Dataset link

https://archive.ics.uci.edu/ml/datasets/lris#targetText=UCl%20Machine%20Learning%20Repository%3A% (https://archive.ics.uci.edu/ml/datasets/lris#targetText=UCl%20Machine%20Learning%20Repository%3A% (https://archive.ics.uci.edu/ml/datasets/lris#targetText=UCl%20Machine%20Learning%20Repository%2Datasets/lris#targetText=Ucl%20Machine%2Datasets/lris#targetText=Ucl%2Datasets/lris#targetText

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In [20]:
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```
%matplotlib inline
```

In [21]:

```
from sklearn import datasets
import numpy as np
import csv
import pandas as pd
from matplotlib.colors import ListedColormap
import matplotlib.pyplot as plt
from sklearn.metrics import accuracy_score
```

In [22]:

```
iris = datasets.load_iris()
x = iris.data[:, [0,2]]
y = iris.target
x_moded=np.delete(x,[100,101,102,103,104,105,106,107,108,109,110,111,112,113,114,115,116,11
y_moded=np.delete(y,[100,101,102,103,104,105,106,107,108,109,110,111,112,113,114,115,116,11
# splitting data into test and train data set
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x_moded,y_moded, test_size=0.3, random_
#print(x_train)
#print(y_train)

weights = np.zeros((1, 2))
gradient_matrix = np.zeros((2, 1))
hessian_matrix = np.zeros((2, 2))
result_matrix = np.zeros((2, 1))
```

In [23]:

```
#creating activation function sigmoid function
def sigmoid(x):
    return 1/(1+np.exp(-x))
```

In [24]:

```
training set
= [0 for i in range(len(x_test))]
              #-----epochs ar
br k in range(50):
  for i in range (0,len(x_train)-60): #-----training
     actual_result = weights[0][0]*x_train[i][0] + weights[0][1]*x_train[i][1]
     if actual_result > 0:
        fx = 1
          # print("greater than 0")
     if actual result <= 0:</pre>
        fx = 0
           #print("less than 0")
     #----
     d=y_train[i]
     w1=weights[0][0]
     w2=weights[0][1]
     x1=x_train[i][0]
     x2=x_train[i][1]
     gradient_matrix[0][0] = -x1*(d-w1*x1-w2*x2) #------putting value
     gradient matrix[1][0] = -x2*(d-w1*x1-w2*x2)
     hessian_matrix[0][0] = x1**2
     hessian_matrix[1][0] = x2*x1
     hessian_matrix[1][1] = x2**2
     A_inv = np.linalg.pinv(hessian_matrix)#pseudo matrix for matrix cannot be find
     #print(A inv)
     result_matrix[0][0] = (A_inv[0][0]*gradient_matrix[0][0])+ (A_inv[0][1]*gradient_matr
     #-----
     if fx != y_train[i]:
        error = y train[i] - fx
        weights[0][1] = weights[0][1] - result_matrix[1][0]#------
        \#weights[0][0] = weights[0][0] + error*x_train[i][0]
        \#weights[0][1] = weights[0][1] + error*x_train[i][1]
           #print(weights[:])
        p=p+1
        print(p)
        print("learning")
        #plt.plot(p)
     else:
        weights[0][0] = weights[0][0]
        weights[0][1] = weights[0][1]
        print("no error")
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Accuracy 0.7

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