```
In [1]:
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

In [2]:

```
import io
%cd "C:\Users\deepe\OneDrive\Desktop\Python Datasets\Car Dekh"
```

C:\Users\deepe\OneDrive\Desktop\Python Datasets\Car Dekh

In [3]:

```
cardata=pd.read_csv("Car details v3.csv")
```

In [4]:

```
# About the data cardata.info()
```

```
RangeIndex: 8128 entries, 0 to 8127
Data columns (total 13 columns):
    Column
                    Non-Null Count
#
                                    Dtype
                    _____
0
    name
                    8128 non-null
                                    object
1
                    8128 non-null
    year
                                    int64
2
    selling_price 8128 non-null
                                    int64
3
    km_driven
                    8128 non-null
                                    int64
4
    fuel
                    8128 non-null
                                    object
5
    seller type
                   8128 non-null
                                    object
    transmission
6
                   8128 non-null
                                    object
7
    owner
                    8128 non-null
                                    object
8
    mileage
                    7907 non-null
                                    float64
9
                                    float64
    engine
                    7907 non-null
10
                    7913 non-null
                                    object
    max_power
11
                    7640 non-null
                                    object
    torque
12
    seats
                    7907 non-null
                                    float64
dtypes: float64(3), int64(3), object(7)
memory usage: 825.6+ KB
```

<class 'pandas.core.frame.DataFrame'>

In [5]:

```
# Columns of data cardata.columns
```

```
Out[5]:
```

In [6]:

cardata.head()

Out[6]:

	name	year	selling_price	km_driven	fuel	seller_type	transmission	owner	mileag
0	Maruti Swift Dzire VDI	2014	450000	145500	Diesel	Individual	Manual	First Owner	23.4
1	Skoda Rapid 1.5 TDI Ambition	2014	370000	120000	Diesel	Individual	Manual	Second Owner	21.1
2	Honda City 2017- 2020 EXi	2006	158000	140000	Petrol	Individual	Manual	Third Owner	17.7
3	Hyundai i20 Sportz Diesel	2010	225000	127000	Diesel	Individual	Manual	First Owner	23.0
4	Maruti Swift VXI BSIII	2007	130000	120000	Petrol	Individual	Manual	First Owner	16.1
4									•

In [7]:

Finding the missing values percentage in each column
cardata.isnull().sum().sort_values(ascending=False)/cardata.shape[0]

Out[7]:

torque	0.060039
mileage	0.027190
engine	0.027190
seats	0.027190
max_power	0.026452
name	0.000000
year	0.000000
selling_price	0.000000
km_driven	0.000000
fuel	0.000000
seller_type	0.000000
transmission	0.000000
owner	0.000000
dtype: float64	

In [8]:

```
# Creating new column car_age in the data
cardata["car_age"]=2023-cardata.year
```

In [9]:

```
# Removing the column year from the data
cardata=cardata.drop(["year"],axis=1)
```

In [10]:

```
cardata.info()
```

```
RangeIndex: 8128 entries, 0 to 8127
Data columns (total 13 columns):
                   Non-Null Count Dtype
#
    Column
    _____
                   -----
0
                   8128 non-null
                                   object
    name
1
    selling_price 8128 non-null
                                   int64
2
    km_driven
                   8128 non-null
                                   int64
3
    fuel
                   8128 non-null
                                   object
4
    seller_type
                   8128 non-null
                                   object
5
    transmission 8128 non-null
                                   object
6
    owner
                   8128 non-null
                                   object
7
    mileage
                   7907 non-null
                                   float64
                                  float64
8
    engine
                   7907 non-null
    max_power
9
                   7913 non-null
                                   object
10
                   7640 non-null
                                   object
    torque
11
   seats
                   7907 non-null
                                   float64
12 car age
                  8128 non-null
                                   int64
dtypes: float64(3), int64(3), object(7)
memory usage: 825.6+ KB
```

<class 'pandas.core.frame.DataFrame'>

In [11]:

```
cardata.columns
```

Out[11]:

In [12]:

```
# Splitting the data into numeric and object columns
numcols=cardata[['car_age' ,'selling_price', 'km_driven','max_power', 'torque','mileage'
```

In [13]:

```
numcols.describe()
# Here we are not finding the 'max_power' and 'torque' due to the datatypes
```

Out[13]:

	car_age	selling_price	km_driven	mileage	engine
count	8128.000000	8.128000e+03	8.128000e+03	7907.000000	7907.000000
mean	9.195989	6.382718e+05	6.981951e+04	19.418783	1458.625016
std	4.044249	8.062534e+05	5.655055e+04	4.037145	503.916303
min	3.000000	2.999900e+04	1.000000e+00	0.000000	624.000000
25%	6.000000	2.549990e+05	3.500000e+04	16.780000	1197.000000
50%	8.000000	4.500000e+05	6.000000e+04	19.300000	1248.000000
75%	12.000000	6.750000e+05	9.800000e+04	22.320000	1582.000000
max	40.000000	1.000000e+07	2.360457e+06	42.000000	3604.000000

In [14]:

```
numcols['torque']=pd.to_numeric(numcols.torque,errors='coerce')
```

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\3664150394.py:1: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

numcols['torque']=pd.to_numeric(numcols.torque,errors='coerce')

In [15]:

```
numcols['max_power']=pd.to_numeric(numcols.max_power,errors='coerce')
```

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\3198012263.py:1: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

numcols['max_power']=pd.to_numeric(numcols.max_power,errors='coerce')

In [16]:

```
numcols.describe()
# here we are finding the 'max_power' and 'torque' by changing the data type
```

Out[16]:

	car_age	selling_price	km_driven	max_power	torque	mileage	
count	8128.000000	8.128000e+03	8.128000e+03	7912.000000	7634.000000	7907.000000	79
mean	9.195989	6.382718e+05	6.981951e+04	91.517919	3089.943673	19.418783	14
std	4.044249	8.062534e+05	5.655055e+04	35.822499	923.452758	4.037145	5
min	3.000000	2.999900e+04	1.000000e+00	0.000000	500.000000	0.000000	6
25%	6.000000	2.549990e+05	3.500000e+04	68.050000	2500.000000	16.780000	11
50%	8.000000	4.500000e+05	6.000000e+04	82.000000	3000.000000	19.300000	12
75%	12.000000	6.750000e+05	9.800000e+04	102.000000	4000.000000	22.320000	15
max	40.000000	1.000000e+07	2.360457e+06	400.000000	21800.000000	42.000000	36
4							•

In [17]:

```
objcols=cardata[['fuel', 'seller_type', 'seats', 'transmission', 'owner']]
```

In [18]:

```
# missing values in numcols
numcols.isnull().sum().sort_values(ascending=False)/numcols.shape[0]
```

Out[18]:

```
torque 0.060778
mileage 0.027190
engine 0.027190
max_power 0.026575
car_age 0.000000
selling_price 0.000000
km_driven 0.000000
```

dtype: float64

In [19]:

```
# fill in the missing values by using the median
for col in numcols.columns:
    numcols[col]=numcols[col].fillna(numcols[col].median())
```

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\2985587695.py:3: Setting
WithCopyWarning:

```
A value is trying to be set on a copy of a slice from a DataFrame. 
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
numcols[col]=numcols[col].fillna(numcols[col].median())
```

In [20]:

```
# finding the frequency of objcols and finding the missing values
for col in objcols.columns:
    freq=objcols[col].value_counts(dropna=False)
    print(freq)
          4402
Diesel
Petrol
          3631
CNG
            57
LPG
            38
Name: fuel, dtype: int64
Individual
                    6766
Dealer
                    1126
Trustmark Dealer
                     236
Name: seller_type, dtype: int64
        6254
5.0
7.0
        1120
         236
8.0
NaN
         221
4.0
         133
9.0
          80
6.0
          62
10.0
          19
           2
2.0
14.0
           1
Name: seats, dtype: int64
Manual
             7078
Automatic
             1050
Name: transmission, dtype: int64
First Owner
                        5289
Second Owner
                        2105
Third Owner
                         555
Fourth & Above Owner
                         174
Test Drive Car
                           5
Name: owner, dtype: int64
In [21]:
# Merging of the levels due to very few no.
objcols.fuel=objcols.fuel.replace(["Petrol","CNG","LPG"],"PCL")
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\2720792656.py:2: Setting
WithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-doc
s/stable/user guide/indexing.html#returning-a-view-versus-a-copy (https://
pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
view-versus-a-copy)
  objcols.fuel=objcols.fuel.replace(["Petrol","CNG","LPG"],"PCL")
```

In [22]:

```
objcols.seats=objcols.seats.replace([4.0,9.0,6.0,10.0,2.0,14.0],24691014)
```

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\992892221.py:1: SettingW
ithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

objcols.seats=objcols.seats.replace([4.0,9.0,6.0,10.0,2.0,14.0],24691014)

In [23]:

```
objcols.owner=objcols.owner.replace(["Third Owner", "Fourth & Above Owner", "Test Drive Ca
```

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\1184832894.py:1: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

objcols.owner=objcols.owner.replace(["Third Owner", "Fourth & Above Owner", "Test Drive Car"], "3&4 Test")

In [24]:

```
objcols.isnull().sum().sort_values(ascending=False)/numcols.shape[0]
```

Out[24]:

 seats
 0.02719

 fuel
 0.00000

 seller_type
 0.00000

 transmission
 0.00000

 owner
 0.00000

dtype: float64

In [25]:

```
for col in objcols.columns:
    objcols[col]=objcols[col].fillna(objcols[col].value_counts().idxmax())
```

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\1716936729.py:2: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

objcols[col]=objcols[col].fillna(objcols[col].value_counts().idxmax())
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\1716936729.py:2: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

objcols[col]=objcols[col].fillna(objcols[col].value_counts().idxmax())
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\1716936729.py:2: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

objcols[col]=objcols[col].fillna(objcols[col].value_counts().idxmax())
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\1716936729.py:2: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

objcols[col]=objcols[col].fillna(objcols[col].value_counts().idxmax())
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\1716936729.py:2: Setting
WithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

objcols[col]=objcols[col].fillna(objcols[col].value counts().idxmax())

In [26]:

```
# combining the numcols and objcols
cardatadf=pd.concat([numcols,objcols],axis=1)
```

In [27]:

```
cardatadf.columns
```

Out[27]:

In [28]:

```
# EDA
# average selling price of fuel
cardatadf.selling_price.groupby(cardatadf.fuel).mean()
```

Out[28]:

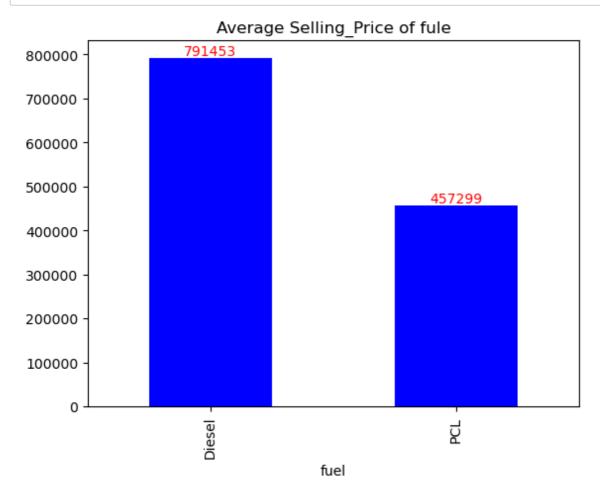
fuel

Diesel 791452.921627 PCL 457299.380569

Name: selling_price, dtype: float64

In [29]:

ax=cardatadf.selling_price.groupby(cardatadf.fuel).mean().plot(kind="bar",color="blue",t
for i in ax.containers:
 ax.bar_label(i,fontsize=10,color="Red")

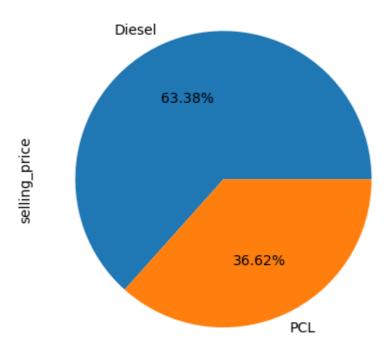


In [30]:

cardatadf.selling_price.groupby(cardatadf.fuel).mean().plot(kind="pie",autopct="%.2f%%")

Out[30]:

<Axes: ylabel='selling_price'>



In [31]:

Average selling_price of seller-type
cardatadf.selling_price.groupby(cardatadf.seller_type).mean()

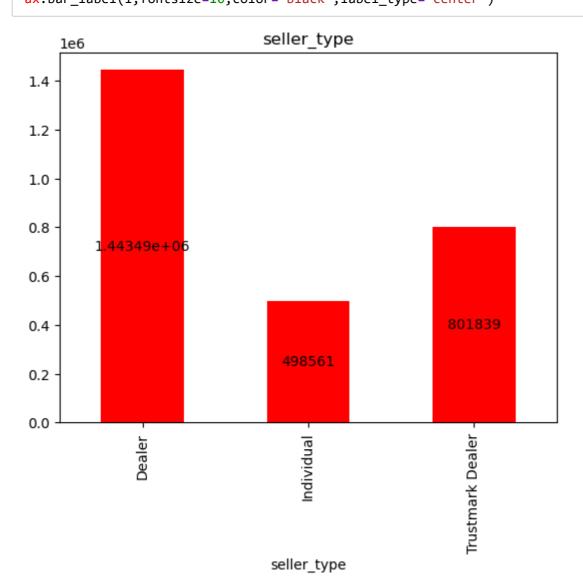
Out[31]:

seller_type

Dealer 1.443494e+06
Individual 4.985612e+05
Trustmark Dealer 8.018390e+05
Name: selling_price, dtype: float64

In [32]:

ax=cardatadf.selling_price.groupby(cardatadf.seller_type).mean().plot(kind="bar",color="
for i in ax.containers:
 ax.bar_label(i,fontsize=10,color="black",label_type="center")

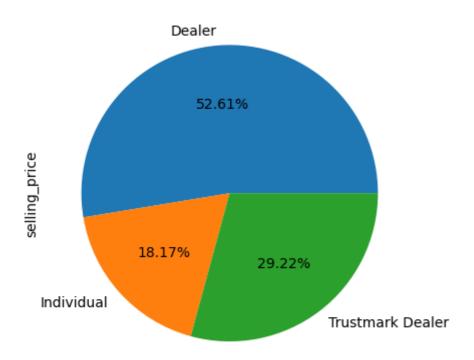


In [33]:

cardatadf.selling_price.groupby(cardatadf.seller_type).mean().plot(kind="pie",autopct="%")

Out[33]:

<Axes: ylabel='selling_price'>



In [34]:

Average selling_price of transmission
cardatadf.selling_price.groupby(cardatadf.transmission).mean()

Out[34]:

transmission

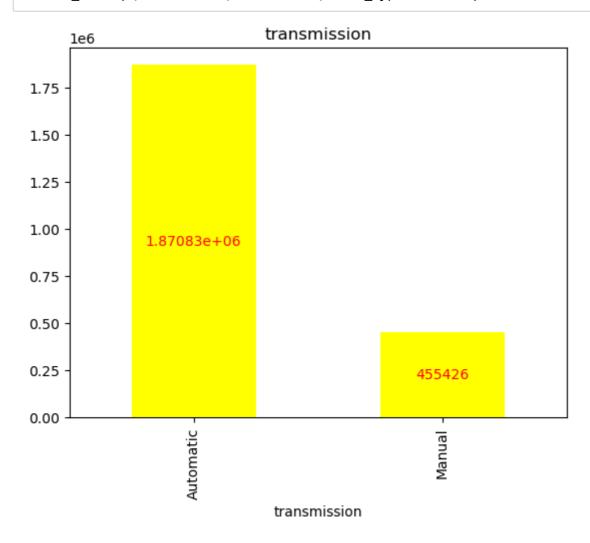
Automatic 1.870828e+06 Manual 4.554259e+05

Name: selling_price, dtype: float64

In [35]:

ax=cardatadf.selling_price.groupby(cardatadf.transmission).mean().plot(kind="bar",color=
for i in ax.containers:

ax.bar_label(i,fontsize=10,color="red",label_type="center")

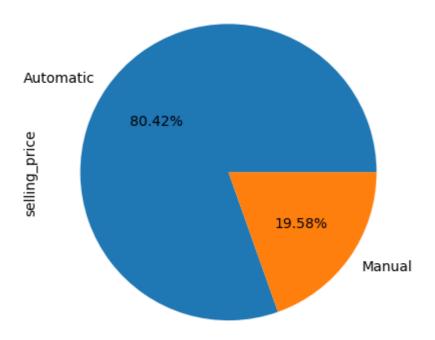


In [36]:

cardatadf.selling_price.groupby(cardatadf.transmission).mean().plot(kind="pie",autopct="

Out[36]:

<Axes: ylabel='selling_price'>



In [37]:

Average selling_price of owner
cardatadf.selling_price.groupby(cardatadf.owner).mean()

Out[37]:

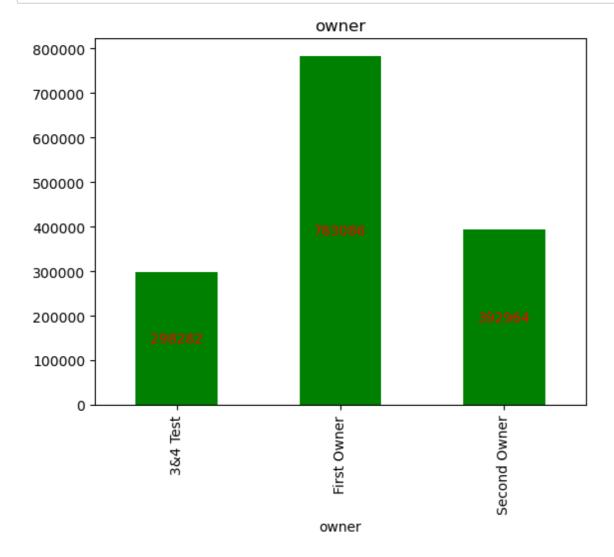
owner

3&4 Test 298282.017711 First Owner 783086.414445 Second Owner 392964.468409

Name: selling_price, dtype: float64

In [38]:

ax=cardatadf.selling_price.groupby(cardatadf.owner).mean().plot(kind="bar",color="green"
for i in ax.containers:
 ax.bar_label(i,fontsize=10,color="red",label_type="center")

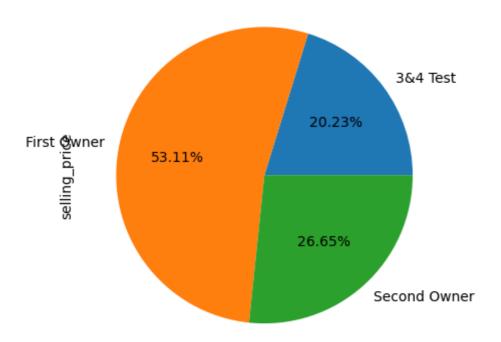


In [39]:

cardatadf.selling_price.groupby(cardatadf.owner).mean().plot(kind="pie",autopct="%.2f%"

Out[39]:

<Axes: ylabel='selling_price'>



In [40]:

Average selling_price of seats
cardatadf.selling_price.groupby(cardatadf.seats).mean()

Out[40]:

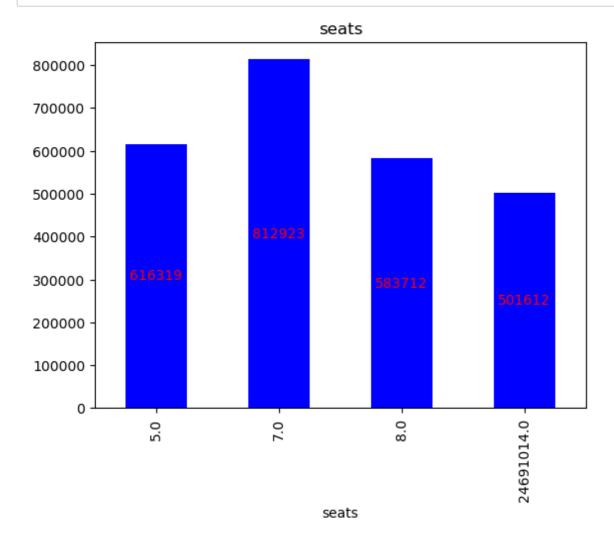
seats

5.0 616318.857606 7.0 812923.184821 8.0 583711.826271 24691014.0 501611.757576

Name: selling_price, dtype: float64

In [41]:

ax=cardatadf.selling_price.groupby(cardatadf.seats).mean().plot(kind="bar",color="blue",
for i in ax.containers:
 ax.bar_label(i,fontsize=10,color="red",label_type="center")

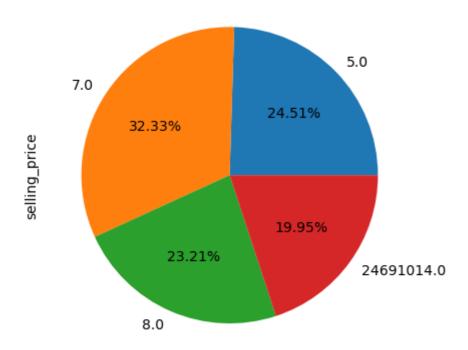


In [42]:

cardatadf.selling_price.groupby(cardatadf.seats).mean().plot(kind="pie",autopct="%.2f%%"

Out[42]:

<Axes: ylabel='selling_price'>



In [43]:

avg maileage of fule
cardatadf.mileage.groupby(cardatadf.fuel).mean()

Out[43]:

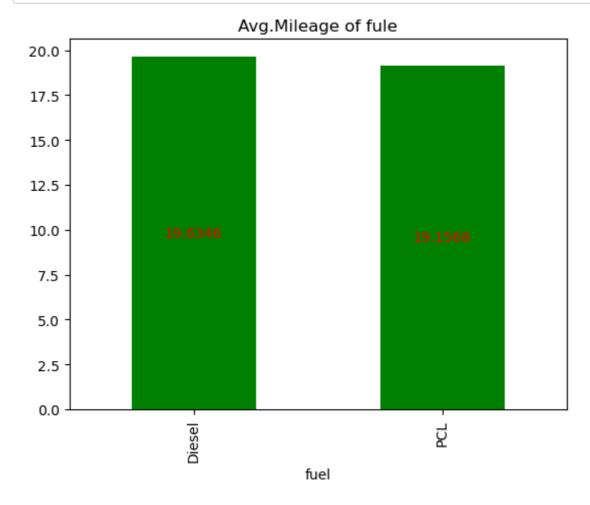
fuel

Diesel 19.634607 PCL 19.156758

Name: mileage, dtype: float64

In [44]:

ax=cardatadf.mileage.groupby(cardatadf.fuel).mean().plot(kind="bar",color="green",title=
for i in ax.containers:
 ax.bar_label(i,fontsize=10,color="red",label_type="center")

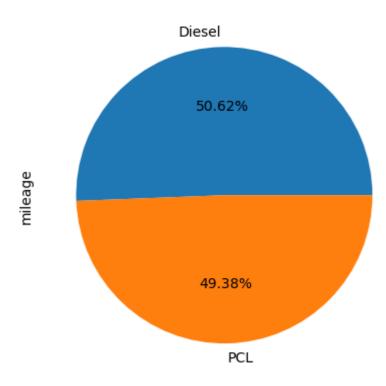


In [45]:

```
cardatadf.mileage.groupby(cardatadf.fuel).mean().plot(kind="pie",autopct="%.2f%")
```

Out[45]:

<Axes: ylabel='mileage'>



In [46]:

what is avg engine of transmission
cardatadf.engine.groupby(cardatadf.transmission).mean()

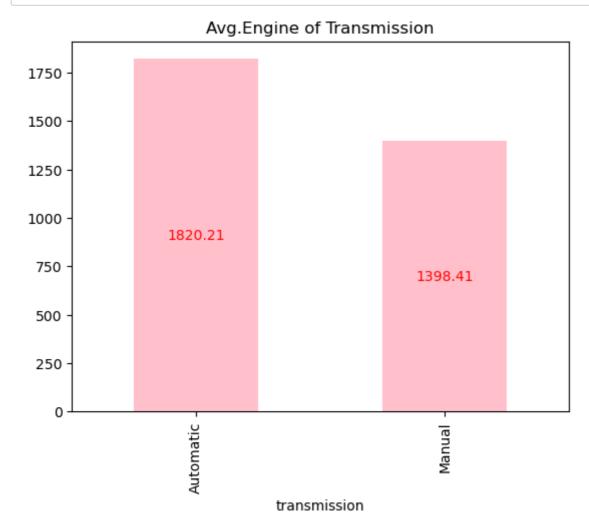
Out[46]:

transmission

Automatic 1820.206667 Manual 1398.409014 Name: engine, dtype: float64

In [47]:

ax=cardatadf.engine.groupby(cardatadf.transmission).mean().plot(kind="bar",color="pink",
for i in ax.containers:
 ax.bar_label(i,fontsize=10,color="red",label_type="center")

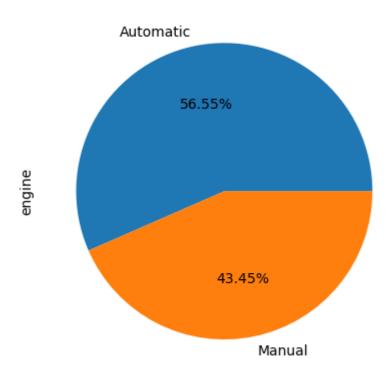


In [48]:

cardatadf.engine.groupby(cardatadf.transmission).mean().plot(kind="pie",autopct="%.2f%%"

Out[48]:

<Axes: ylabel='engine'>



In [49]:

correlation
cardatadf.corr()

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\1947867239.py:2: FutureW
arning: The default value of numeric_only in DataFrame.corr is deprecated.
In a future version, it will default to False. Select only valid columns o
r specify the value of numeric_only to silence this warning.
 cardatadf.corr()

Out[49]:

	car_age	selling_price	km_driven	max_power	torque	mileage	engine
car_age	1.000000	-0.414092	0.418006	-0.224704	-0.073480	-0.311657	-0.035851
selling_price	-0.414092	1.000000	-0.225534	0.747935	-0.034770	-0.125040	0.458345
km_driven	0.418006	-0.225534	1.000000	-0.040599	-0.206214	-0.171746	0.198911
max_power	-0.224704	0.747935	-0.040599	1.000000	-0.010493	-0.372897	0.703612
torque	-0.073480	-0.034770	-0.206214	-0.010493	1.000000	-0.186292	-0.178529
mileage	-0.311657	-0.125040	-0.171746	-0.372897	-0.186292	1.000000	-0.574128
engine	-0.035851	0.458345	0.198911	0.703612	-0.178529	-0.574128	1.000000
seats	0.132600	-0.033012	0.014551	-0.107845	-0.124052	-0.104514	0.053855
4							•

In [50]:

import seaborn as sns

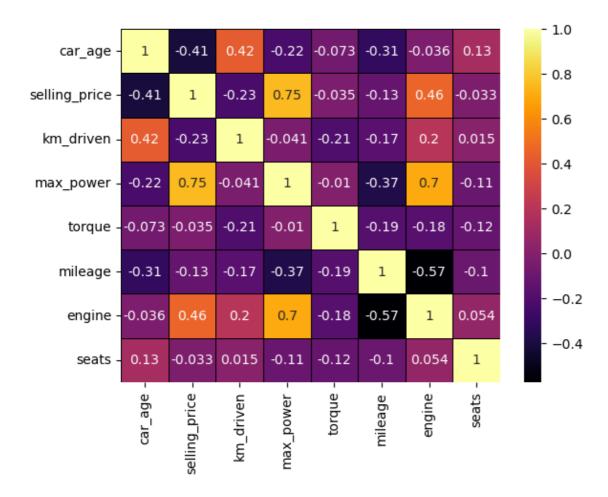
In [51]:

sns.heatmap(data=cardatadf.corr(),annot=True,cmap="inferno",linecolor="black",linewidths

C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\680232014.py:1: FutureWa
rning: The default value of numeric_only in DataFrame.corr is deprecated.
In a future version, it will default to False. Select only valid columns o
r specify the value of numeric_only to silence this warning.
 sns.heatmap(data=cardatadf.corr(),annot=True,cmap="inferno",linecolor="b
lack",linewidths=0.5)

Out[51]:

<Axes: >



In [52]:

```
# Relation between selling_price and max_power
cardatadf[["selling_price","max_power"]].corr()
```

Out[52]:

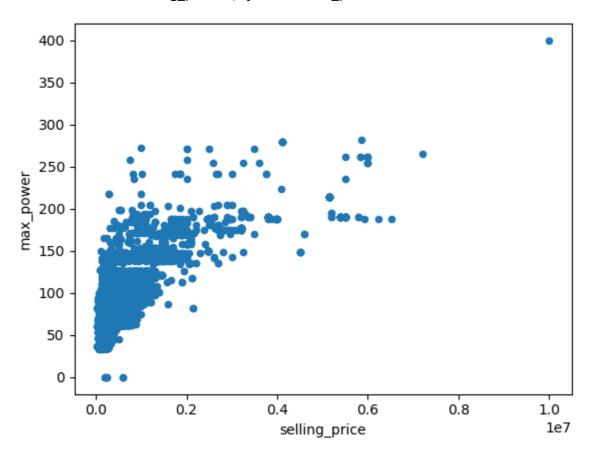
	selling_price	max_power
selling_price	1.000000	0.747935
max_power	0.747935	1.000000

In [53]:

```
cardatadf.plot(x="selling_price",y="max_power",kind="scatter")
```

Out[53]:

<Axes: xlabel='selling_price', ylabel='max_power'>



In [54]:

```
# Relation between age_car and km_driven
cardatadf[["car_age","km_driven"]].corr()
```

Out[54]:

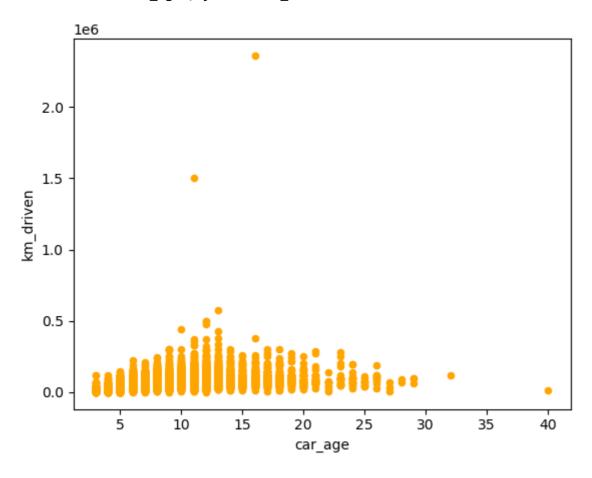
	car_age	km_driven
car_age	1.000000	0.418006
km driven	0.418006	1.000000

In [55]:

```
cardatadf.plot(x="car_age",y="km_driven",kind="scatter",color="orange")
```

Out[55]:

<Axes: xlabel='car_age', ylabel='km_driven'>



In [56]:

```
# Hypothis Testing
from scipy.stats import ttest_ind
```

In [57]:

```
cardatadf.selling_price.groupby(cardatadf.transmission).mean()
```

Out[57]:

transmission

Automatic 1.870828e+06 Manual 4.554259e+05

Name: selling_price, dtype: float64

In [58]:

```
# Spliting data
```

transmissionA=cardatadf[cardatadf.transmission=="Automatic"]
transmissionM=cardatadf[cardatadf.transmission=="Manual"]

```
22/07/2023, 16:16
                                                Car Dehk - Jupyter Notebook
  In [59]:
  ttest_ind(transmissionA.selling_price,transmissionM.selling_price,equal_var=False)
  Out[59]:
  Ttest_indResult(statistic=27.888813193478946, pvalue=8.531946079710315e-12
  In [60]:
  # What is null avg. Selling_price and fule
  cardatadf.selling price.groupby(cardatadf.fuel).mean()
  Out[60]:
  fuel
            791452.921627
  Diesel
  PCL
            457299.380569
  Name: selling_price, dtype: float64
  In [61]:
  fuleDiesel=cardatadf[cardatadf.fuel=="Diesel"]
 fulePCL=cardatadf[cardatadf.fuel=="PCL"]
  In [62]:
  ttest ind(fuleDiesel.selling price,fulePCL.selling price,equal var=False)
```

Out[62]:

Ttest_indResult(statistic=19.618295786057498, pvalue=1.1099823034116554e-8

In [63]:

```
from scipy.stats import f_oneway
```

In [64]:

```
cardatadf.selling_price.groupby(cardatadf.seats).mean()
```

Out[64]:

seats 5.0 616318.857606 7.0 812923.184821 583711.826271 8.0 24691014.0 501611.757576

Name: selling_price, dtype: float64

In [65]:

```
seats5=cardatadf[cardatadf.seats==5.0]
seats7=cardatadf[cardatadf.seats==7.0]
seats8=cardatadf[cardatadf.seats==8.0]
seats24=cardatadf[cardatadf.seats==24691014.0]
```

In [66]:

f_oneway(seats5.selling_price,seats7.selling_price,seats8.selling_price,seats24.selling_

Out[66]:

F_onewayResult(statistic=22.50039628849741, pvalue=1.6749292345739754e-14)

In [67]:

```
cardatadf.selling_price.groupby(cardatadf.seller_type).mean()
```

Out[67]:

seller_type

Dealer 1.443494e+06
Individual 4.985612e+05
Trustmark Dealer 8.018390e+05
Name: selling_price, dtype: float64

In [68]:

```
dea=cardatadf[cardatadf.seller_type=="Dealer"]
ind=cardatadf[cardatadf.seller_type=="Individual"]
tru=cardatadf[cardatadf.seller_type=="Trustmark Dealer"]
```

In [69]:

```
f_oneway(dea.selling_price,ind.selling_price,tru.selling_price)
```

Out[69]:

F_onewayResult(statistic=799.2183234565078, pvalue=1.3789446e-317)

In [70]:

```
# Cross Tabulation
# Test null no Association between seller_type and seats
pd.crosstab(cardatadf.seller_type,cardatadf.seats)
```

Out[70]:

seats 5.0 7.0 8.0 24691014.0

seller_type

Dealer	987	107	17	15
Individual	5283	982	219	282
Trustmark Dealer	205	31	0	0

```
In [71]:
```

```
from scipy.stats import chi2_contingency
```

```
In [72]:
```

```
chi2_contingency(pd.crosstab(cardatadf.seller_type,cardatadf.seats))
```

Out[72]:

```
Chi2ContingencyResult(statistic=77.45365977478873, pvalue=1.19829679202214 5e-14, dof=6, expected_freq=array([[ 897.00418307, 155.15748031, 32.693 89764, 41.14443898], [5389.9913878, 932.32283465, 196.45374016, 247.2320374], [ 188.00442913, 32.51968504, 6.8523622, 8.62352362]]))
```

In [73]:

```
# Test null no Association between seller_type and Transmission
pd.crosstab(cardatadf.seller_type,cardatadf.transmission)
```

Out[73]:

transmission Automatic Manual seller_type Dealer 459 667

Individual	492	6274
Trustmark Dealer	99	137

In [74]:

```
chi2_contingency(pd.crosstab(cardatadf.seller_type,cardatadf.transmission))
```

Out[74]:

In [75]:

```
cardatadf.selling_price.kurt()
```

Out[75]:

21.081289012186755

In [76]:

```
cardatadf.selling_price.skew()
```

Out[76]:

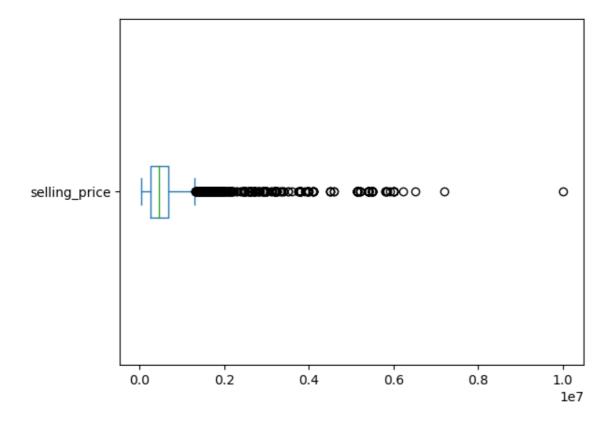
4.193533440675855

In [77]:

cardatadf.selling_price.plot(kind="box",vert=False)

Out[77]:

<Axes: >

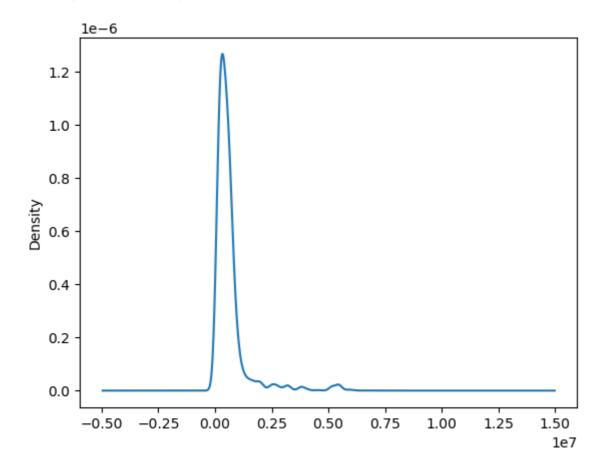


In [78]:

```
cardatadf.selling_price.plot(kind="density")
```

Out[78]:

<Axes: ylabel='Density'>

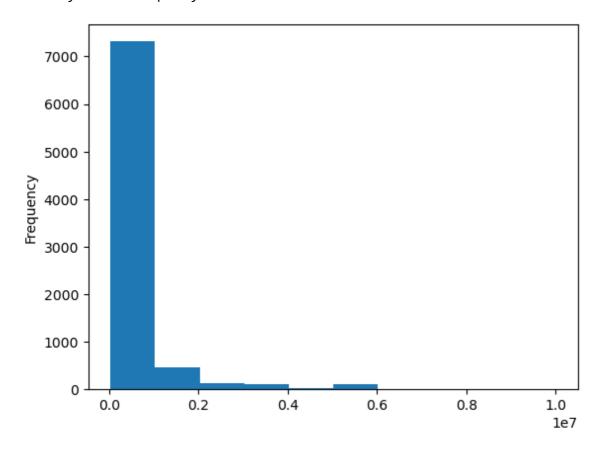


In [79]:

```
cardatadf.selling_price.plot(kind="hist")
```

Out[79]:

<Axes: ylabel='Frequency'>

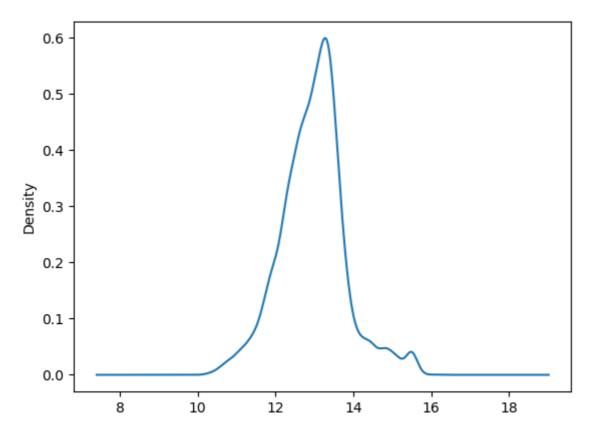


In [80]:

```
np.log(cardatadf.selling_price).plot(kind='density')
```

Out[80]:

<Axes: ylabel='Density'>



In [81]:

Here we are dropping the dependent variable
numcols=numcols.drop(["selling_price"],axis=1)

In [82]:

numcols.head()

Out[82]:

	car_age	km_driven	max_power	torque	mileage	engine
0	9	145500	74.00	2000.0	23.40	1248.0
1	9	120000	103.52	2500.0	21.14	1498.0
2	17	140000	78.00	2700.0	17.70	1497.0
3	13	127000	90.00	3000.0	23.00	1396.0
4	16	120000	88.20	4500.0	16.10	1298.0

```
In [83]:
```

Dummy Encoding and Scaling

from sklearn.preprocessing import StandardScaler

In [84]:

car_stdsca=StandardScaler()

In [85]:

numcols_stdsc=car_stdsca.fit_transform(numcols)

In [86]:

numcols_stdsc=pd.DataFrame(numcols_stdsc,columns=numcols.columns)

In [87]:

numcols_stdsc.head()

Out[87]:

	car_age	km_driven	max_power	torque	mileage	engine
0	-0.048464	1.338363	-0.488067	-1.211504	1.000696	-0.411305
1	-0.048464	0.887411	0.346439	-0.652938	0.433095	0.090536
2	1.929775	1.241098	-0.374990	-0.429512	-0.430862	0.088528
3	0.940656	1.011202	-0.035760	-0.094372	0.900235	-0.114215
4	1.682495	0.887411	-0.086645	1.581325	-0.832703	-0.310937

In [88]:

from sklearn.preprocessing import RobustScaler

In [89]:

car_rus=RobustScaler()

In [90]:

numcols_Rus=car_rus.fit_transform(numcols)

In [91]:

numcols_Rus=pd.DataFrame(numcols_Rus,columns=numcols.columns)

```
In [92]:
```

```
numcols_Rus.head()
```

Out[92]:

	car_age	km_driven	max_power	torque	mileage	engine
0	0.166667	1.357143	-0.241327	-0.666667	0.748517	0.000000
1	0.166667	0.952381	0.649170	-0.333333	0.335920	0.649351
2	1.500000	1.269841	-0.120664	-0.200000	-0.292104	0.646753
3	0.833333	1.063492	0.241327	0.000000	0.675491	0.384416
4	1.333333	0.952381	0.187029	1.000000	-0.584208	0.129870

In [93]:

```
objcols.columns
```

Out[93]:

```
Index(['fuel', 'seller_type', 'seats', 'transmission', 'owner'], dtype='ob
ject')
```

In [94]:

```
objcols_dummy=pd.get_dummies(objcols,columns=['fuel', 'seller_type', 'transmission', 'ow
```

In [95]:

```
objcols_dummy.head()
```

Out[95]:

	fuel_Diesel	fuel_PCL	seller_type_Dealer	seller_type_Individual	seller_type_Trustmark Dealer	traı
0	1	0	0	1	0	
1	1	0	0	1	0	
2	0	1	0	1	0	
3	1	0	0	1	0	
4	0	1	0	1	0	
4						•

In [96]:

```
car_clean_data=pd.concat([numcols_stdsc,objcols_dummy,cardatadf.selling_price],axis=1)
```

In [97]:

```
car_clean_data.shape
```

Out[97]:

(8128, 21)

```
In [98]:
# split data X=car claen data
X=car_clean_data.drop("selling_price",axis=1)
y=car_clean_data.selling_price
In [99]:
# Linear Regression
from sklearn.linear_model import LinearRegression
In [100]:
le=LinearRegression()
In [101]:
leregmodel=le.fit(X,y)
In [102]:
leregmodel.score(X,y)
Out[102]:
0.6886651745932868
In [103]:
predict=leregmodel.predict(X)
In [104]:
pd.DataFrame(np.exp(predict)).to_csv("reg.csv")
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\2257968620.py:1: Runtime
Warning: overflow encountered in exp
  pd.DataFrame(np.exp(predict)).to_csv("reg.csv")
In [105]:
leregmodel=le.fit(X,np.log(y))
In [106]:
leregmodel.score(X,np.log(y))
Out[106]:
0.8685987228220362
In [107]:
predict=leregmodel.predict(X)
```

```
In [108]:
pd.DataFrame(np.exp(predict)).to_csv("reglog.csv")
In [109]:
# Decision Tree
from sklearn.tree import DecisionTreeRegressor
In [110]:
DTR=DecisionTreeRegressor(max_depth=4)
In [111]:
DTR_model=DTR.fit(X,y)
In [112]:
DTR_model.score(X,y)
Out[112]:
0.9035585844165369
In [113]:
predict1=DTR_model.predict(X)
In [114]:
pd.DataFrame(np.exp(predict1)).to_csv("cardtr.csv")
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\4088083815.py:1: Runtime
Warning: overflow encountered in exp
  pd.DataFrame(np.exp(predict1)).to_csv("cardtr.csv")
In [115]:
DTR_model=DTR.fit(X,np.log(y))
In [116]:
DTR_model.score(X,np.log(y))
Out[116]:
0.8170200025834815
In [117]:
predict1=DTR_model.predict(X)
```

```
In [118]:
pd.DataFrame(np.exp(predict1)).to_csv("cardtrlog.csv")
In [119]:
# Random Forest
from sklearn.ensemble import RandomForestRegressor
In [120]:
RFR=RandomForestRegressor(n_estimators=2000)
In [121]:
RFR_model=RFR.fit(X,y)
In [122]:
RFR_model.score(X,y)
Out[122]:
0.9956802695767057
In [123]:
predict2=RFR_model.predict(X)
In [124]:
pd.DataFrame(np.exp(predict2)).to_csv("car_random.csv")
C:\Users\deepe\AppData\Local\Temp\ipykernel_13108\3624464597.py:1: Runtime
Warning: overflow encountered in exp
  pd.DataFrame(np.exp(predict2)).to_csv("car_random.csv")
In [125]:
RFR_model=RFR.fit(X,np.log(y))
In [126]:
RFR_model.score(X,np.log(y))
Out[126]:
0.9902514165617993
In [127]:
predict2=RFR_model.predict(X)
```

In [128]:				
od.DataFrame(np.exp(predict2)).to_csv("car_randomlog.csv")				
In []:				