

```
In [2]: #reading the data
import pandas as pd
car_info= pd.read_csv("car%20data.csv")
```

```
In [3]: car_info.head()
```

```
Out[3]:
```

	Car_Name	Year	Selling_Price	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	1
0	ritz	2014	3.35	5.59	27000	Petrol	Dealer	
1	sx4	2013	4.75	9.54	43000	Diesel	Dealer	
2	ciaz	2017	7.25	9.85	6900	Petrol	Dealer	
3	wagon r	2011	2.85	4.15	5200	Petrol	Dealer	
4	swift	2014	4.60	6.87	42450	Diesel	Dealer	

```
In [4]: #check data shape
car_info.shape
```

```
Out[4]: (301, 9)
```

```
In [5]: #check null value
car_info.isnull().sum()
```

```
Out[5]: Car_Name      0
Year      0
Selling_Price  0
Present_Price  0
Kms_Driven  0
Fuel_Type    0
Seller_Type  0
Transmission  0
Owner        0
dtype: int64
```

```
In [6]: car_info.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   Kms_Driven           301 non-null    int64
5   Fuel_Type            301 non-null    object
6   Seller_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
```

```
In [7]: #mathamatics infomation
car_info.describe()
```

```
Out[7]:
```

	Year	Selling_Price	Present_Price	Kms_Driven	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	8.644115	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

```
In [8]: #remove case sensitive (capital/small)
car_info.columns = [cols.lower() for cols in car_info.columns]
```

```
In [9]: car_info.head()
```

```
Out[9]:
```

	car_name	year	selling_price	present_price	kms_driven	fuel_type	seller_type	tra
0	ritz	2014	3.35	5.59	27000	Petrol	Dealer	
1	sx4	2013	4.75	9.54	43000	Diesel	Dealer	
2	ciaz	2017	7.25	9.85	6900	Petrol	Dealer	
3	wagon r	2011	2.85	4.15	5200	Petrol	Dealer	
4	swift	2014	4.60	6.87	42450	Diesel	Dealer	

```
In [10]: #none mathamatical value (catagical data)
print(car_info["fuel_type"].value_counts())
print(car_info["seller_type"].value_counts())
print(car_info["transmission"].value_counts())
```

```
fuel_type
Petrol      239
Diesel       60
CNG          2
Name: count, dtype: int64
seller_type
Dealer       195
Individual   106
Name: count, dtype: int64
transmission
Manual       261
Automatic     40
Name: count, dtype: int64
```

```
In [11]: #create catagrical data graph
import matplotlib.pyplot as plt
from matplotlib import style
style.available
```

```
Out[11]: ['Solarize_Light2',
'_classic_test_patch',
'_mpl-gallery',
'_mpl-gallery-nogrid',
'bmh',
'classic',
'dark_background',
'fast',
'fivethirtyeight',
'ggplot',
'grayscale',
'seaborn-v0_8',
'seaborn-v0_8-bright',
'seaborn-v0_8-colorblind',
'seaborn-v0_8-dark',
'seaborn-v0_8-dark-palette',
'seaborn-v0_8-darkgrid',
'seaborn-v0_8-deep',
'seaborn-v0_8-muted',
'seaborn-v0_8-notebook',
'seaborn-v0_8-paper',
'seaborn-v0_8-pastel',
'seaborn-v0_8-poster',
'seaborn-v0_8-talk',
'seaborn-v0_8-ticks',
'seaborn-v0_8-white',
'seaborn-v0_8-whitegrid',
'tableau-colorblind10']
```

```
In [12]: fuel_info = car_info["fuel_type"]
seller_info = car_info["seller_type"]
transmission_info = car_info["transmission"]
selling_price = car_info["selling_price"]
```

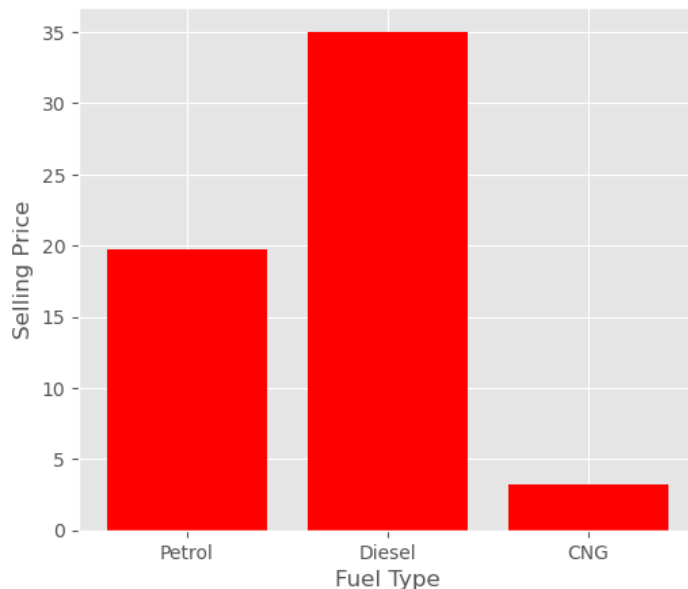
```
In [13]: style.use("ggplot")
chart = plt.figure(figsize=(20, 5))
chart.suptitle("Categorical chart info")

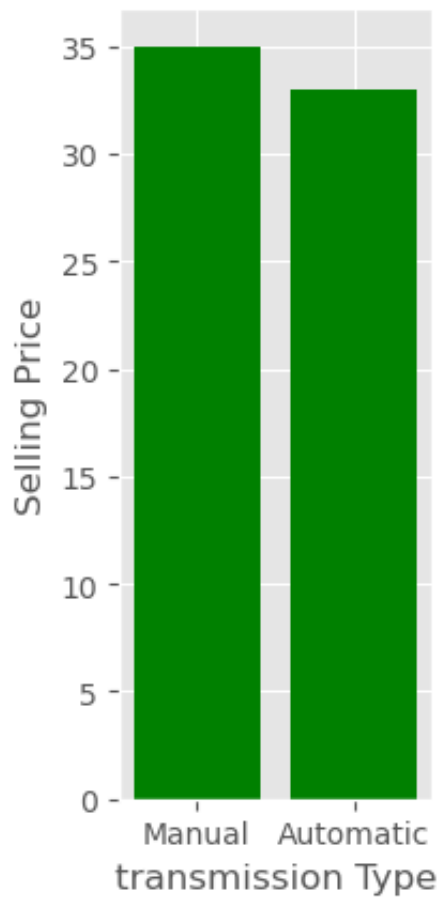
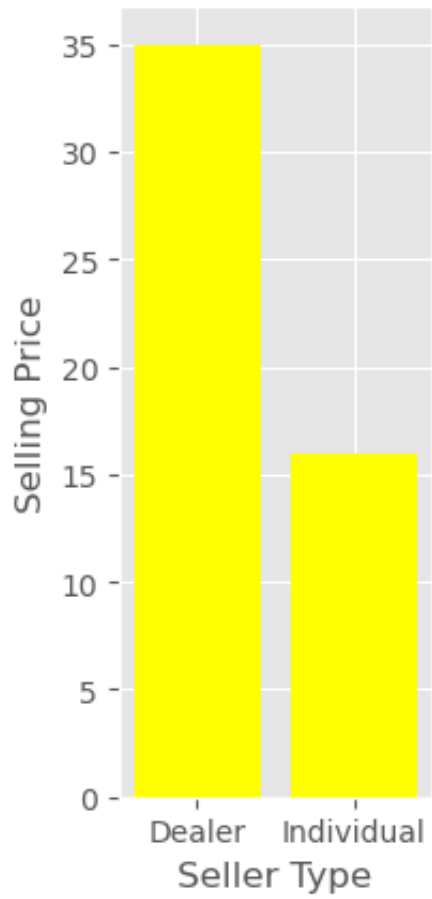
plt.subplot(1, 3, 1)
plt.bar(fuel_info, selling_price, color="Red")
plt.xlabel("Fuel Type")
plt.ylabel("Selling Price")
plt.show()

plt.subplot(1, 3, 2)
plt.bar(seller_info, selling_price, color="yellow")
plt.xlabel("Seller Type")
plt.ylabel("Selling Price")
plt.show()

plt.subplot(1, 3, 3)
plt.bar(transmission_info, selling_price, color="green")
plt.xlabel("transmission Type")
plt.ylabel("Selling Price")
plt.show()
```

Categorical chart info





```
In [14]: petrol_data = car_info.groupby("fuel_type").get_group("Petrol")
petrol_data.describe()
```

```
Out[14]:
```

	year	selling_price	present_price	kms_driven	owner
count	239.000000	239.000000	239.000000	239.000000	239.000000
mean	2013.539749	3.264184	5.583556	33528.937238	0.050209
std	3.042674	3.135537	5.290685	40308.984886	0.270368
min	2003.000000	0.100000	0.320000	500.000000	0.000000
25%	2012.000000	0.600000	0.940000	13850.000000	0.000000
50%	2014.000000	2.650000	4.600000	25870.000000	0.000000
75%	2016.000000	5.200000	7.980000	44271.000000	0.000000
max	2017.000000	19.750000	23.730000	500000.000000	3.000000

```
In [15]: Diesel_data = car_info.groupby("fuel_type").get_group("Diesel")
Diesel_data.describe()
```

```
Out[15]:
```

	year	selling_price	present_price	kms_driven	owner
count	60.000000	60.000000	60.000000	60.000000	60.000000
mean	2014.000000	10.278500	15.814500	50369.916667	0.016667
std	2.201694	7.185159	13.484289	30021.446979	0.129099
min	2005.000000	3.100000	5.700000	2071.000000	0.000000
25%	2013.000000	5.137500	8.912500	38750.000000	0.000000
50%	2014.000000	7.750000	10.585000	45000.000000	0.000000
75%	2015.000000	12.600000	17.010000	59250.000000	0.000000
max	2018.000000	35.000000	92.600000	197176.000000	1.000000

```
In [16]: cng_data = car_info.groupby("fuel_type").get_group("CNG")
cng_data.describe()
```

Out[16]:

	year	selling_price	present_price	kms_driven	owner
count	2.000000	2.000000	2.000000	2.000000	2.0
mean	2013.000000	3.100000	6.415000	42749.000000	0.0
std	2.828427	0.212132	1.873833	10251.634114	0.0
min	2011.000000	2.950000	5.090000	35500.000000	0.0
25%	2012.000000	3.025000	5.752500	39124.500000	0.0
50%	2013.000000	3.100000	6.415000	42749.000000	0.0
75%	2014.000000	3.175000	7.077500	46373.500000	0.0
max	2015.000000	3.250000	7.740000	49998.000000	0.0

convert letter to numerical value

(assign the numbers eg: Petrol-1,diseal-2,.....)

```
In [17]: print(car_info["fuel_type"], car_info["seller_type"], car_info["transmiss
```

```

0      Petrol
1      Diesel
2      Petrol
3      Petrol
4      Diesel
...
296    Diesel
297    Petrol
298    Petrol
299    Diesel
300    Petrol
Name: fuel_type, Length: 301, dtype: object 0      Dealer
1      Dealer
2      Dealer
3      Dealer
4      Dealer
...
296    Dealer
297    Dealer
298    Dealer
299    Dealer
300    Dealer
Name: seller_type, Length: 301, dtype: object 0      Manual
1      Manual
2      Manual
3      Manual
4      Manual
...
296    Manual
297    Manual
298    Manual
299    Manual
300    Manual
Name: transmission, Length: 301, dtype: object

```

```

In [18]: car_info.replace({"fuel_type":{"Petrol":0, "Diesel":1, "CNG":2}}, inplace=True)
car_info.replace({"seller_type":{"Dealer":1, "Individual":2}}, inplace=True)
car_info.replace({"transmission":{"Manual":1, "Automatic":2}}, inplace=True)

print(car_info["fuel_type"], car_info["seller_type"], car_info["transmission"])

```



```

0      0
1      1
2      0
3      0
4      1
..
296    1
297    0
298    0
299    1
300    0
Name: fuel_type, Length: 301, dtype: int64 0      1
1      1
2      1
3      1
4      1
..
296    1
297    1
298    1
299    1
300    1
Name: seller_type, Length: 301, dtype: int64 0      1
1      1
2      1
3      1
4      1
..
296    1
297    1
298    1
299    1
300    1
Name: transmission, Length: 301, dtype: int64

```

In [19]: `car_info.describe()`

Out[19]:

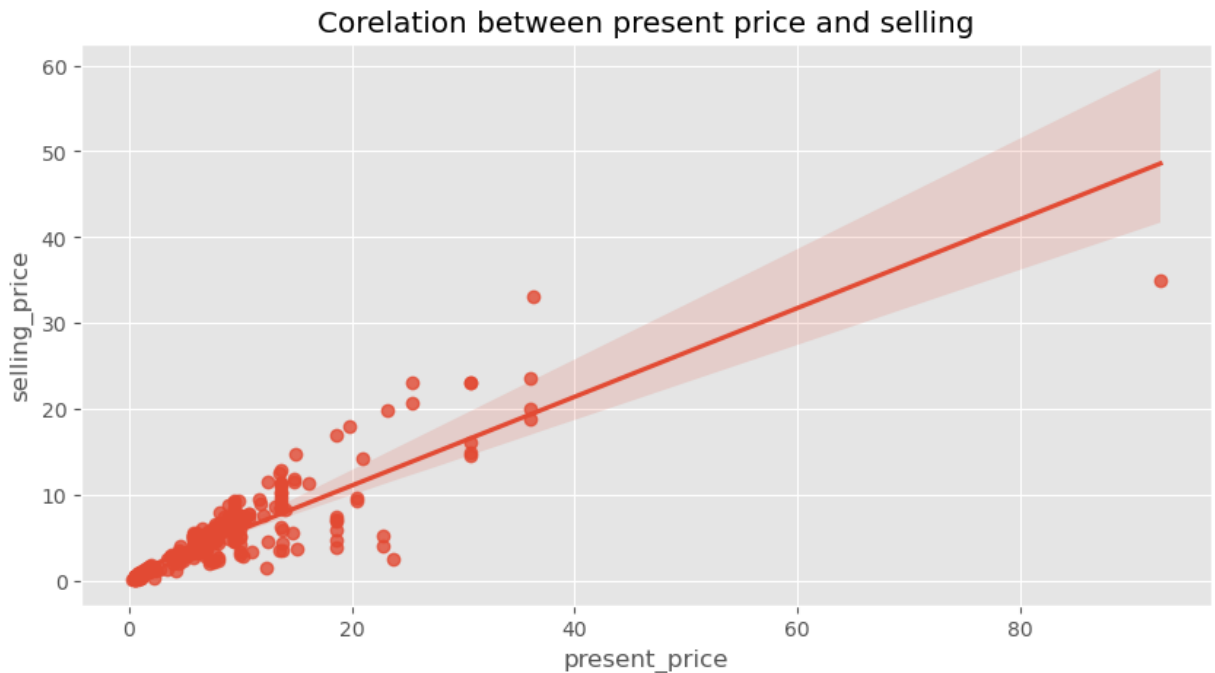
	year	selling_price	present_price	kms_driven	fuel_type	seller_typ
count	301.000000	301.000000	301.000000	301.000000	301.000000	301.000000
mean	2013.627907	4.661296	7.628472	36947.205980	0.212625	1.35215
std	2.891554	5.082812	8.644115	38886.883882	0.425801	0.47843
min	2003.000000	0.100000	0.320000	500.000000	0.000000	1.00000
25%	2012.000000	0.900000	1.200000	15000.000000	0.000000	1.00000
50%	2014.000000	3.600000	6.400000	32000.000000	0.000000	1.00000
75%	2016.000000	6.000000	9.900000	48767.000000	0.000000	2.00000
max	2018.000000	35.000000	92.600000	500000.000000	2.000000	2.00000

In []:

In []:

```
In [20]: import seaborn as sea
fig = plt.figure(figsize=(10,5))
plt.title("Corelation between present price and selling")
sea.regplot(x = "present_price" , y = "selling_price" ,data = car_info)
```

Out[20]: <Axes: title={'center': 'Corelation between present price and selling'},
xlabel='present_price', ylabel='selling_price'>



```
In [21]: input_data = car_info.drop(columns = ["car_name" , "selling_price"])
out_data = car_info ["selling_price"]
```

```
In [22]: print (input_data.shape)
print (out_data.shape)
```

```
(301, 7)
(301,)
```

```
In [23]: print (input_data)
print (out_data)
```

```

n \
0   2014          5.59      27000          0          1
1
1   2013          9.54      43000          1          1
1
2   2017          9.85       6900          0          1
1
3   2011          4.15       5200          0          1
1
4   2014          6.87      42450          1          1
1
..   ...          ...          ...          ...          ...
...
296 2016         11.60      33988          1          1
1
297 2015          5.90      60000          0          1
1
298 2009         11.00      87934          0          1
1
299 2017         12.50       9000          1          1
1
300 2016          5.90       5464          0          1
1

```

```

owner

```

```

0      0
1      0
2      0
3      0
4      0
..     ...
296    0
297    0
298    0
299    0
300    0

```

```

[301 rows x 7 columns]

```

```

0      3.35
1      4.75
2      7.25
3      2.85
4      4.60
...
296    9.50
297    4.00
298    3.35
299   11.50
300    5.30

```

```

Name: selling_price, Length: 301, dtype: float64

```

change number simliar range of number (StandardScaler)

```
In [24]: from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
input_data = scaler.fit_transform(input_data)
print (input_data)

[[ 0.128897 -0.23621461 -0.25622446 ... -0.73728539 -0.39148015
  -0.17450057]
 [-0.21751369  0.22150462  0.1559105 ... -0.73728539 -0.39148015
  -0.17450057]
 [ 1.16812909  0.25742689 -0.77396901 ... -0.73728539 -0.39148015
  -0.17450057]
 ...
 [-1.60315648  0.39068691  1.31334003 ... -0.73728539 -0.39148015
  -0.17450057]
 [ 1.16812909  0.56450434 -0.7198763 ... -0.73728539 -0.39148015
  -0.17450057]
 [ 0.8217184 -0.20029235 -0.81095812 ... -0.73728539 -0.39148015
  -0.17450057]]
```

Model test (allocation datas test,train)

```
In [25]: from sklearn.model_selection import train_test_split
input_data_train , input_data_test , output_data_train , output_data_test
```

```
In [26]: print(input_data_train.shape)
print(output_data_train.shape)
print(input_data_test.shape)
print(output_data_test.shape)

(210, 7)
(210,)
(91, 7)
(91,)
```

```
In [27]: from sklearn.linear_model import LinearRegression
from sklearn import metrics
model = LinearRegression()
model.fit(input_data_train , output_data_train)
```

```
Out[27]: ▼ LinearRegression ⓘ ?
LinearRegression()
```

predict the data

```
In [28]: predited_sellingprice = model.predict(input_data_test)
```

commpare predit value and assign value

```
In [29]: from sklearn.metrics import mean_absolute_error , mean_squared_error , r2
```

```
In [30]: print("Mean Absolute Error:" , (metrics.mean_absolute_error(predited_sell  
print("Mean Squared Error:" , (metrics.mean_squared_error(predited_sellin  
print("R2 Score:" , (metrics.r2_score(predited_sellingprice, output_data_
```

Mean Absolute Error: 1.4492978940813612

Mean Squared Error: 5.635334450464463

R2 Score: 0.8508873413267803

```
In [35]: from sklearn.linear_model import LinearRegression  
import joblib  
car_model = LinearRegression()  
car_model.fit(input_data,out_data)  
joblib.dump(car_model , "Car_Selling_Price-Identifier")
```

```
Out[35]: ['Car_Selling_Price-Identifier']
```

```
In [ ]:
```

```
In [ ]:
```