

TASK1:-CAR PREDICTION WITH MACHINE LEARNING.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df=pd.read_csv("C:/Users/HP/Desktop/oasis/car_data.csv")
df
```

	Car_Name	Year	Selling_Price	Present_Price	Driven_kms	Fuel_Type	Selling_type	Transmission	Owner
0	ritz	2014	3.35	5.59	27000	Petrol	Dealer	Manual	0
1	sx4	2013	4.75	9.54	43000	Diesel	Dealer	Manual	0
2	ciaz	2017	7.25	9.85	6900	Petrol	Dealer	Manual	0
3	wagon r	2011	2.85	4.15	5200	Petrol	Dealer	Manual	0
4	swift	2014	4.60	6.87	42450	Diesel	Dealer	Manual	0
...
296	city	2016	9.50	11.60	33988	Diesel	Dealer	Manual	0
297	brio	2015	4.00	5.90	60000	Petrol	Dealer	Manual	0
298	city	2009	3.35	11.00	87934	Petrol	Dealer	Manual	0
299	city	2017	11.50	12.50	9000	Diesel	Dealer	Manual	0
300	brio	2016	5.30	5.90	5464	Petrol	Dealer	Manual	0

301 rows × 9 columns

```
df=pd.read_csv("C:/Users/HP/Desktop/oasis/car_data.csv")
df.head(10)
```

	Car_Name	Year	Selling_Price	Present_Price	Driven_kms	Fuel_Type	Selling_type	Transmission	Owner
0	ritz	2014	3.35	5.59	27000	Petrol	Dealer	Manual	0
1	sx4	2013	4.75	9.54	43000	Diesel	Dealer	Manual	0
2	ciaz	2017	7.25	9.85	6900	Petrol	Dealer	Manual	0
3	wagon r	2011	2.85	4.15	5200	Petrol	Dealer	Manual	0
4	swift	2014	4.60	6.87	42450	Diesel	Dealer	Manual	0
5	vitara brezza	2018	9.25	9.83	2071	Diesel	Dealer	Manual	0
6	ciaz	2015	6.75	8.12	18796	Petrol	Dealer	Manual	0
7	s cross	2015	6.50	8.61	33429	Diesel	Dealer	Manual	0
8	ciaz	2016	8.75	8.89	20273	Diesel	Dealer	Manual	0
9	ciaz	2015	7.45	8.92	42367	Diesel	Dealer	Manual	0

```
df.isnull().sum()
```

```
Car_Name      0
Year          0
Selling_Price 0
Present_Price 0
Driven_kms    0
Fuel_Type     0
Selling_type  0
Transmission  0
Owner         0
dtype: int64
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 301 entries, 0 to 300
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Car_Name        301 non-null   object
1   Year            301 non-null   int64
2   Selling_Price   301 non-null   float64
3   Present_Price   301 non-null   float64
4   Driven_kms      301 non-null   int64
5   Fuel_Type       301 non-null   object
6   Selling_type    301 non-null   object
7   Transmission    301 non-null   object
8   Owner           301 non-null   int64
dtypes: float64(2), int64(3), object(4)
memory usage: 21.3+ KB
```

```
print(df.describe())
```

	Year	Selling_Price	Present_Price	Driven_kms	Owner
count	301.000000	301.000000	301.000000	301.000000	301.000000
mean	2013.627907	4.661296	7.628472	36947.205980	0.043189
std	2.891554	5.082812	8.642584	38886.883882	0.247915
min	2003.000000	0.100000	0.320000	500.000000	0.000000
25%	2012.000000	0.900000	1.200000	15000.000000	0.000000
50%	2014.000000	3.600000	6.400000	32000.000000	0.000000
75%	2016.000000	6.000000	9.900000	48767.000000	0.000000
max	2018.000000	35.000000	92.600000	500000.000000	3.000000

```
print(df["Transmission"].describe())
```

```
count      301
unique       2
top      Manual
freq        261
Name: Transmission, dtype: object
```

```
print(df["Fuel_Type"].describe())
```

```
count          301
unique           3
top      Petrol
freq           239
Name: Fuel_Type, dtype: object
```

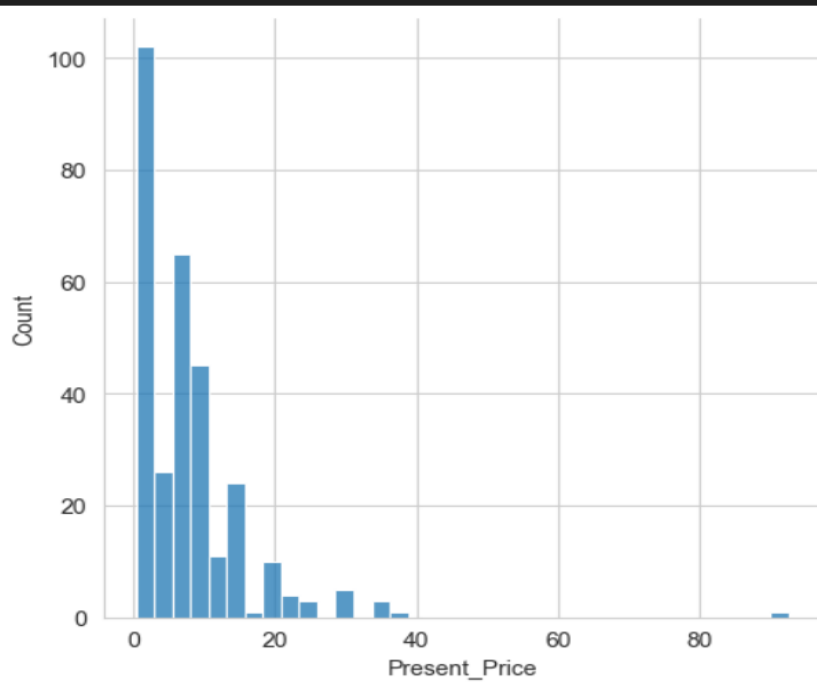
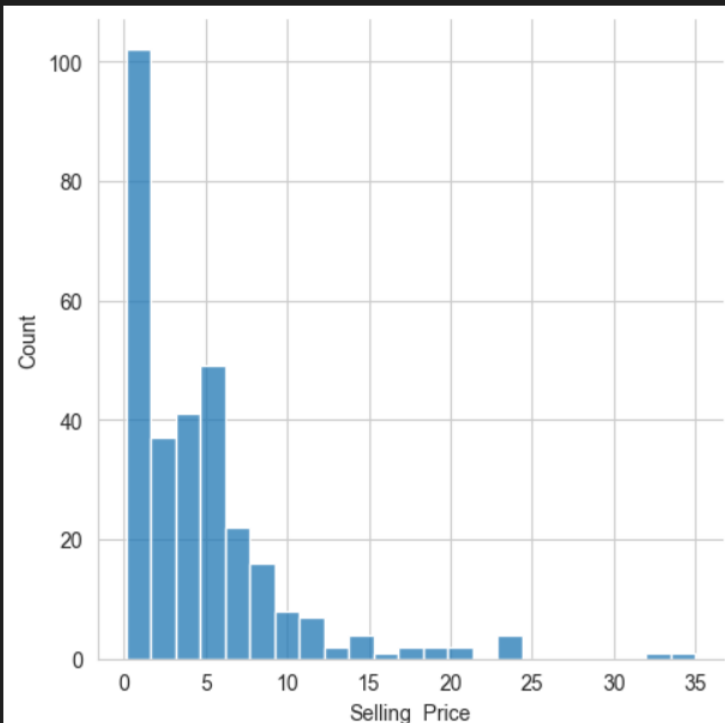
```
df.Car_Name.unique()
```

```
array(['ritz', 'sx4', 'ciaz', 'wagon r', 'swift', 'vitara brezza',
      's cross', 'alto 800', 'ertiga', 'dzire', 'alto k10', 'ignis',
      '800', 'baleno', 'omni', 'fortuner', 'innova', 'corolla altis',
      'etios cross', 'etios g', 'etios liva', 'corolla', 'etios gd',
      'camry', 'land cruiser', 'Royal Enfield Thunder 500',
      'UM Renegade Mojave', 'KTM RC200', 'Bajaj Dominar 400',
      'Royal Enfield Classic 350', 'KTM RC390', 'Hyosung GT250R',
      'Royal Enfield Thunder 350', 'KTM 390 Duke ',
      'Mahindra Mojo XT300', 'Bajaj Pulsar RS200',
      'Royal Enfield Bullet 350', 'Royal Enfield Classic 500',
      'Bajaj Avenger 220', 'Bajaj Avenger 150', 'Honda CB Hornet 160R',
      'Yamaha FZ S V 2.0', 'Yamaha FZ 16', 'TVS Apache RTR 160',
      'Bajaj Pulsar 150', 'Honda CBR 150', 'Hero Extreme',
      'Bajaj Avenger 220 dtsi', 'Bajaj Avenger 150 street',
      'Yamaha FZ v 2.0', 'Bajaj Pulsar NS 200', 'Bajaj Pulsar 220 F',
      'TVS Apache RTR 180', 'Hero Passion X pro', 'Bajaj Pulsar NS 200',
      'Yamaha Fazer ', 'Honda Activa 4G', 'TVS Sport ',
      'Honda Dream Yuga ', 'Bajaj Avenger Street 220',
      'Hero Splender iSmart', 'Activa 3g', 'Hero Passion Pro',
      'Honda CB Trigger', 'Yamaha FZ S ', 'Bajaj Pulsar 135 LS',
      'Activa 4g', 'Honda CB Unicorn', 'Hero Honda CBZ extreme',
      'Honda Karizma', 'Honda Activa 125', 'TVS Jupyter',
      'Hero Honda Passion Pro', 'Hero Splender Plus', 'Honda CB Shine',
      'Bajaj Discover 100', 'Suzuki Access 125', 'TVS Wego',
      'Honda CB twister', 'Hero Glamour', 'Hero Super Splendor',
      'Bajaj Discover 125', 'Hero Hunk', 'Hero Ignitor Disc',
      'Hero CBZ Xtreme', 'Bajaj ct 100', 'i20', 'grand i10', 'i10',
      'eon', 'xcent', 'elantra', 'creta', 'verna', 'city', 'brio',
      'amaze', 'jazz'], dtype=object)
```

+

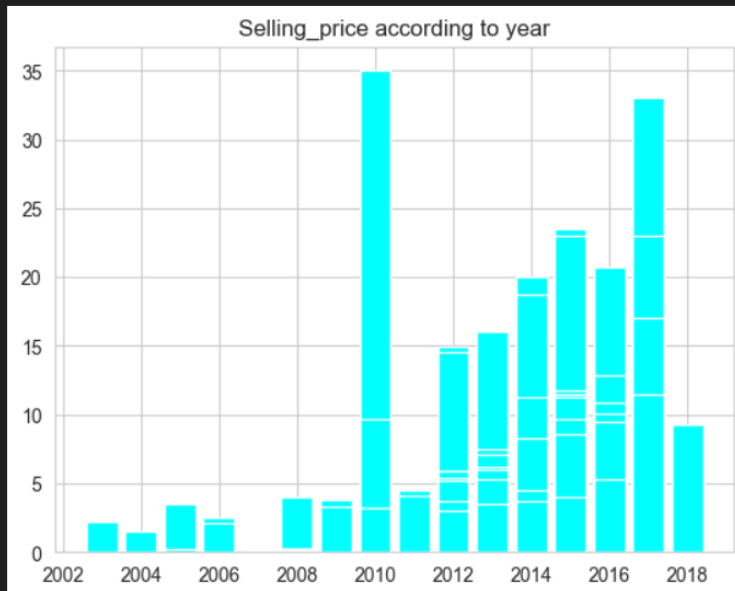
```
sns.set_style("whitegrid")
plt.figure(figsize=(10,15))
sns.displot(df.Selling_Price)
sns.displot(df.Present_Price)
plt.show()
```

<Figure size 1000x1500 with 0 Axes>



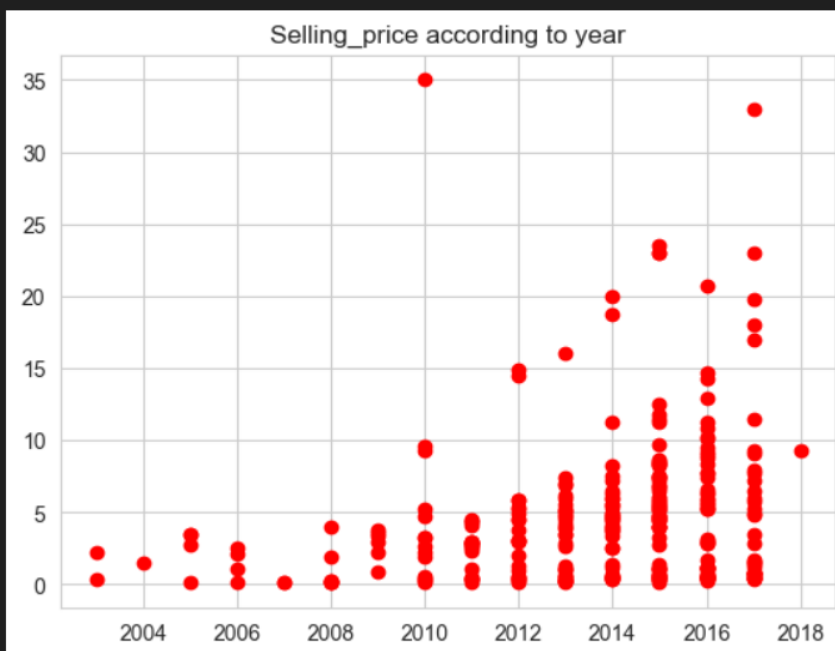
```
plt.bar(df["Year"],df["Selling_Price"],color='cyan')
plt.title("Selling_price according to year")
```

```
Text(0.5, 1.0, 'Selling_price according to year')
```



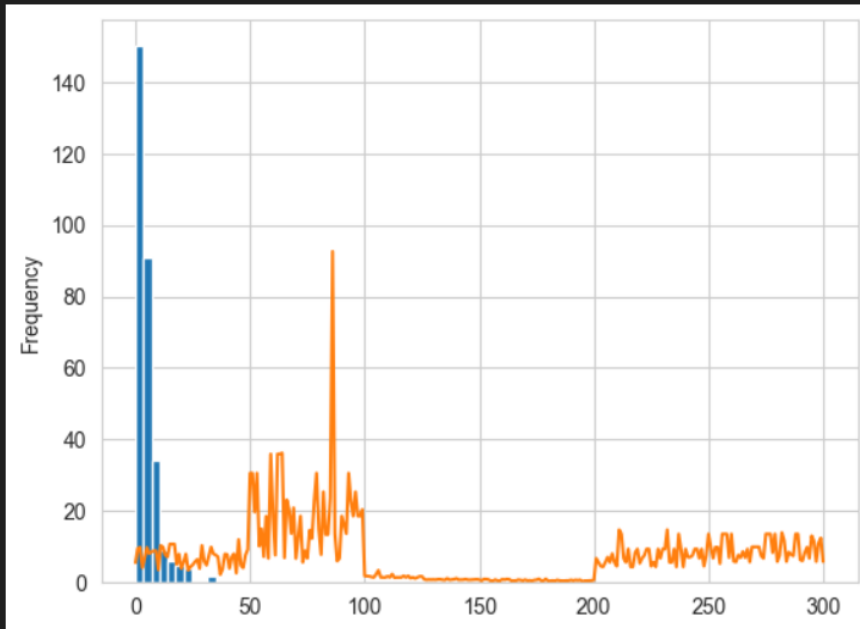
```
plt.scatter(df["Year"],df["Selling_Price"],color='red')
plt.title("Selling_price according to year")
```

```
Text(0.5, 1.0, 'Selling_price according to year')
```



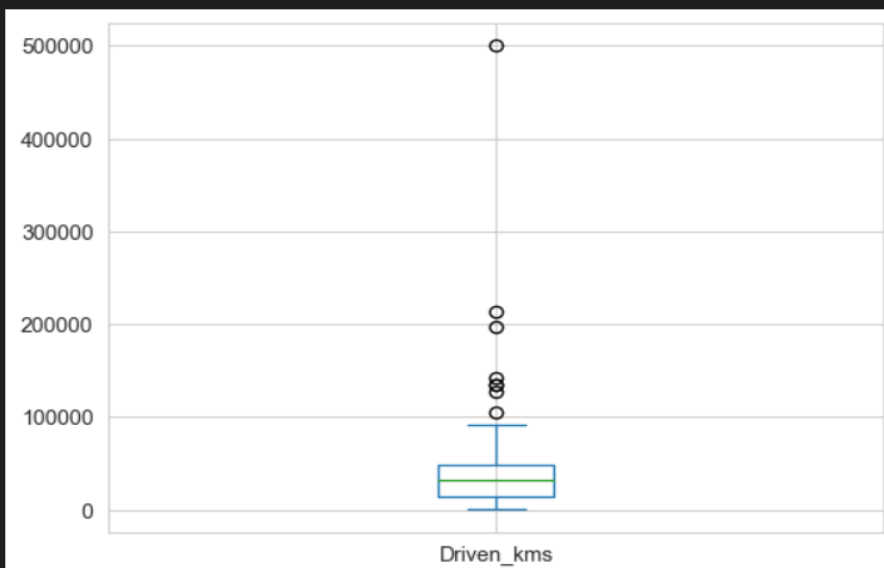
```
df["Selling_Price"].plot.hist()  
df["Present_Price"].plot.line()
```

<AxesSubplot:ylabel='Frequency'>



```
df["Driven_kms"].plot.box()
```

<AxesSubplot:>



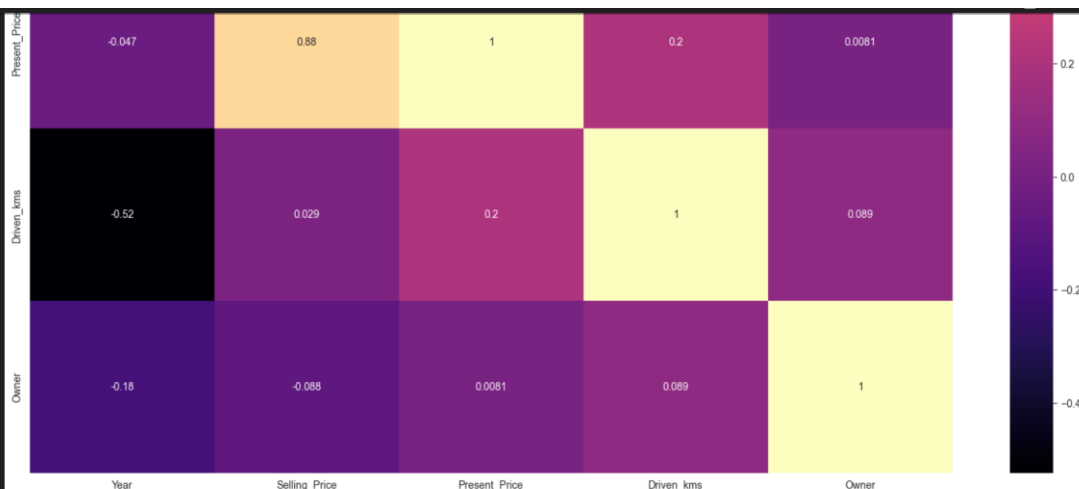
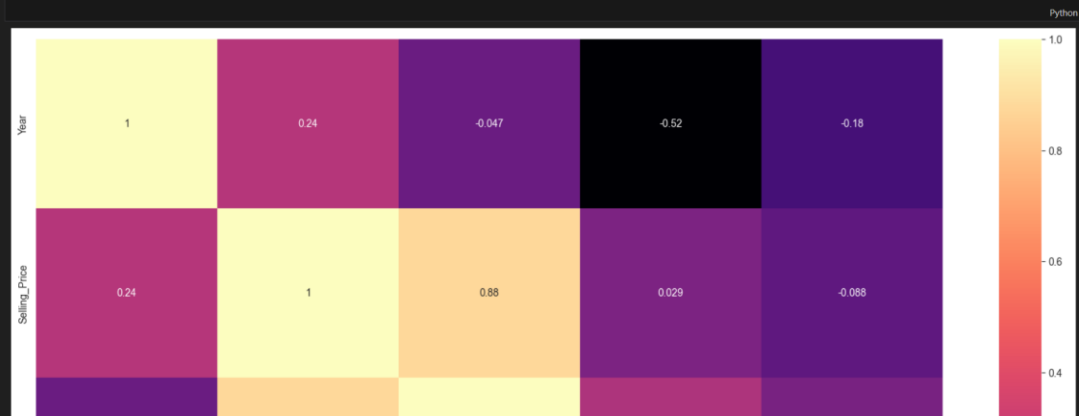
```
print(df)
```

	Car_Name	Year	Selling_Price	Present_Price	Driven_kms	Fuel_Type	\
0	ritz	2014	3.35	5.59	27000	Petrol	
1	sx4	2013	4.75	9.54	43000	Diesel	
2	ciaz	2017	7.25	9.85	6900	Petrol	
3	wagon r	2011	2.85	4.15	5200	Petrol	
4	swift	2014	4.60	6.87	42450	Diesel	
..	
296	city	2016	9.50	11.60	33988	Diesel	
297	brio	2015	4.00	5.90	60000	Petrol	
298	city	2009	3.35	11.00	87934	Petrol	
299	city	2017	11.50	12.50	9000	Diesel	
300	brio	2016	5.30	5.90	5464	Petrol	

	Selling_type	Transmission	Owner
0	Dealer	Manual	0
1	Dealer	Manual	0
2	Dealer	Manual	0
3	Dealer	Manual	0
4	Dealer	Manual	0
..
296	Dealer	Manual	0
297	Dealer	Manual	0
298	Dealer	Manual	0
299	Dealer	Manual	0
300	Dealer	Manual	0

[301 rows x 9 columns]

```
plt.figure(figsize=(20,15))
correlations=df.corr()
sns.heatmap(correlations,cmap="magma",annot=True)
plt.show()
```



```
x=df.drop(columns=["Present_Price"])
y=df["Present_Price"]
```

```
import sklearn
```

```
categorical=df.select_dtypes(include=["object"])
```

```
dummy=pd.get_dummies(categorical,drop_first=True)
x=df.drop(list(categorical.columns),axis=1)
x=pd.concat([x,dummy],axis=1)
```

```
from sklearn.model_selection import train_test_split as tts
x_train,x_test,y_train,y_test=tts(x,y,random_state=101)
```

```
from sklearn.tree import DecisionTreeRegressor
model = DecisionTreeRegressor()
model.fit(x_train, y_train)
predictions = model.predict(x_test)
from sklearn.metrics import mean_absolute_error
model.score(x_test, predictions)
```

1.0