```
In [58]:
         import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
         dataset=pd.read_csv('car data.csv')
In [59]:
In [60]:
         dataset.head(2)
Out[60]:
             Car Name
                       Year Present_Price Kms_Driven Fuel_Type Seller_Type Transmission Owne
          0
                   ritz 2014
                                    5.59
                                              27000
                                                       Petrol
                                                                  Dealer
                                                                             Manual
           1
                                    9.54
                                              43000
                      2013
                                                       Diesel
                                                                  Dealer
                                                                             Manual
                  sx4
In [61]: |dataset.isnull().sum()
Out[61]: Car_Name
                           0
                           0
          Year
          Present_Price
                           0
          Kms_Driven
                           0
          Fuel_Type
                           0
          Seller_Type
                           0
          Transmission
                           0
          Owner
                           0
          Selling Price
          dtype: int64
In [62]: dataset.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
                              301 non-null
                                               int64
               Year
           2
               Present_Price
                              301 non-null
                                               float64
           3
               Kms_Driven
                              301 non-null
                                               int64
           4
               Fuel Type
                              301 non-null
                                               object
           5
               Seller_Type
                              301 non-null
                                               object
           6
               Transmission
                              301 non-null
                                               object
           7
               0wner
                              301 non-null
                                               int64
               Selling_Price 301 non-null
                                               float64
          dtypes: float64(2), int64(3), object(4)
          memory usage: 21.3+ KB
          Car Name
```

In [63]: from sklearn.preprocessing import LabelEncoder

```
In [64]: LE= LabelEncoder()
```

In [65]: LE

Out[65]: LabelEncoder()

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On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [66]: dataset['Car_Name']=LE.fit_transform(dataset['Car_Name'])
```

In [67]: dataset.head()

Out[67]:

	Car_Name	Year	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	Transmission	Owne
0	90	2014	5.59	27000	Petrol	Dealer	Manual	(
1	93	2013	9.54	43000	Diesel	Dealer	Manual	(
2	68	2017	9.85	6900	Petrol	Dealer	Manual	(
3	96	2011	4.15	5200	Petrol	Dealer	Manual	(
4	92	2014	6.87	42450	Diesel	Dealer	Manual	(
4								

Fuel_Type

```
In [68]: dataset['Fuel_Type'].unique()
```

Out[68]: array(['Petrol', 'Diesel', 'CNG'], dtype=object)

```
In [ ]:
```

```
In [69]: dataset['Fuel_Type']=LE.fit_transform(dataset['Fuel_Type'])
```

```
In [70]: dataset.head(1)
```

Out[70]: Car_Name Year Present_Price Kms_Driven Fuel_Type Seller_Type Transmission Owne

0 90 2014 5.59 27000 2 Dealer Manual

Seller_Type

```
In [71]: dataset['Seller_Type']=LE.fit_transform(dataset['Seller_Type'])
```

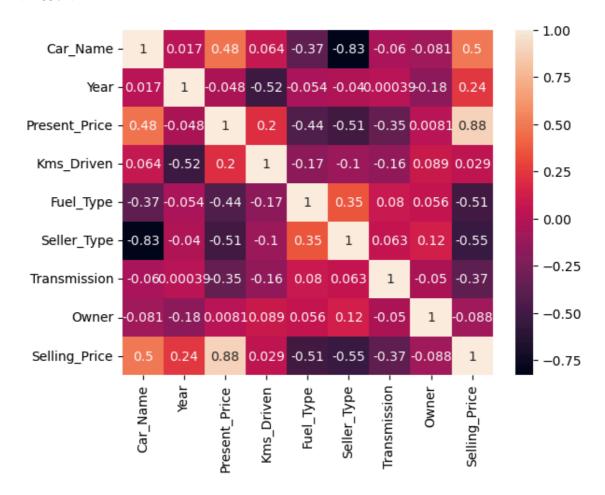
```
In [72]:
          dataset.head(1)
Out[72]:
              Car_Name Year Present_Price Kms_Driven Fuel_Type Seller_Type Transmission Owne
           0
                     90
                        2014
                                       5.59
                                                  27000
                                                                2
                                                                           0
                                                                                    Manual
          Transmission
          dataset['Transmission']=LE.fit_transform(dataset['Transmission'])
In [74]: dataset.head(1)
Out[74]:
              Car_Name Year Present_Price Kms_Driven Fuel_Type Seller_Type Transmission Owne
                     90 2014
                                       5.59
                                                  27000
          x=dataset.iloc[:,:-1]
In [75]:
In [76]:
Out[76]:
                           Year Present_Price Kms_Driven Fuel_Type Seller_Type Transmission Ow
                Car_Name
                           2014
                                                                  2
             0
                       90
                                         5.59
                                                    27000
                                                                              0
                                                                                           1
             1
                       93
                           2013
                                         9.54
                                                    43000
                                                                  1
                                                                              0
                                                                                           1
                           2017
                                                                  2
             2
                       68
                                         9.85
                                                    6900
                                                                              0
                                                                                           1
             3
                       96
                          2011
                                         4.15
                                                    5200
                                                                  2
                                                                              0
                                                                                           1
             4
                       92 2014
                                         6.87
                                                    42450
                                                                              0
                                                                                           1
                           2016
           296
                       69
                                        11.60
                                                    33988
                                                                  1
                                                                              0
                                                                                           1
           297
                       66
                           2015
                                         5.90
                                                    60000
                                                                  2
                                                                              0
                                                                                           1
           298
                           2009
                                        11.00
                                                    87934
                                                                  2
                       69
           299
                       69
                           2017
                                        12.50
                                                    9000
           300
                       66 2016
                                         5.90
                                                    5464
                                                                  2
          301 rows × 8 columns
In [77]:
          y=dataset['Selling_Price']
```

```
In [78]:
Out[78]: 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
In [79]: from sklearn.preprocessing import StandardScaler
In [80]:
          SS=StandardScaler()
In [81]:
          SS
Out[81]: StandardScaler()
          In a Jupyter environment, please rerun this cell to show the HTML representation or
          trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this page
          with nbviewer.org.
In [82]: x=pd.DataFrame(SS.fit_transform(x), columns=x.columns)
In [83]:
          x.head()
Out[83]:
              Car_Name
                             Year
                                  Present_Price Kms_Driven Fuel_Type
                                                                      Seller_Type Transmission
           0
               1.074323
                        0.128897
                                      -0.236215
                                                  -0.256224
                                                             0.500183
                                                                        -0.737285
                                                                                      0.39148 -
           1
               1.191828 -0.217514
                                      0.221505
                                                  0.155911
                                                            -1.852241
                                                                        -0.737285
                                                                                      0.39148 -
               0.212627
           2
                                      0.257427
                                                             0.500183
                         1.168129
                                                  -0.773969
                                                                        -0.737285
                                                                                      0.39148 -
           3
               1.309332 -0.910335
                                      -0.403079
                                                  -0.817758
                                                             0.500183
                                                                        -0.737285
                                                                                      0.39148 -
               1.152659
                         0.128897
                                      -0.087890
                                                  0.141743
                                                            -1.852241
                                                                        -0.737285
                                                                                      0.39148 -
In [84]: from sklearn.model selection import train test split
In [85]:
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_star
```

```
In [86]: from sklearn.linear_model import LinearRegression, Lasso, Ridge, ElasticNet
    from sklearn.tree import DecisionTreeRegressor
    from sklearn.neighbors import KNeighborsRegressor
    from sklearn.svm import SVR
    from sklearn.ensemble import RandomForestRegressor
```

In [87]: | sns.heatmap(data=dataset.corr(),annot=True)

Out[87]: <Axes: >



LinearRegression()

```
In [88]: LR=LinearRegression()
```

In [89]: LR

Out[89]: LinearRegression()

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```
In [90]: LR.fit(x_train,y_train)
LR.score(x_train,y_train)*100 , LR.score(x_test,y_test)*100
```

Out[90]: (88.40630578239454, 84.65539666857805)

Lasso

```
In [91]: LR1= Lasso(alpha=0.05)
LR1.fit(x_train,y_train)
LR1.score(x_train,y_train)*100 , LR1.score(x_test,y_test)*100
```

Out[91]: (88.35433202380113, 84.42023265451037)

Ridge

```
In [92]: LR2= Ridge(alpha=10)
    LR2.fit(x_train,y_train)
    LR2.score(x_train,y_train)*100 , LR2.score(x_test,y_test)*100
```

Out[92]: (88.28628537091495, 84.16213595432282)

ElasticNet

```
In [93]: LR3= ElasticNet(alpha=0.5)
LR3.fit(x_train,y_train)
LR3.score(x_train,y_train)*100 , LR3.score(x_test,y_test)*100
```

Out[93]: (84.00059239671332, 78.3177718663528)

Decision Tree

```
In [94]: dt=DecisionTreeRegressor(max_depth=13)
    dt.fit(x_train,y_train)
    dt.score(x_train,y_train)*100 , dt.score(x_test,y_test)*100
```

Out[94]: (99.99983614119861, 95.47372436648206)

In [95]: from sklearn.metrics import mean_squared_error, mean_absolute_error

```
In [96]: mean_squared_error(y_test,dt.predict(x_test)), mean_absolute_error(y_test,d
```

Out[96]: (1.042654049180328, 0.6585737704918032)

Random Forest Regressor

```
In [97]: rf=RandomForestRegressor(n_estimators=100)
    rf.fit(x_train,y_train)
    rf.score(x_train,y_train)*100 , rf.score(x_test,y_test)*100
```

Out[97]: (98.30481566012408, 96.41847578267432)

```
car prediction Regression supervised ML - Jupyter Notebook
          mean_squared_error(y_test,rf.predict(x_test)), mean_absolute_error(y_test,r
In [98]:
Out[98]: (0.8250250381967219, 0.5829622950819674)
          Support Vector regression
 In [99]:
          SV=SVR()
          SV.fit(x_train,y_train)
          SV.score(x train,y train)*100 , SV.score(x test,y test)*100
Out[99]: (66.00840380338376, 78.48466914602926)
          KNeighborsRegressor
In [100]:
          Knn=KNeighborsRegressor()
          Knn.fit(x_train,y_train)
          Knn.score(x_train,y_train)*100 , Knn.score(x_test,y_test)*100
  In [ ]:
In [101]: rf.predict([[-1.275759,0.821718,-0.817924,-0.333500,0.500183,1.356327,-2.554
          C:\Users\adity\anaconda3\New folder\Lib\site-packages\sklearn\base.py:464:
          UserWarning: X does not have valid feature names, but RandomForestRegresso
          r was fitted with feature names
```

```
Out[100]: (91.06681012800678, 93.29996797075346)
             warnings.warn(
Out[101]: array([0.4543])
In [102]: y test
Out[102]: 177
                   0.35
           289
                  10.11
           228
                   4.95
           198
                   0.15
                   6.95
           60
           234
                   5.50
           296
                   9.50
                   2.10
           281
           285
                   7.40
                   0.30
           182
           Name: Selling_Price, Length: 61, dtype: float64
In [107]:
          new_data=pd.DataFrame([['ritz',2014,5.59,27000,'Petrol','Dealer','Manual',0
           new data
Out[
```

[107]:		Car_Name	Year	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	Transmission	Owne
	0	0	2014	5.59	27000	Petrol	Dealer	Manual	(
	4					_			

```
In [ ]:
In [109]: new_data['Fuel_Type']=LE.fit_transform(new_data['Fuel_Type'])
In [110]: new_data['Seller_Type']=LE.fit_transform(new_data['Seller_Type'])
In [111]: new_data['Transmission']=LE.fit_transform(new_data['Transmission'])
In [112]:
           new_data
Out[112]:
              Car Name
                        Year Present_Price Kms_Driven Fuel_Type Seller_Type Transmission Owne
            0
                        2014
                                      5.59
                                                27000
          new_data=pd.DataFrame(SS.transform(new_data), columns=new_data.columns)
In [114]: | new_data
Out[114]:
              Car_Name
                            Year
                                 Present_Price Kms_Driven Fuel_Type Seller_Type Transmission
            0
                 -2.4508 0.128897
                                    -0.236215
                                                -0.256224
                                                          -4.204665
                                                                     -0.737285
                                                                                 -2.554408 -0
In [115]: rf.predict(new_data)
Out[115]: array([3.4745])
  In [ ]:
  In [ ]:
```