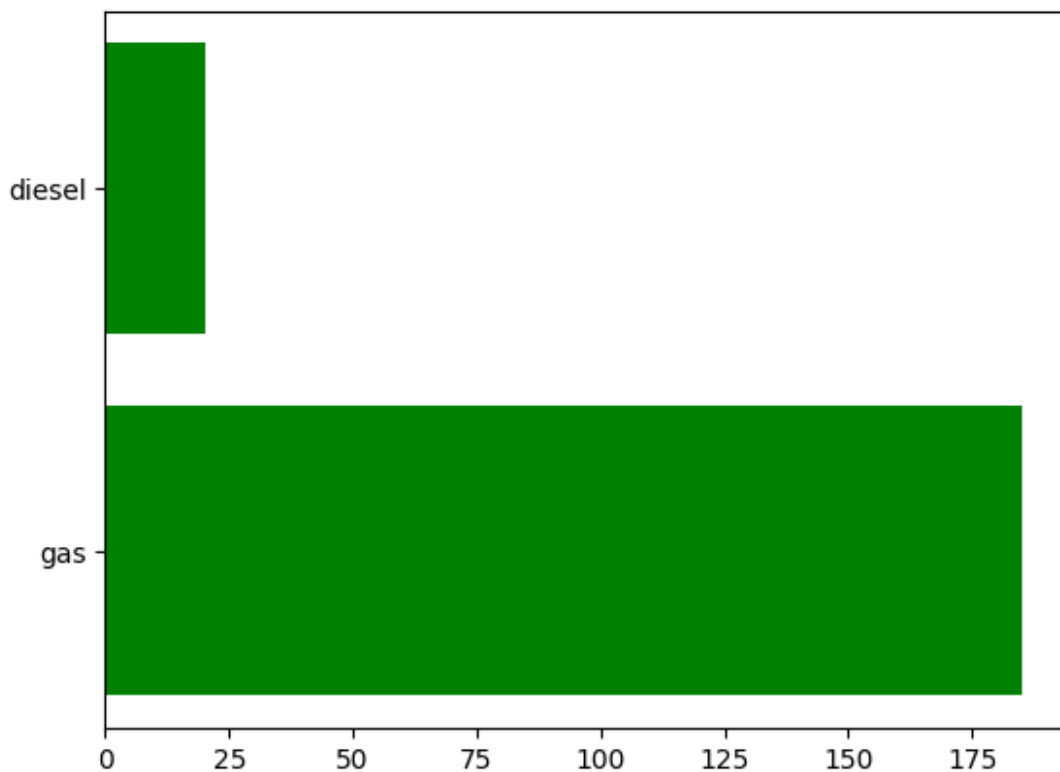
df['fueltype'].value_counts()

fueltype
gas      185
diesel    20
Name: count, dtype: int64

plt.barh(df['fueltype'].value_counts().index,df['fueltype'].value_counts(),color='green')

<BarContainer object of 2 artists>

```



```

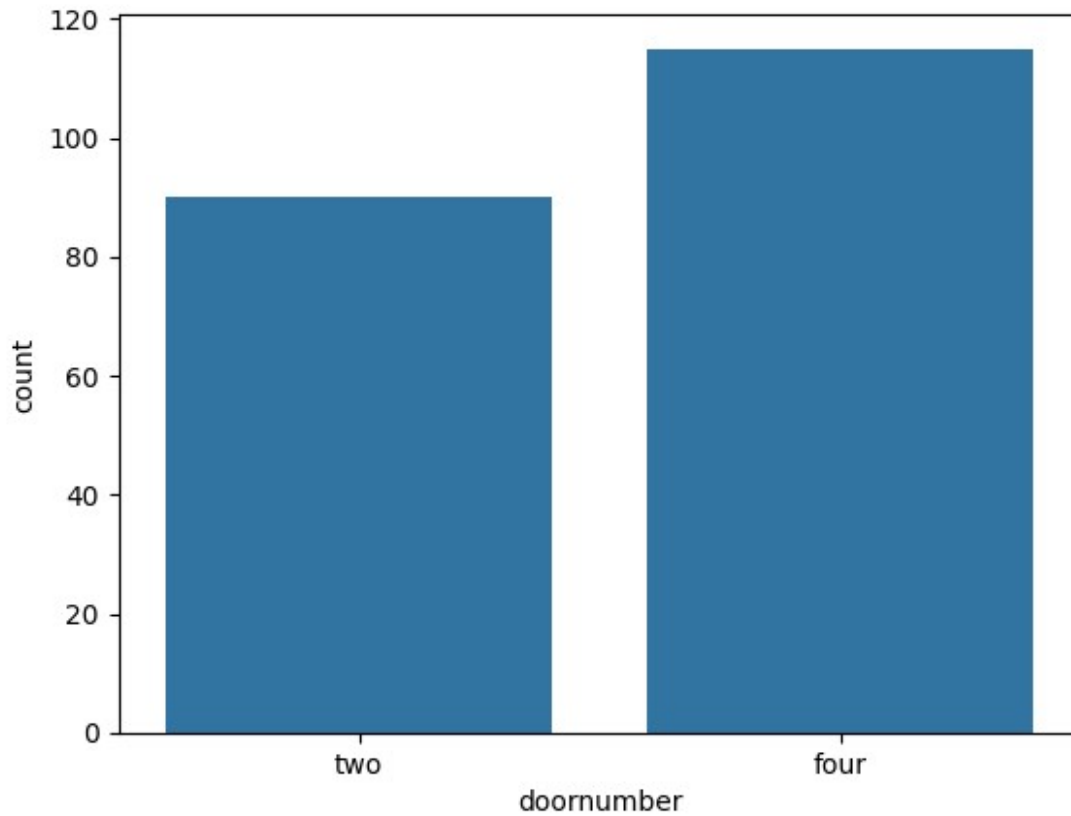
df['doornumber'].value_counts()

doornumber
four      115
two       90
Name: count, dtype: int64

sns.countplot(x=df['doornumber'])

<Axes: xlabel='doornumber', ylabel='count'>

```

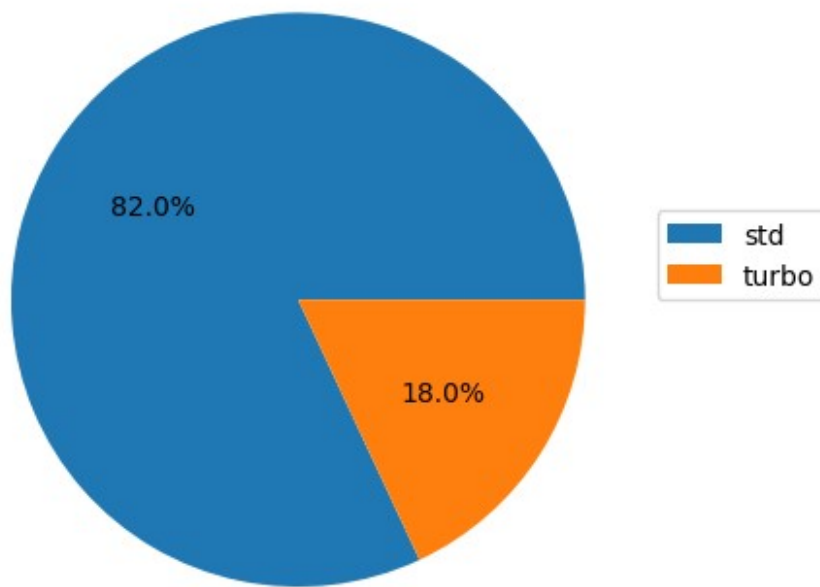



```
df['aspiration'].value_counts()
aspiration
std      168
turbo     37
Name: count, dtype: int64

plt.pie(df['aspiration'].value_counts(),autopct='%1.1f%%')
plt.legend(df['aspiration'].value_counts().index,loc=(1,0.5))
plt.title('aspiration graph',color='black')

Text(0.5, 1.0, 'aspiration graph')
```

aspiration graph

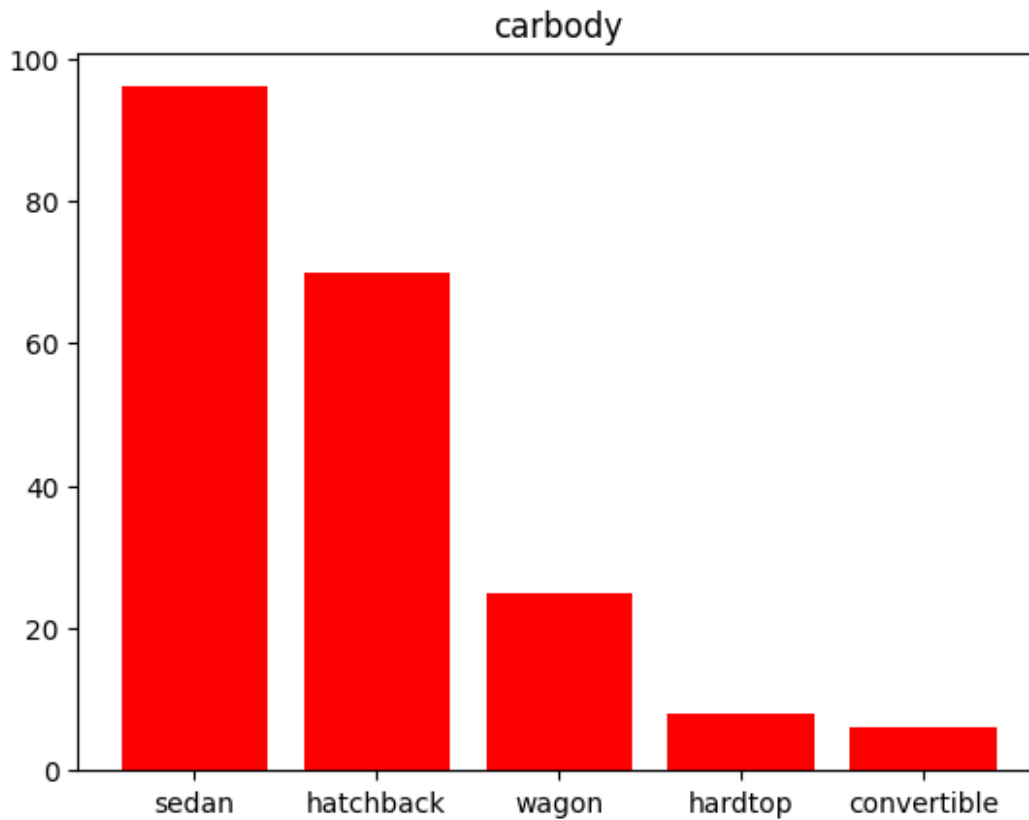


```
df['carbody'].value_counts()

carbody
sedan      96
hatchback  70
wagon      25
hardtop     8
convertible 6
Name: count, dtype: int64

plt.bar(df['carbody'].value_counts().index,df['carbody'].value_counts(
),color='red')
plt.title('carbody')

Text(0.5, 1.0, 'carbody')
```



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

```
drivewheel
```

```
fwd    120
```

```
rwd     76
```

```
4wd      9
```

```
Name: count, dtype: int64
```

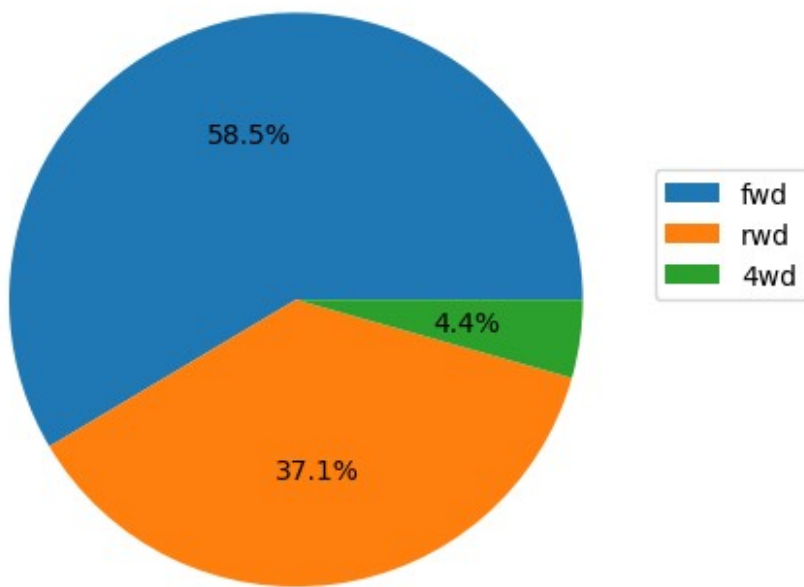
```
plt.pie(df['drivewheel'].value_counts(),autopct='%1.1f%%')
```

```
plt.legend(df['drivewheel'].value_counts().index,loc=(1,0.5))
```

```
plt.title('drivewheel graph',color='red')
```

```
Text(0.5, 1.0, 'drivewheel graph')
```

drivewheel graph



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

```
enginetype
```

```
ohc      148
```

```
ohcf      15
```

```
ohcv      13
```

```
dohc      12
```

```
l         12
```

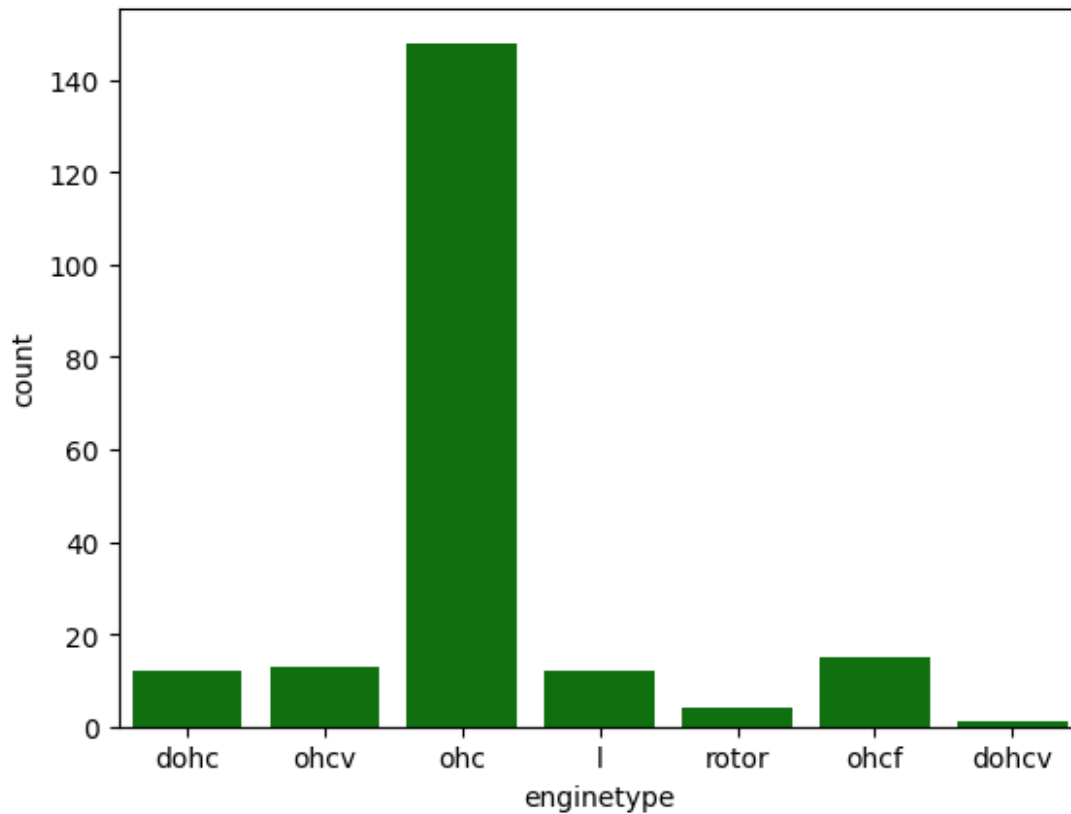
```
rotor      4
```

```
dohcv      1
```

```
Name: count, dtype: int64
```

```
sns.countplot(x=df['enginetype'],data=df,color='green')
```

```
<Axes: xlabel='enginetype', ylabel='count'>
```



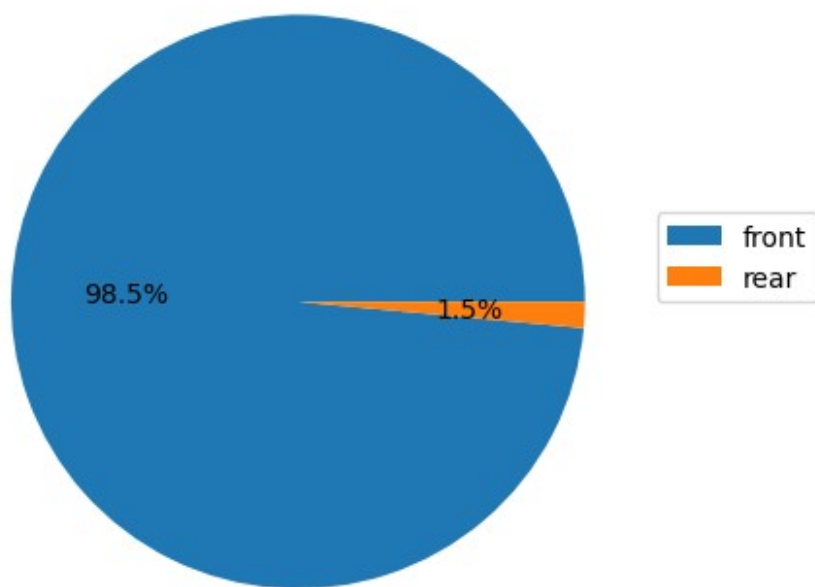
```
df['enginelocation'].value_counts()

enginelocation
front    202
rear      3
Name: count, dtype: int64

plt.pie(df['enginelocation'].value_counts(),autopct='%1.1f%%')
plt.legend(df['enginelocation'].value_counts().index,loc=(1,0.5))
plt.title('engine location graph',color='green')

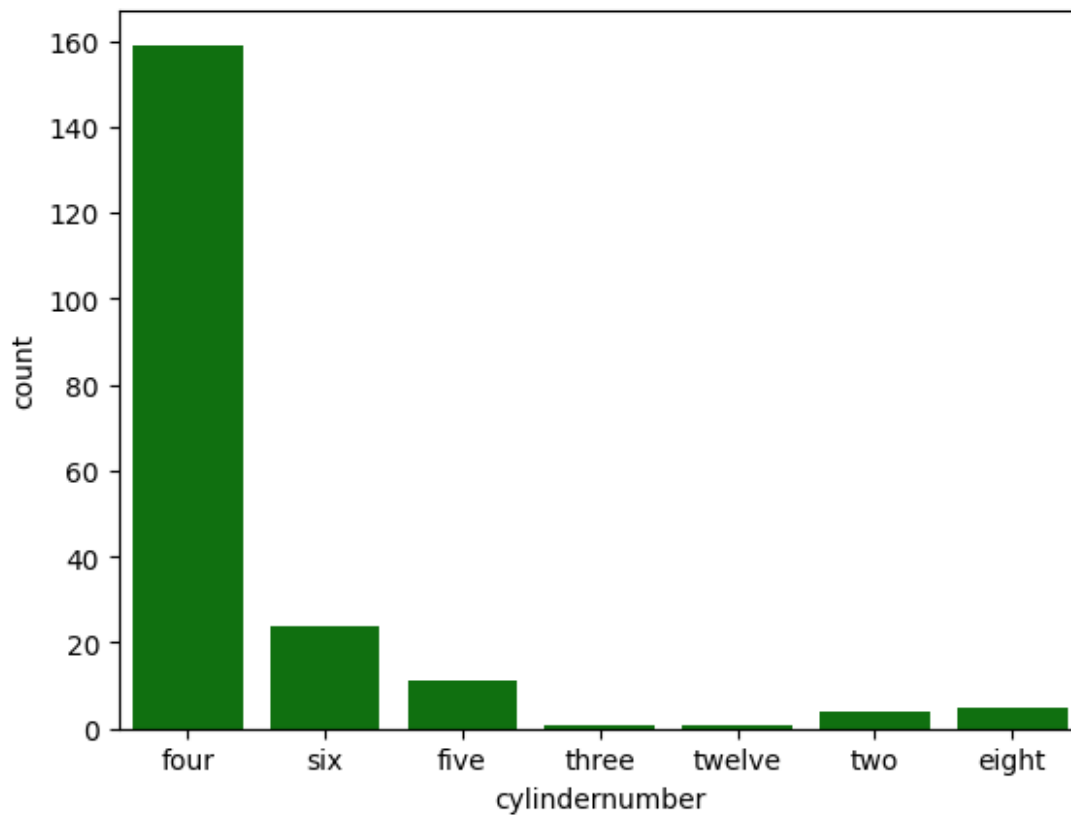
Text(0.5, 1.0, 'engine location graph')
```

engine location graph



```
df['cylindernumber'].value_counts()
cylindernumber
four      159
six       24
five      11
eight      5
two        4
three      1
twelve     1
Name: count, dtype: int64

sns.countplot(x=df['cylindernumber'],data=df,color='green')
<Axes: xlabel='cylindernumber', ylabel='count'>
```



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

```
fuelsystem
```

```
mpfi      94
```

```
2bbl      66
```

```
idi       20
```

```
1bbl      11
```

```
spdi       9
```

```
4bbl       3
```

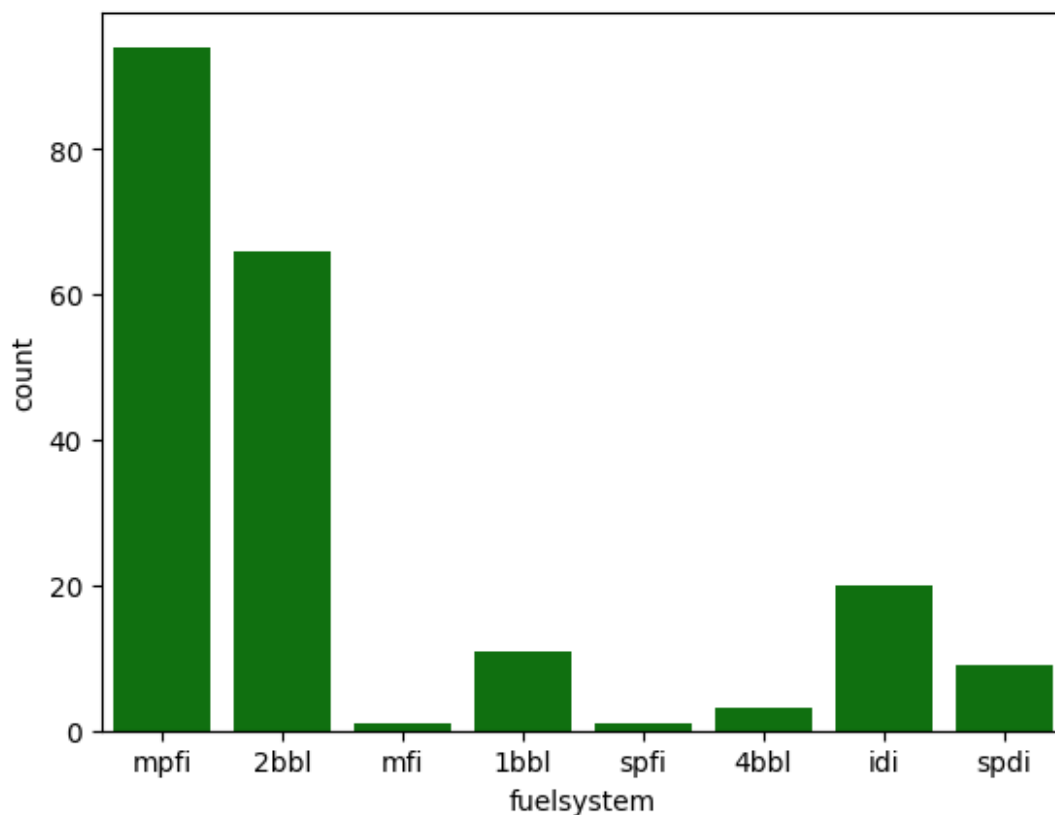
```
mfi        1
```

```
spfi       1
```

```
Name: count, dtype: int64
```

```
sns.countplot(x=df['fuelsystem'],data=df,color='green')
```

```
<Axes: xlabel='fuelsystem', ylabel='count'>
```



ENCODING

```
#Label encoding
lab=LabelEncoder()
lst=['fueltype','aspiration','doornumber','engine location']

for i in lst:
    if df[i].dtype=='object':
        df[i]=lab.fit_transform(df[i])

#getdummies
df1=pd.get_dummies(df[['carbody','cylindernumber','drivewheel','engine
type','fuelsystem']],drop_first=True)
df1=df1.astype(int)
df1
```

	carbody_hardtop	carbody_hatchback	carbody_sedan	carbody_wagon
0	0	0	0	0
1	0	0	0	0
2	0	1	0	0
3	0	0	1	0

4	0	0	1	0
..
200	0	0	1	0
201	0	0	1	0
202	0	0	1	0
203	0	0	1	0
204	0	0	1	0

	cylindernumber_five	cylindernumber_four	cylindernumber_six	\
0	0	1	0	
1	0	1	0	
2	0	0	1	
3	0	1	0	
4	1	0	0	
..	
200	0	1	0	
201	0	1	0	
202	0	0	1	
203	0	0	1	
204	0	1	0	

	cylindernumber_three	cylindernumber_twelve
cylindernumber_two	...	\
0	0	0
0	...	
1	0	0
0	...	
2	0	0
0	...	
3	0	0
0	...	
4	0	0
0	...	
..
...		
200	0	0
0	...	
201	0	0
0	...	
202	0	0
0	...	
203	0	0

0	...			
204		0		0
0	...			

	enginetype_ohcf	enginetype_ohcv	enginetype_rotor
fuelsystem_2bbl \			
0	0	0	0
0			
1	0	0	0
0			
2	0	1	0
0			
3	0	0	0
0			
4	0	0	0
0			
..
..			
200	0	0	0
0			
201	0	0	0
0			
202	0	1	0
0			
203	0	0	0
0			
204	0	0	0
0			

	fuelsystem_4bbl	fuelsystem_idi	fuelsystem_mfi	fuelsystem_mphi
\				
0	0	0	0	1
1	0	0	0	1
2	0	0	0	1
3	0	0	0	1
4	0	0	0	1
..
200	0	0	0	1
201	0	0	0	1
202	0	0	0	1
203	0	1	0	0

204	0	0	0	1
-----	---	---	---	---

	fuelsystem_spdi	fuelsystem_spfi
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0
..
200	0	0
201	0	0
202	0	0
203	0	0
204	0	0

[205 rows x 25 columns]

#concatination

df2=pd.concat([df,df1],axis=1)

df2

	car_ID	symboling	CarName	fueltype	aspiration
\					
0	1	3	alfa-romero giulia	1	0
1	2	3	alfa-romero stelvio	1	0
2	3	1	alfa-romero Quadrifoglio	1	0
3	4	2	audi 100 ls	1	0
4	5	2	audi 100ls	1	0
..
200	201	-1	volvo 145e (sw)	1	0
201	202	-1	volvo 144ea	1	1
202	203	-1	volvo 244dl	1	0
203	204	-1	volvo 246	0	1
204	205	-1	volvo 264gl	1	1

	doornumber	carbody	drivewheel	enginelocation
wheelbase	...	\		
0	1	convertible	rwd	0

88.6	...				
1		1	convertible	rwd	0
88.6	...				
2		1	hatchback	rwd	0
94.5	...				
3		0	sedan	fwd	0
99.8	...				
4		0	sedan	4wd	0
99.4	...				
..	
.					
200		0	sedan	rwd	0
109.1	...				
201		0	sedan	rwd	0
109.1	...				
202		0	sedan	rwd	0
109.1	...				
203		0	sedan	rwd	0
109.1	...				
204		0	sedan	rwd	0
109.1	...				

	enginetype_ohcf	enginetype_ohcv	enginetype_rotor
fuelsystem_2bbl	\		
0	0	0	0
0			
1	0	0	0
0			
2	0	1	0
0			
3	0	0	0
0			
4	0	0	0
0			
..
..			
200	0	0	0
0			
201	0	0	0
0			
202	0	1	0
0			
203	0	0	0
0			
204	0	0	0
0			

	fuelsystem_4bbl	fuelsystem_idi	fuelsystem_mfi	fuelsystem_mphi	\
0	0	0	0	1	

1	0	0	0	1
2	0	0	0	1
3	0	0	0	1
4	0	0	0	1
..
200	0	0	0	1
201	0	0	0	1
202	0	0	0	1
203	0	1	0	0
204	0	0	0	1

	fuelsystem_spdi	fuelsystem_spfi
0	0	0
1	0	0
2	0	0
3	0	0
4	0	0
..
200	0	0
201	0	0
202	0	0
203	0	0
204	0	0

[205 rows x 51 columns]

#dropping unwanted columns

```
df2.drop(['carbody','cylindernumber','drivewheel','enginetype','fuelsystem','car_ID','CarName'],axis=1,inplace=True)
```

df2.dtypes

symboling	int64
fueltype	int64
aspiration	int64
doornumber	int64
enginelocation	int64
wheelbase	float64
carlength	float64
carwidth	float64
carheight	float64
curbweight	int64
enginesize	int64
boreratio	float64
stroke	float64
compressionratio	float64
horsepower	int64
peakrpm	int64
citympg	int64
highwaympg	int64

```

price float64
carbody_hardtop int64
carbody_hatchback int64
carbody_sedan int64
carbody_wagon int64
cylindernumber_five int64
cylindernumber_four int64
cylindernumber_six int64
cylindernumber_three int64
cylindernumber_twelve int64
cylindernumber_two int64
drivewheel_fwd int64
drivewheel_rwd int64
enginetype_dohcv int64
enginetype_l int64
enginetype_ohc int64
enginetype_ohcf int64
enginetype_ohcv int64
enginetype_rotor int64
fuelsystem_2bbl int64
fuelsystem_4bbl int64
fuelsystem_idi int64
fuelsystem_mfi int64
fuelsystem_mphi int64
fuelsystem_spdi int64
fuelsystem_spfi int64
dtype: object

```

SEPERATING X AND Y

```

#x as input variable
x=df2.drop('price',axis=True)
x

```

	symboling	fueltype	aspiration	doornumber	enginelocation
wheelbase \					
0	3	1	0	1	0
88.6					
1	3	1	0	1	0
88.6					
2	1	1	0	1	0
94.5					
3	2	1	0	0	0
99.8					
4	2	1	0	0	0
99.4					
..
...					
200	-1	1	0	0	0

109.1					
201	-1	1	1	0	0
109.1					
202	-1	1	0	0	0
109.1					
203	-1	0	1	0	0
109.1					
204	-1	1	1	0	0
109.1					

	carlength	carwidth	carheight	curbweight	...	enginetype_ohcf
\						
0	168.8	64.1	48.8	2548	...	0
1	168.8	64.1	48.8	2548	...	0
2	171.2	65.5	52.4	2823	...	0
3	176.6	66.2	54.3	2337	...	0
4	176.6	66.4	54.3	2824	...	0
..
200	188.8	68.9	55.5	2952	...	0
201	188.8	68.8	55.5	3049	...	0
202	188.8	68.9	55.5	3012	...	0
203	188.8	68.9	55.5	3217	...	0
204	188.8	68.9	55.5	3062	...	0

	enginetype_ohcv	enginetype_rotor	fuelsystem_2bbl
fuelsystem_4bbl \			
0	0	0	0
0			
1	0	0	0
0			
2	1	0	0
0			
3	0	0	0
0			
4	0	0	0
0			
..
..			
200	0	0	0

```

0
201      0      0      0
0
202      1      0      0
0
203      0      0      0
0
204      0      0      0
0

```

```

      fuelsystem_idi  fuelsystem_mfi  fuelsystem_mpfi  fuelsystem_spdi
\
0      0      0      1      0
1      0      0      1      0
2      0      0      1      0
3      0      0      1      0
4      0      0      1      0
..      ...      ...      ...      ...
200     0      0      1      0
201     0      0      1      0
202     0      0      1      0
203     1      0      0      0
204     0      0      1      0

```

```

      fuelsystem_spfi
0      0
1      0
2      0
3      0
4      0
..      ...
200     0
201     0
202     0
203     0
204     0

```

```
[205 rows x 43 columns]
```



```
y=df2['price']
```

```
y
```

```
0      13495.0
1      16500.0
2      16500.0
3      13950.0
4      17450.0
```

```
...
200     16845.0
201     19045.0
202     21485.0
203     22470.0
204     22625.0
```

```
Name: price, Length: 205, dtype: float64
```

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=.30,random_state=42)
```

```
x_train
```

	symboling	fueltype	aspiration	doornumber	engine	location
wheelbase \						
177	-1	1	0	0		0
102.4						
75	1	1	1	1		0
102.7						
174	-1	0	1	0		0
102.4						
31	2	1	0	1		0
86.6						
12	0	1	0	1		0
101.2						
..
...						
106	1	1	0	1		0
99.2						
14	1	1	0	0		0
103.5						
92	1	1	0	0		0
94.5						
179	3	1	0	1		0
102.9						
102	0	1	0	0		0
100.4						

	carlength	carwidth	carheight	curbweight	...	engine	type_ohcf
\							
177	175.6	66.5	53.9	2458	...		0
75	178.4	68.0	54.8	2910	...		0

174	175.6	66.5	54.9	2480	...	0
31	144.6	63.9	50.8	1819	...	0
12	176.8	64.8	54.3	2710	...	0
..
106	178.5	67.9	49.7	3139	...	0
14	189.0	66.9	55.7	3055	...	0
92	165.3	63.8	54.5	1938	...	0
179	183.5	67.7	52.0	3016	...	0
102	184.6	66.5	56.1	3296	...	0

	enginetype_ohcv	enginetype_rotor	fuelsystem_2bbl
fuelsystem_4bbl \			
177	0	0	0
0			
75	0	0	0
0			
174	0	0	0
0			
31	0	0	0
0			
12	0	0	0
0			
..
..			
106	1	0	0
0			
14	0	0	0
0			
92	0	0	1
0			
179	0	0	0
0			
102	1	0	0
0			

	fuelsystem_idi	fuelsystem_mfi	fuelsystem_mpfi	fuelsystem_spdi
\				
177	0	0	1	0
75	0	0	1	0

174	1	0	0	0
31	0	0	0	0
12	0	0	1	0
..
106	0	0	1	0
14	0	0	1	0
92	0	0	0	0
179	0	0	1	0
102	0	0	1	0

```

fuelsystem_spfi
177      0
75       0
174      0
31       0
12       0
..      ...
106      0
14       0
92       0
179      0
102      0

```

[143 rows x 43 columns]

x_test

	symboling	fueltype	aspiration	doornumber	engine location
wheelbase \					
15	0	1	0	0	0
103.5					
9	0	1	1	1	0
99.5					
100	0	1	0	0	0
97.2					
132	3	1	0	1	0
99.1					
68	-1	0	1	0	0
110.0					
..
...					
56	3	1	0	1	0

95.3					
128	3	1	0	1	1
89.5					
76	2	1	0	1	0
93.7					
144	0	1	0	0	0
97.0					
104	3	1	0	1	0
91.3					

	carlength	carwidth	carheight	curbweight	...	enginetype_ohcf
\						
15	189.0	66.9	55.7	3230	...	0
9	178.2	67.9	52.0	3053	...	0
100	173.4	65.2	54.7	2302	...	0
132	186.6	66.5	56.1	2658	...	0
68	190.9	70.3	58.7	3750	...	0
..
56	169.0	65.7	49.6	2380	...	0
128	168.9	65.0	51.6	2800	...	1
76	157.3	64.4	50.8	1918	...	0
144	172.0	65.4	54.3	2385	...	1
104	170.7	67.9	49.7	3071	...	0

	enginetype_ohcv	enginetype_rotor	fuelsystem_2bbl
fuelsystem_4bbl \			
15	0	0	0
0			
9	0	0	0
0			
100	0	0	1
0			
132	0	0	0
0			
68	0	0	0
0			
..
..			
56	0	1	0

```

1
128      0      0      0
0
76      0      0      1
0
144     0      0      1
0
104     1      0      0
0

```

```

      fuelsystem_idi  fuelsystem_mfi  fuelsystem_mpfi  fuelsystem_spdi
\
15      0      0      1      0
9       0      0      1      0
100     0      0      0      0
132     0      0      1      0
68      1      0      0      0
..      ...      ...      ...      ...
56      0      0      0      0
128     0      0      1      0
76      0      0      0      0
144     0      0      0      0
104     0      0      1      0

```

```

      fuelsystem_spfi
15      0
9       0
100     0
132     0
68      0
..      ...
56      0
128     0
76      0
144     0
104     0

```

```
[62 rows x 43 columns]
```

```
y_train
```

```

177    11248.0
75     16503.0
174    10698.0
31      6855.0
12     20970.0
...
106    18399.0
14     24565.0
92      6849.0
179    15998.0
102    14399.0
Name: price, Length: 143, dtype: float64

```

```

y_test
15     30760.000
9      17859.167
100     9549.000
132    11850.000
68     28248.000
...
56     11845.000
128    37028.000
76      5389.000
144     9233.000
104    17199.000
Name: price, Length: 62, dtype: float64

```

MODEL CREATION

```

model=LinearRegression()
model.fit(x_train,y_train)

LinearRegression()

y_pred=model.predict(x_test)
y_pred
array([28789.49001435, 20620.63614303, 10439.81708869, 13080.97596562,
       26508.2560389 , 5652.87320458, 7404.34732707, 7890.50218351,
       8438.20240933, 8216.04952813, 17334.1081073 , 7184.10590596,
       16666.77788623, 8893.95047362, 43439.33332466, 6208.90429031,
       4343.7299397 , 13909.3648381 , 10627.96608776, 9946.2926437 ,
       11147.03041795, 17219.97172739, 6768.53318996, 2729.25616142,
       7122.39123972, 29343.08293322, 13780.99757454, 16430.89280338,
       4480.75941091, 17718.51563485, 27328.21316198, 6159.78288962,
       6296.97996495, 20020.84431001, 7166.41970743, 27042.81604006,
       12530.32557035, 13626.0244425 , 6590.98812919, 14245.2197143 ,
       6490.90029973, 12925.43351039, 17683.59601539, 5755.82063462,
       6581.46477642, 9223.09768198, 6159.78288962, 6036.7981726 ,

```

```
16875.9076384 , 15768.03687359, 5503.07995485, 20543.75731459,
6077.39489533, 9443.66303585, 4431.42296141, 16396.17011025,
13796.17802054, 13626.0244425 , 35197.27571501, 5962.1148521 ,
9325.34969991, 17429.3797627 ])
```

```
df3=pd.DataFrame({"Actual_value":y_test,"pedicted_value":y_pred,"diffa
rance":y_test-y_pred})
df3
```

	Actual_value	pedicted_value	diffarance
15	30760.000	28789.490014	1970.509986
9	17859.167	20620.636143	-2761.469143
100	9549.000	10439.817089	-890.817089
132	11850.000	13080.975966	-1230.975966
68	28248.000	26508.256039	1739.743961
..
56	11845.000	13626.024442	-1781.024442
128	37028.000	35197.275715	1830.724285
76	5389.000	5962.114852	-573.114852
144	9233.000	9325.349700	-92.349700
104	17199.000	17429.379763	-230.379763

```
[62 rows x 3 columns]
```

PERFORMANCE EVALUATION

```
print("MAE is:",mean_absolute_error(y_test,y_pred))
```

```
MAE is: 1889.8051330426135
```

```
print("mean_absolute_percentage_error
is:",mean_absolute_percentage_error(y_test,y_pred))
```

```
mean_absolute_percentage_error is: 0.16075239872131536
```

```
print("MSE is:",mean_squared_error(y_test,y_pred))
```

```
MSE is: 7052032.685191335
```

```
print("r2_score is:",r2_score(y_test,y_pred))
```

```
r2_score is: 0.8982161227892793
```