```
In [28]:
          import numpy as np
          import pandas as pd
          df = pd.read csv("https://raw.githubusercontent.com/simransinghgulati/chemical-segregation/master/Logistic X Train.cs
Out[28]:
                     f1
                              f2
                                       f3
             0 -1.239375 0.749101 -0.528515
             1 -1.036070 0.801436 -1.283712
             2 -0.615579 1.579521 -1.391927
            3 1.335978 1.348651 1.433564
             4 0.658925 1.300019 0.571603
          2995 -0.455628 1.302303 -1.338027
          2996 -0.434551 1.597813 -1.748643
               0.088277 1.638789 -2.193641
               1.525155 0.859234 1.505308
          2999 -0.979817 0.563954 -1.539394
         3000 rows × 3 columns
In [41]:
          ones = np.ones((df.shape[0],1))
          x = df
          x = np.hstack((ones,x))
          x = np.asarray(x)
          Χ
Out[41]: array([[ 1.
                              , -1.23937466, 0.74910149, -0.52851491],
                 [ 1.
                              , -1.03607028, 0.80143631, -1.28371152],
                 [ 1.
                              , -0.61557914, 1.57952053, -1.39192706],
                 . . . ,
                 [ 1.
                              , 0.08827657, 1.63878904, -2.19364056],
```

```
[ 1.
                            , 1.52515501, 0.85923429, 1.50530791],
                ſ 1.
                             , -0.97981744, 0.56395376, -1.53939416]])
         x.shape
In [42]:
Out[42]: (3000, 4)
In [46]: y = pd.read csv("https://raw.githubusercontent.com/simransinghqulati/chemical-segregation/master/Logistic Y Train.csv
          y = np.array(y)
          y = y.reshape((3000,))
In [52]:
          def hypothesis(x,theta):
              z = np.dot(x, theta)
              y = sigmoid(z)
              return y
          def sigmoid(z):
              return 1/(1+np.exp(-z))
          def error(x,theta,y):
              m = len(x)
              y = hypothesis(x, theta)
              err = np.dot(-y, np.log(y)) - np.dot((1-y), np.log(y))
              return err/m
          def gradient(x,theta,y):
              y = hypothesis(x,theta)
              m = len(x)
              grad = np.dot(x.T,y -y)
              return grad/m
          def gradient des(x,y,learning rate=0.1,epochs=50):
              m,n = x.shape
              theta = np.zeros((n,))
              err list = []
              for i in range(epochs):
                  err = error(x, theta, y)
                  err list.append(err)
                  grad = gradient(x, theta, y)
                  theta = theta - learning rate * grad
              return theta, err list
          theta, err_list = gradient_des(x,y)
In [55]:
```

```
import matplotlib.pyplot as plt
In [56]:
          plt.plot(err list)
Out[56]: [<matplotlib.lines.Line2D at 0x1bb4a8afd90>]
         1.1
         1.0
          0.9
          0.8
          0.7
                                      30
                     10
                              20
                                              40
                                                      50
In [58]:
          theta
Out[58]: array([-0.10242894, -0.9628054, 0.23282612, -1.00459897])
         y_{=} hypothesis(x,theta)
In [59]:
Out[59]: array([0.85768617, 0.91461367, 0.90519568, ..., 0.91666396, 0.0529986,
                0.92544389])
In [60]:
          y_{y_{0.5}=1}
          y_[y_<0.5] = 0
          У_
Out[60]: array([1., 1., 1., ..., 1., 0., 1.])
          np.mean(y_==y)*100
In [61]:
         96.73333333333333
```

```
Out[61]:
In [70]: y_pred = hypothesis(x[8],theta)
y_pred
Out[70]: 0.8916807809091882
In []:
```