```
In [1]: 

import pandas as pd
import numpy as np
import random
import matplotlib.pyplot as plt
import time
```

#### Task1

# **Reading datasets**

```
A=pd.read_csv("pp3data/pp3data/A.csv",header=None)

In [2]:
            A=A.to_numpy()
            labels_A=pd.read_csv("pp3data/pp3data/labels-A.csv",header=None)
            labels_A=labels_A.to_numpy().flatten()
         ▶ B=pd.read_csv("pp3data/pp3data/B.csv", header=None)
In [3]:
            B=B.to_numpy()
            labels_B=pd.read_csv("pp3data/pp3data/labels-B.csv",header=None)
            labels B=labels B.to numpy().flatten()
In [4]:

    USPS=pd.read_csv("pp3data/pp3data/usps.csv",header=None)

            USPS=USPS.to numpy()
            labels USPS=pd.read csv("pp3data/pp3data/labels-usps.csv",header=None)
            labels_USPS=labels_USPS.to_numpy().flatten()
In [5]:

    data=[A,B,USPS]

            labels=[labels A, labels B, labels USPS]
In [6]:
         ▶ generative=[]
            discriminative=[]
            average=[]
            partition=[0.1,0.2,0.3,0.4,0.5,0.6,0.7,0.8,0.9,1.0]
```

# **Calculating generative results**

```
In [7]:

    for i in range(3):

                X=data[i]
                Y=labels[i]
                N = X.shape[0]
                N_{\text{test}} = int(N/3)
                N_train = N-N_test
                index = []
                for j in range(N):
                    index.append(j)
                generative_result = {}
                for t in range(1, 31):
                    test_index = random.sample(index, N_test)
                    train_index = [x for x in index if x not in test_index]
                    xtrain = X[train index,:]
                    ytrain = Y[train_index]
                    xtest = X[test_index,:]
                    ytest = Y[test_index]
                    for p in partition:
                         p_N = int(p*N_train)
                        p_x = xtrain[:p_N,]
                         p_y = ytrain[:p_N]
                         class1 = p_x[(np.where(p_y==1)[0]),:]
                        mu1 = np.mean(class1, axis=0)
                        class2 = p_x[(np.where(p_y==0)[0]),:]
                        mu2 = np.mean(class2, axis=0)
                         d = X.shape[1]
                         pred = np.zeros(xtest.shape[0])
                        total1 = np.zeros((d,d))
                         for i in range(class1.shape[0]):
                             total1+= np.outer((class1[i,:]-mu1), (class1[i,:]-mu1))
                        S1 = total1/class1.shape[0]
                        total2 = np.zeros((d,d))
                         for i in range(class2.shape[0]):
                             total2+= np.outer((class2[i,:]-mu2), (class2[i,:]-mu2))
                        S2 = total2/class2.shape[0]
                         S = ((class1.shape[0]/p_N)*S1) + ((class2.shape[0]/p_N)*S2)
                        PC1 = class1.shape[0]/p_N
                        PC2 = class2.shape[0]/p N
                        W = np.dot(np.linalg.inv(S),(mu1-mu2))
                        W0=np.log(PC1/PC2)+0.5*np.dot(np.dot(mu2, np.linalg.inv(S)),mu2)-
                         for i in range(pred.size):
                             x test = xtest[i,:]
                             pred[i] = np.dot(W,x_test) + W0
                         prediction = np.where(pred>=0,1,0)
                         error = (np.sum(np.logical_xor(ytest, prediction)))/ytest.size
                         generative_result[(str(t),str(p))] = error
                generative.append(generative_result)
```

### Calculating discriminative results

```
In [8]:

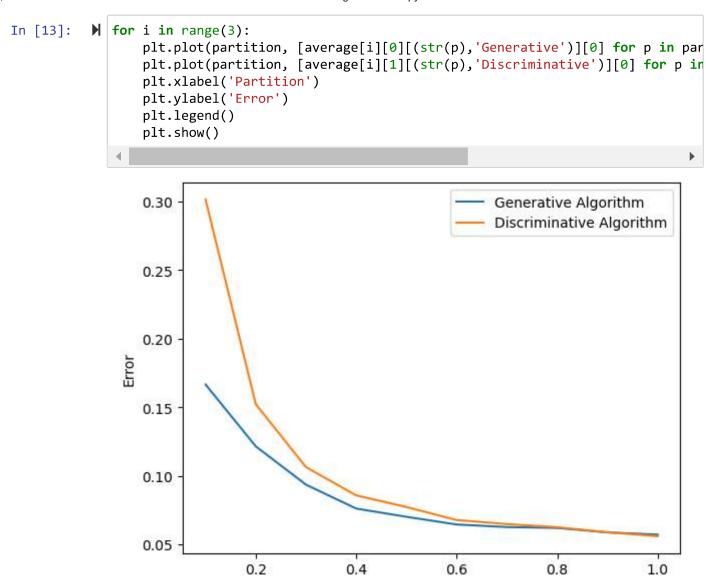
    for i in range(3):

                X=data[i]
                Y=labels[i]
                N = X.shape[0]
                N \text{ test} = int(N/3)
                N_train = N-N_test
                index=[]
                for j in range(N):
                    index.append(j)
                discriminative_result = {}
                logisticR X = np.c [np.ones(N), X]
                for times in range(1, 31):
                    test_index = random.sample(index, N_test)
                    train index = [x for x in index if x not in test index]
                    xtrain = logisticR_X[train_index,:]
                    ytrain = Y[train_index]
                    xtest = logisticR_X[test_index,:]
                    ytest = Y[test_index]
                    for pv in partition:
                         pv N = int(pv*N train)
                        p_x = xtrain[:pv_N,]
                         p_y = ytrain[:pv_N]
                        W old = np.zeros(p x.shape[1])
                        W_new = np.matrix(np.zeros(p_x.shape[1]))
                         d = X.shape[1]
                         for iteration in range(100):
                             if not(np.linalg.norm(W new-W old)**2)/(np.linalg.norm(W old)
                             a = np.dot(p x, W old)
                             y = 1/(1 + np.exp(-a))
                             R = np.diag(y*(1-y))
                             W new = W old - np.dot(np.linalg.inv(0.1*np.identity(d+1) + n
                             W old=W new
                         a = np.dot(p x, W new)
                        y = 1/(1 + np.exp(-a))
                        total = np.zeros((d+1,d+1))
                         for i in range(y.size):
                             total+= (y[i]*(1-y[i]))*np.dot(p x[i,:], p x[i,:])
                         SN = 0.1*np.identity(d+1) + total
                         prob = np.zeros(xtest.shape[0])
                         for i in range(prob.size):
                             phi = xtest[i,:]
                             mu = np.dot(W_new.T, phi)
                             sigma_square = np.dot(np.dot(phi.T, SN), phi)
                             k = 1/np.sqrt(1 + (np.pi*sigma square/8))
                             prob[i] = 1/(1 + np.exp(-(k*mu)))
                         pred_labels = np.where(prob>=0.5,1,0)
                         misclassified_instances = np.sum(np.logical_xor(ytest, pred_label
                         test error rate = misclassified instances/ytest.size
                        discriminative result[(str(times), str(pv))] = test error rate
                discriminative.append(discriminative_result)
```

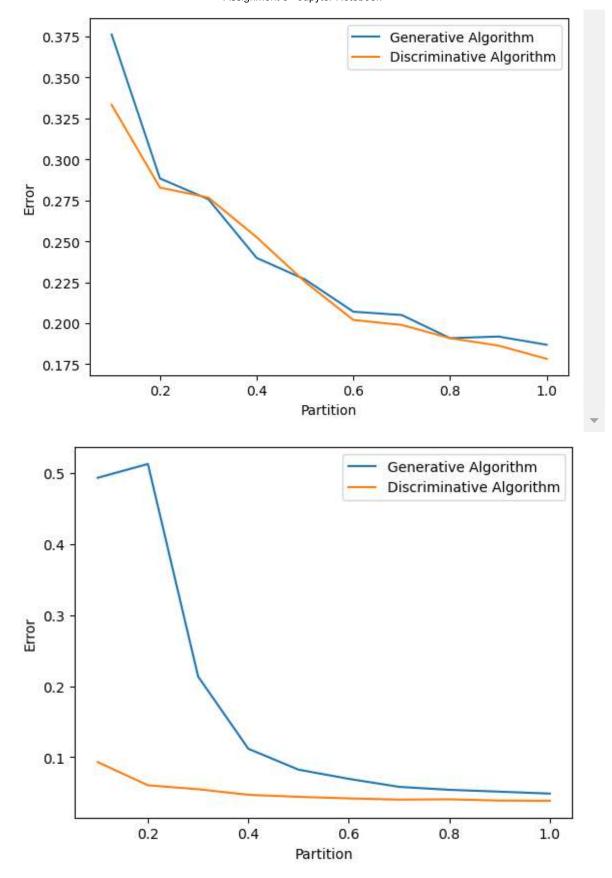
```
C:\Users\Aaryan Agarwal\AppData\Local\Temp\ipykernel_23140\1645444686.py:2
7: RuntimeWarning: invalid value encountered in true_divide
  if not(np.linalg.norm(W_new-W_old)**2)/(np.linalg.norm(W_old)**2>0.001):
```

## Calculating average

# **Graphs**



Partition



# Task2

```
In [8]:

    for i in range(1):

                new_d=np.c_[np.ones(newdata[i].shape[0]),newdata[i]]
                N = \text{new d.shape}[0]
                train_size = int((2/3) * len(newdata[i]))
                xtrain = new_d[:train_size]
                ytrain = newlabel[0][:train size]
                xtest = new_d[train_size:]
                ytest = newlabel[0][train_size:]
                W old=np.zeros(xtrain.shape[1])
                weight_vector = {}
                errors = \{\}
                for iteration in range(6000):
                    d = new_d.shape[1]
                    a = np.dot(xtrain, W_old)
                    y = 1/(1 + 1/np.exp(a))
                    R = np.diag(y*(1-y))
                    W_new = W_old - np.dot(np.linalg.inv(0.1*np.identity(d) + np.dot(np.d)
                    ts = time.time()
                    weight_vector[ts] = W_old
                    if ((np.linalg.norm(W new-W old)**2)/(np.linalg.norm(W old)**2)<0.001
                         break
                    else:
                        W old = W new
                         a = np.dot(xtrain, W new)
                        y = 1/(1 + np.exp(-a))
                        total = np.zeros((d,d))
                         for i in range(y.size):
                             total+= (y[i]*(1-y[i]))*np.dot(xtrain[i,:], xtrain[i,:])
                        SN = 0.1*np.identity(d) + total
                        pred = np.zeros(xtest.shape[0])
                         for i in range(pred.size):
                             phi = xtest[i,:]
                             mu = np.dot(W_new.T, phi)
                             sigma sq = np.dot(np.dot(phi.T, SN), phi)
                             a = 1/np.sqrt(1 + (np.pi*sigma sq/8))
                             pred[i] = 1/(1 + np.exp(-(a*mu)))
                         prediction = np.where(pred>=0.5,1,0)
                         error = np.sum(np.logical_xor(ytest, prediction))
                         test error = error/ytest.size
                         errors[ts] = test error
                newton.append((errors, weight vector))
```

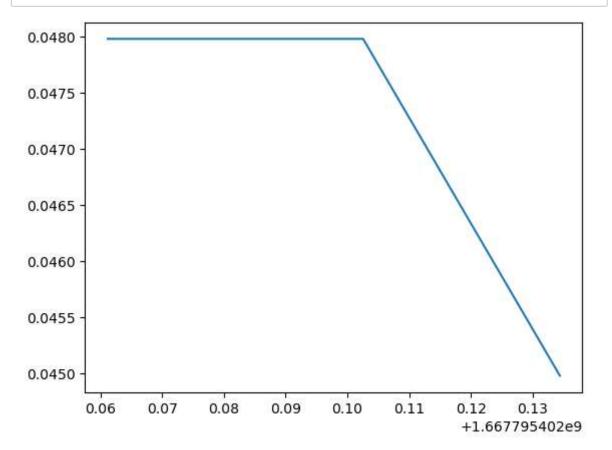
```
C:\Users\Aaryan Agarwal\AppData\Local\Temp\ipykernel_29020\1619481033.py:2
0: RuntimeWarning: divide by zero encountered in double_scalars
  if ((np.linalg.norm(W_new-W_old)**2)/(np.linalg.norm(W_old)**2)<0.001):</pre>
```

```
In [11]:

    for i in range(1):

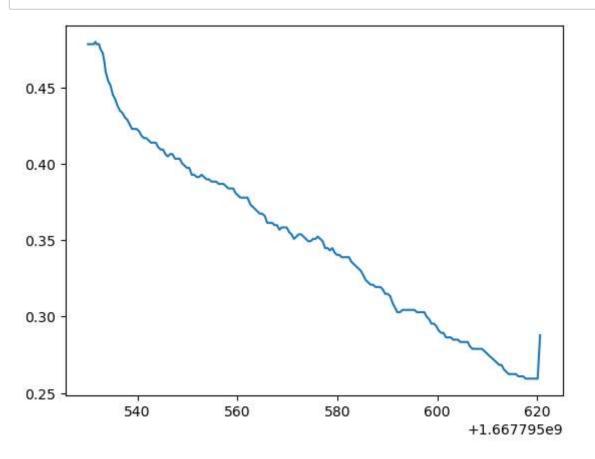
                 new_d=np.c_[np.ones(newdata[i].shape[0]),newdata[i]]
                 N = new_d.shape[0]
                 d = new d.shape[1]
                 train_size = int((2/3) * len(newdata[i]))
                 xtrain = new_d[:train_size]
                 ytrain = newlabel[0][:train size]
                 xtest = new_d[train_size:]
                 ytest = newlabel[0][train_size:]
                 W_old = np.zeros(xtrain.shape[1])
                 weight_vector = {}
                 errors = {}
                 for iteration in range(6000):
                      a = np.dot(xtrain, W old)
                     y = 1/(1 + 1/np.exp(a))
                     term_1 = np.dot(xtrain.T, (y-ytrain)) + 0.1*W_old
                     W_{new} = W_{old} - 0.001*(term_1)
                      if(iteration%10 == 0):
                          ts = time.time()
                          weight vector[ts] = W old
                      if ((np.linalg.norm(W_new-W_old)**2)/(np.linalg.norm(W_old)**2)<0.001</pre>
                          break
                     else:
                          W_old = W_new
                          # SN calculation
                          a = np.dot(xtrain, W new)
                          y = 1/(1 + np.exp(-a))
                          summation = np.zeros((d,d))
                          for i in range(y.size):
                              summation+= (y[i]*(1-y[i]))*np.dot(xtrain[i,:], xtrain[i,:])
                          SN = 0.1*np.identity(d) + summation
                          #predictions
                          pred = np.zeros(xtest.shape[0])
                          for i in range(pred.size):
                              phi = xtest[i,:]
                              mu = np.dot(W new.T, phi)
                              sigma sq = np.dot(np.dot(phi.T, SN), phi)
                              a = 1/np.sqrt(1 + (np.pi*sigma sq/8))
                              pred[i] = 1/(1 + np.exp(-(a*mu)))
                          prediction = np.where(pred>=0.5,1,0)
                          error = np.sum(np.logical_xor(ytest, prediction))
                          test error = error/ytest.size
                          errors[ts] = test error
                 gradient.append((errors, weight_vector))
```

```
C:\Users\Aaryan Agarwal\AppData\Local\Temp\ipykernel_29020\107682710.py:21:
RuntimeWarning: divide by zero encountered in double_scalars
  if ((np.linalg.norm(W_new-W_old)**2)/(np.linalg.norm(W_old)**2)<0.001):</pre>
```



IndexError: list index out of range

In [12]: plt.plot(gradient[1][0].keys(),gradient[1][0].values())
 plt.show()



In []: **M**