# Classification of Fuel type of the car (Petrol or Diesel) using 4 layer Backpropagation

```
In [29]:
import pandas as pd
import numpy as np
import math
from sklearn.preprocessing import OneHotEncoder
from sklearn.metrics import accuracy_score
import time
from sklearn.metrics import mean_squared_error
import statistics
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
In [ ]:
In [30]:
data = pd.read_csv('ToyotaCorolla.csv')#-----Loading the dataset
In [31]:
labelencoder = LabelEncoder()#-----using Labelencode
data['FuelType'] = labelencoder.fit_transform(data['FuelType'])
In [32]:
y = data.FuelType#-----label set
X = data#-----without label data
In [33]:
X= X.drop(columns="FuelType")
In [34]:
X_train, X_test, y_train, y_test = train_test_split(X, y,test_size=0.7)
print(y_train.shape)
print(y_test.shape)
(430,)
```

(1006,)

```
In [35]:
```

```
X_train[1:10]
```

#### Out[35]:

|      | Price | Age | KM     | CC   | Weight |
|------|-------|-----|--------|------|--------|
| 1222 | 8250  | 79  | 84966  | 1600 | 1070   |
| 378  | 6500  | 53  | 216000 | 1900 | 1110   |
| 1182 | 9900  | 80  | 92255  | 1600 | 1105   |
| 227  | 11690 | 34  | 65345  | 1400 | 1060   |
| 199  | 11950 | 39  | 98823  | 1600 | 1119   |
| 799  | 8250  | 65  | 74179  | 1600 | 1050   |
| 1244 | 6950  | 70  | 81663  | 1600 | 1050   |
| 282  | 12850 | 39  | 45713  | 1400 | 1085   |
| 502  | 9900  | 53  | 57475  | 1600 | 1040   |

#### In [36]:

```
y_train[1:10]
#print(y_train.nunique)
```

#### Out[36]:

```
1222
        1
378
        2
1182
        1
227
        1
199
        0
799
        1
1244
        1
282
        1
502
        1
Name: FuelType, dtype: int32
```

#### In [37]:

```
X_train=pd.DataFrame(X_train).to_numpy()#-----converting pandas dataframe to numpy
X_test=pd.DataFrame(X_test).to_numpy()
y_train=pd.DataFrame(y_train).to_numpy()
y_test=pd.DataFrame(y_test).to_numpy()
```

#### In [38]:

```
def sigmoid(s, deriv=False):
   if(deriv == True):
     return s * (1 - s)
   return 1/(1 + np.exp(-s))
```

```
In [ ]:
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In [39]:
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```
Input_layer = 5
Output_layer = 1
Hidden_layer_1 = 100
Hidden_layer_2 = 100
Hidden_layer_3 = 100
Hidden_layer_4 = 100
learning_rate = 0.01
Weight_1 = np.random.normal(size=[Input_layer, Hidden_layer_1])
Weight_2 = np.random.normal(size=[Hidden_layer_1, Hidden_layer_2])
Weight_3 = np.random.normal(size=[Hidden_layer_2, Hidden_layer_3])
Weight_4 = np.random.normal(size=[Hidden_layer_3, Hidden_layer_4])
Weight_5 = np.random.normal(size=[Hidden_layer_4, Output_layer])
```

```
In [ ]:
```

### Main Neural network class starts here

#### In [40]:

```
import numpy as np
from random import seed
learning rate = 0.0001
def __init__(self):
       #parameters
       self.Input_layer = 5
       self.Output_layer = 1
       self.Hidden_layer_1 = 10#------
       self.Hidden_layer_2 = 20
       self.Hidden_layer_3 = 30
       self.Hidden_layer_4 = 15
       learning_rate = 0.0001#---
       #weights
       self.Weight_1 = np.random.normal(size=[self.Input_layer, self.Hidden_layer_1])
       self.Weight_2 = np.random.normal(size=[self.Hidden_layer_1, self.Hidden_layer_2]) #
       self.Weight_3 = np.random.normal(size=[self.Hidden_layer_2, self.Hidden_layer_3])
       self.Weight_4 = np.random.normal(size=[self.Hidden_layer_3, self.Hidden_layer_4])
       self.Weight_5 = np.random.normal(size=[self.Hidden_layer_4, self.Output_layer])
   def feedForward(self, X_train):#-----Creating function
       #forward propogation through the network
       self.Forward = np.dot(X_train, self.Weight_1)
       self.Forward2 = self.sigmoid(self.Forward)
       self.Forward3 = np.dot(self.Forward2, self.Weight_2) #-----
       self.Forward4 = self.sigmoid(self.Forward3) #------Activation function
       self.Forward5 = np.dot(self.Forward4, self.Weight_3)
       self.Forward6 = self.sigmoid(self.Forward5)
       self.Forward7 = np.dot(self.Forward6, self.Weight 4)
       self.Forward8 = self.sigmoid(self.Forward7)
       self.Forward9 = np.dot(self.Forward8, self.Weight_5)
       output = self.sigmoid(self.Forward9)
       return output
   def sigmoid(self, s, deriv=False):#-----Defining Sigma
       if (deriv == True):
          return s * (1 - s)
       return 1/(1 + np.exp(-s))
   def Backpropagate(self, X train, y train, output):#-----
       #Backpropagate propogate through the network
       self.output_error = y_train - output
       self.output_delta = self.output_error * self.sigmoid(output, deriv=True)#------
       self.Forward8_error = self.output_delta.dot(self.Weight_5.T)
       self.Forward8_delta = self.Forward8_error * self.sigmoid(self.Forward8, deriv=True)
       self.Forward6 error = self.Forward8 delta.dot(self.Weight 4.T)
       self.Forward6_delta = self.Forward6_error * self.sigmoid(self.Forward6, deriv=True)
       self.Forward4_error = self.Forward6_delta.dot(self.Weight_3.T)
       self.Forward4 delta = self.Forward4 error * self.sigmoid(self.Forward4, deriv=True)
```

## **Final Training Accuracy**

```
In [53]:
NN = Four layer NeuralNetwork()
for i in range(30): #----- number of epochs for neural network and
    if i%1==0:
     print("Training Accuracy: epoch", + i+1, str(accuracy_score(y_train.round(),NN.feedFo
   NN.train(X_train, y_train)
print("Accuracy: " + str(accuracy score(y train.round(),NN.feedForward(X train).round())*10
Training Accuracy: epoch 1 0.9302325581395349
Training Accuracy: epoch 2 0.9302325581395349
Training Accuracy: epoch 3 0.9302325581395349
Training Accuracy: epoch 4 0.9302325581395349
Training Accuracy: epoch 5 1.8604651162790697
Training Accuracy: epoch 6 1.8604651162790697
Training Accuracy: epoch 7 1.8604651162790697
Training Accuracy: epoch 8 77.90697674418605
Training Accuracy: epoch 9 81.16279069767441
Training Accuracy: epoch 10 86.27906976744187
Training Accuracy: epoch 11 86.27906976744187
Training Accuracy: epoch 12 86.27906976744187
Training Accuracy: epoch 13 87.44186046511628
Training Accuracy: epoch 14 87.44186046511628
Training Accuracy: epoch 15 87.44186046511628
Training Accuracy: epoch 16 87.44186046511628
Training Accuracy: epoch 17 87.44186046511628
Training Accuracy: epoch 18 87.44186046511628
Training Accuracy: epoch 19 87.44186046511628
Training Accuracy: epoch 20 87.44186046511628
Training Accuracy: epoch 21 87.44186046511628
Training Accuracy: epoch 22 87.44186046511628
Training Accuracy: epoch 23 87.44186046511628
Training Accuracy: epoch 24 87.44186046511628
Training Accuracy: epoch 25 87.44186046511628
Training Accuracy: epoch 26 87.44186046511628
Training Accuracy: epoch 27 87.44186046511628
Training Accuracy: epoch 28 87.44186046511628
Training Accuracy: epoch 29 87.44186046511628
Training Accuracy: epoch 30 87.44186046511628
Accuracy: 87.44186046511628
<ipython-input-40-534bfa9dd00c>:44: RuntimeWarning: overflow encountered in
  return 1/(1 + np.exp(-s))
In [ ]:
```

## **Final Testing Accuracy**

```
In [57]:
#---testing accuracy
NN = Four_layer_NeuralNetwork()
print("Testing Accuracy: " + str(accuracy_score(y_test.round(),NN.feedForward(X_test).round
Testing Accuracy: 79.72166998011929
<ipython-input-40-534bfa9dd00c>:44: RuntimeWarning: overflow encountered in
  return 1/(1 + np.exp(-s))
Type Markdown and LaTeX: \alpha^2
In [ ]:
```

| In [ ]: |  |  |
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|         |  |  |