```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
```

```
In [2]: df=pd.read_csv(r"c7_used_cars.csv")
df
```

Out[2]:

		Unnamed: 0	model	year	price	transmission	mileage	fuelType	tax	mpg	engineSize
٠	0	0	T-Roc	2019	25000	Automatic	13904	Diesel	145	49.6	2.0
	1	1	T-Roc	2019	26883	Automatic	4562	Diesel	145	49.6	2.0
	2	2	T-Roc	2019	20000	Manual	7414	Diesel	145	50.4	2.0
	3	3	T-Roc	2019	33492	Automatic	4825	Petrol	145	32.5	2.0
	4	4	T-Roc	2019	22900	Semi-Auto	6500	Petrol	150	39.8	1.5
								•••			
	99182	10663	А3	2020	16999	Manual	4018	Petrol	145	49.6	1.0
	99183	10664	А3	2020	16999	Manual	1978	Petrol	150	49.6	1.0
	99184	10665	А3	2020	17199	Manual	609	Petrol	150	49.6	1.0
	99185	10666	Q3	2017	19499	Automatic	8646	Petrol	150	47.9	1.4
	99186	10667	Q3	2016	15999	Manual	11855	Petrol	150	47.9	1.4

99187 rows × 11 columns

memory usage: 8.3+ MB

In [3]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 99187 entries, 0 to 99186
Data columns (total 11 columns):

Column Non-Null Count Dtype # --------------0 Unnamed: 0 99187 non-null int64 1 model 99187 non-null object 2 year 99187 non-null int64 3 price 99187 non-null int64 4 transmission 99187 non-null object 5 99187 non-null int64 mileage 6 99187 non-null object fuelType 7 99187 non-null int64 tax 8 mpg 99187 non-null float64 9 99187 non-null float64 engineSize 99187 non-null object 10 Make dtypes: float64(2), int64(5), object(4)

```
In [4]: |df=df.dropna()
 In [5]: |df.describe()
 Out[5]:
                  Unnamed: 0
                                                  price
                                     year
                                                             mileage
                                                                             tax
                                                                                        mpg
           count 99187.000000
                             99187.000000
                                           99187.000000
                                                        99187.000000 99187.000000 99187.000000 9
                  6294.413532
                              2017.087723
                                           16805.347656
                                                        23058.914213
                                                                       120.299838
                                                                                    55.166825
           mean
                  4265.588536
                                                        21148.523721
                                                                       63.150926
                                                                                    16.138522
                                 2.123934
                                            9866.773417
             std
                              1970.000000
            min
                     0.000000
                                             450.000000
                                                            1.000000
                                                                        0.000000
                                                                                     0.300000
            25%
                  2755.000000
                              2016.000000
                                            9999.000000
                                                         7425.000000
                                                                       125.000000
                                                                                    47.100000
            50%
                  5591.000000
                              2017.000000
                                           14495.000000
                                                        17460.000000
                                                                       145.000000
                                                                                    54.300000
            75%
                  9420.000000
                              2019.000000
                                           20870.000000
                                                        32339.000000
                                                                       145.000000
                                                                                    62.800000
            max 17964.000000
                              2060.000000 159999.000000 323000.000000
                                                                      580.000000
                                                                                   470.800000
 In [6]: | df.columns
 Out[6]: Index(['Unnamed: 0', 'model', 'year', 'price', 'transmission', 'mileage',
                  'fuelType', 'tax', 'mpg', 'engineSize', 'Make'],
                dtype='object')
In [21]: x=df[['Unnamed: 0', 'year', 'price', 'mileage',
                'tax', 'mpg', 'engineSize']]
          y=df['fuelType']
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
          lr=LogisticRegression()
          lr.fit(x train,y train)
          C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear model\ logistic.py:
          763: ConvergenceWarning: lbfgs failed to converge (status=1):
          STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
          Increase the number of iterations (max_iter) or scale the data as shown in:
              https://scikit-learn.org/stable/modules/preprocessing.html (https://sciki
          t-learn.org/stable/modules/preprocessing.html)
          Please also refer to the documentation for alternative solver options:
              https://scikit-learn.org/stable/modules/linear model.html#logistic-regres
          sion (https://scikit-learn.org/stable/modules/linear model.html#logistic-regr
          ession)
            n_iter_i = _check_optimize_result(
Out[21]: LogisticRegression()
In [13]: lr.predict(x_test)
Out[13]: array(['Petrol', 'Diesel', 'Diesel', ..., 'Petrol', 'Petrol', 'Diesel'],
                dtype=object)
```

```
In [14]: |lr.score(x_test,y_test)
Out[14]: 0.6925765366132338
In [15]: from sklearn.preprocessing import StandardScaler
         fs=StandardScaler().fit_transform(x)
         logr=LogisticRegression()
         logr.fit(fs,y)
         C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:
         763: ConvergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html (https://sciki
         t-learn.org/stable/modules/preprocessing.html)
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear_model.html#logistic-regres
         sion (https://scikit-learn.org/stable/modules/linear model.html#logistic-regr
           n_iter_i = _check_optimize_result(
Out[15]: LogisticRegression()
In [17]: o=[[6,7,8,34,52,6,7]]
         prediction=logr.predict(o)
         print(prediction)
         ['Diesel']
In [18]: logr.classes
Out[18]: array(['Diesel', 'Electric', 'Hybrid', 'Other', 'Petrol'], dtype=object)
In [19]: logr.predict_proba(o)[0][0]
Out[19]: 0.9795692751051442
In [20]: logr.predict_proba(o)[0][1]
Out[20]: 1.2174728850843716e-22
 In [ ]:
```