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```
#!/usr/bin/python
import numpy as np
import matplotlib.pyplot as plt
import numpy as np
import scipy.stats as sp
from mpl toolkits.mplot3d import Axes3D
import copy
import pprint
from matplotlib.backends.backend pdf import PdfPages
class analyZeData(object):
   def __init__(self, dataSetFile):
    # Load in the dataSet into an NP array
        # It retains the row column convention given the file
        # i j Zij Xij(1) Xij(2)
        self.NGROUPS=4
        self.dataSet=np.loadtxt(dataSetFile, comments="#")
        self.w=np.ones((self.dataSet.shape[0],self.NGROUPS))
        self.ones=np.ones((self.dataSet.shape[0],))
        self.zeros=np.zeros((self.dataSet.shape[0],))
        self.preBias=np.ones((self.dataSet.shape[0],2))
        self.LL=[]
    def bcMultiply(self, ynx2, xnx1):
       [m, n] = ynx2.shape
result = copy.deepcopy(ynx2)
        for i in range(m):
            result[i]=(ynx2[i]*xnx1[i])
        return result
    def weightedMean(self,w,x):
        [mu]=np.sum(self.bcMultiply(x,w), axis=0, keepdims=True) / np.sum(w)
        return mu
    def weightedCov(self,w,x,mu):
        [m, n] = self.w.shape
        sigma= np.mat(self.bcMultiply((x-mu),w)).T*(x-mu) / np.sum(w)
        return sigma
    def EStep(self, mu0, mu1, sigma0, sigma1, phi, lembda):
        self.w=np.ones((self.dataSet.shape[0],self.NGROUPS))
        y0=sp.multivariate normal(mu0, sigma0)
        y1=sp.multivariate_normal(mu1, sigma1)
        [m, n] = self.w.shape
       self.w=np.array([(1-phi)*lembda*y0.pdf(self.X), (1-phi)*(1-lembda)*y1.pdf(se
lf.X), (phi)*(1-lembda)*y0.pdf(self.X), (phi)*lembda*y1.pdf(self.X)]).T
        s=np.sum(self.w, axis=1)
        self.LL.append(np.sum(np.log(np.sum(self.w, axis=1))))
        one=np.ones(s.shape)
        s=np.divide(one,s)
        self.w=self.bcMultiply(self.w, s)
        #print "EStep", self.w[0]
    def precintBIAS(self, mu0, mu1, sigma0, sigma1, phi, lembda):
        y0=sp.multivariate_normal(mu0, sigma0)
        y1=sp.multivariate_normal(mu1, sigma1)
        p=np.array([(1-phi)*lembda*y0.pdf(self.X), (1-phi)*(1-lembda)*y1.pdf(self.X)
p1=np.sum(p[:,2:4], axis=1, keepdims=True)/phi
        s=p0+p1
        self.preBias=np.ones((self.dataSet.shape[0],2))
        nPrecint=int(self.dataSet[-1,0])
        nVoterPerPrecint=int(self.dataSet[-1,1])
        print "############### Precint Preference Table ###############"
        print "-
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print "%-15s|%-20s|%-15s " %("Precint ID", "P(Yi=1/Xi)", "> 0.5")
                   for i in range(nPrecint):
                           p_1=1
                            p_{0}=1
                            for j in range(nVoterPerPrecint):
                                     p 1*=p0[i*nVoterPerPrecint+j]
                                     p_0*=p1[i*nVoterPerPrecint+j]
                            s=p_1+p_0
                            self.preBias[i*nVoterPerPrecint:(i+1)*nVoterPerPrecint,0]*=p 0/s
                            self.preBias[i*nVoterPerPrecint:(i+1)*nVoterPerPrecint,1]*=p_1/s
                            print \|-15s\| -20s\| -15s\| -15
                   print
         def initStep(self,labeled=False):
                   if labeled==True:
                            self.X=self.dataSet[:,3:5]
                            self.Z=self.dataSet[:,2]
                            self.Y=copy.deepcopy(self.ones)
                            N=self.dataSet.shape[0]
                            precinct=0
                            pointer=0
                            for i in range(N):
                                      if self.dataSet[i,0] != precinct :
                                               if np.sum(self.Z[pointer:i]) < 10 and i != 0 :</pre>
                                                         self.Y[pointer:i] = 0 *self.ones[pointer:i]
                                               precinct=self.dataSet[i,0]
                                               pointer=i
                            if np.sum(self.Z[precinct:N]) < 10 :</pre>
                                      self.Y[pointer:N] = 0 *self.ones[pointer:N]
                            self.Z=self.dataSet[:,2]
                            self.w[:,0]=np.array(((self.Z+self.Y) == self.zeros), dtype=int)
                            self.w[:,1]=np.array(((self.Z-self.Y) == self.ones), dtype=int)
self.w[:,2]=np.array(((self.Y-self.Z) == self.ones), dtype=int)
                            self.w[:,3]=np.array((np.multiply(self.Z,self.Y) == self.ones), dtype=in
t)
                            return 0
                   else:
                            # weight Values Would be set in E-Step
                            self.X=self.dataSet[:,2:4]
                            [mu0]=np.mean(self.X, axis=0, keepdims=True)
                            sigma0=np.cov(self.X.T)
                            mu1=mu0
                            sigma1=sigma0
                            phi=0.5
                            lembda=0.5
                            return [mu0.T, mu1.T, sigma0, sigma1, phi, lembda]
         def MStep(self):
                   return self.MLEstimator()
         def MLEstimator(self):
                  mu0=self.weightedMean((self.w[:,0]+self.w[:,2]),self.X)
                   mu1=self.weightedMean((self.w[:,1]+self.w[:,3]),self.X)
                  \label{eq:sigma0} \begin{split} &\text{sigma0} = \text{self.weightedCov}((\text{self.w[:,0]} + \text{self.w[:,2]}), \text{self.X,mu0}) \\ &\text{sigma1} = \text{self.weightedCov}((\text{self.w[:,1]} + \text{self.w[:,3]}), \text{self.X,mu1}) \end{split}
                   \label{eq:phi=sum((self.w[:,2]+self.w[:,3]))/self.w.shape[0]} phi=sum((self.w[:,2]+self.w[:,3]))/self.w.shape[0]
                   lembda=sum((self.w[:,0]+self.w[:,3]))/self.w.shape[0]
                   #self.precintBIAS(mu0, mu1, sigma0, sigma1, phi, lembda)
                   return [mu0, mu1, sigma0, sigma1, phi, lembda]
         def visualizeData(self, ax, mu0, mu1, sigma0, sigma1, phi, lembda):
                   [m, n] = self.w.shape
                   y0=sp.multivariate_normal(mu0, sigma0)
                   y1=sp.multivariate_normal(mu1, sigma1)
                   #print "DEBUG:",mu0
                   #print "DEBUG: , mu1
```

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```
#print "DEBUG:",sigma0
#print "DEBUG:",sigma1
#print "DEBUG:",lembda, phi
        #Defining Mesh for contour Plot
        x, y = np.mgrid[-3:3:0.1, -3:3:0.1]
pos = np.empty(x.shape + (2,))
        pos[:, :, 0] = x; pos[:, :, 1] = y
p_z_1=y1.pdf(self.X)*(1-lembda)
        p_z_2=y0.pdf(self.X)*lembda
        s=(p_z_1+p_z_2)
        p_z_1=np.divide(p_z_1,s)
        p_z_2=np.divide(p_z_1,s)
        p_z_1=p.multiply(p_z_1,self.preBias[:,0])
        p_z_2=np.multiply(p_z_2, self.preBias[:,1])
self.Z=np.array(((p_z_1+p_z_2)> 0.5), dtype=int)
#self.Z=np.array((p_z_1 > p_z_0), dtype=int)
        nPrecint=int(self.dataSet[-1,0])
        nVoterPerPrecint=int(self.dataSet[-1,1])
        plt.scatter(self.X[:,0], self.X[:,1], c=self.Z, alpha=0.8)
plt.scatter(mu0[0], mu0[1], s=70, c='yellow')
        plt.scatter(mu1[0], mu1[1], s=70, c='yellow')
        pie=np.mean(self.Z, axis=0, keepdims=True)
        # Mixture Contour
        plt.contour(x,y, (pie*y1.pdf(pos)+ (1-pie)* y0.pdf(pos)) )
        #ax.plot_surface(x,y, (pie*y1.pdf(pos)+ (1-pie)* y0.pdf(pos)) )
        #return [mu0, mu1, sigma0, sigma1, phi, lembda]
    def runEM(self, itr, initFromMLE=False, epsilon=1):
        self.LL=[]
        ########"
        [ mu0, mu1, sigma0, sigma1, phi, lembda]=self.initStep()
        if initFromMLE:
            mle.initStep(labeled=True)
            [ mu0, mu1, sigma0, sigma1, phi, lembda]=mle.MLEstimator()
        mu0*=epsilon
        mu1*=epsilon
        #sigma0*=epsilon
        #sigma1*=epsilon
        phi*=epsilon
        lembda*=epsilon
        for i in range(itr):
            self.EStep(mu0, mu1, sigma0, sigma1, phi, lembda)
            [mu0, mu1, sigma0, sigma1, phi, lembda]=self.MStep()
#print "DEBUG:", "ITR", i
        self.printEstimates(mu0, mu1, sigma0, sigma1, phi, lembda, epsilon)
        print "########### END
                                              EM Training on UNabled Set ###########
########"
        return [mu0, mu1, sigma0, sigma1, phi, lembda]
    ########"
        print "[LEMDA, PHI]: ",
        pprint.pprint([lembda, phi])
        print "mu0: "
        pprint.pprint(mu0)
        print "mu1: "
        pprint.pprint(mu1)
        print "sigma0: "
        pprint.pprint(sigma0)
        print "sigmal:
```

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```
pprint.pprint(sigma1)
     #########
pp=PdfPages('./Model2.pdf')
fig=plt.figure()
# Train
##"
mle=analyZeData('./surveylabeled.dat')
mle.initStep(labeled=True)
[mu0, mu1, sigma0, sigma1, phi, lembda]=mle.MLEstimator()
##"
print ""
mle.printEstimates(mu0, mu1, sigma0, sigma1, phi, lembda)
# Predict/Label
mle2=analyZeData('./surveyunlabeled.dat')
mle2.initStep(labeled=False)
mle2.precintBIAS(mu0, mu1, sigma0, sigma1, phi, lembda)
ax=fig.add subplot(121)
plt.title("MODEL-2: Distribution - MLE")
mle2.visualizeData(ax, mu0, mu1, sigma0, sigma1, phi, lembda)
##"
print ""
#ax=fig.add subplot(122, projection='3d')
#exit()
em=analyZeData('./surveyunlabeled.dat')
[mu0, mu1, sigma0, sigma1, phi, lembda]=em.runEM(10, initFromMLE=True)
print
em.precintBIAS(mu0, mu1, sigma0, sigma1, phi, lembda)
ax=fig.add_subplot(122)
plt.title("MODEL2: Distribution - EM")
em.visualizeData(ax, mu0, mu1, sigma0, sigma1, phi, lembda)
plt.savefig(pp, format='pdf')
plt.show()
plt.title("MODEL-2: Log Liklihood Over Iterations")
plt.plot(em.LL, label="INIT=MLE")
em.runEM(10, initFromMLE=True, epsilon=0.9)
plt.plot(em.LL, label="INIT=MLE - 10%")
em.runEM(10, initFromMLE=True, epsilon=0.6)
plt.plot(em.LL, label="INIT=MLE - 40%")
plt.legend()
plt.savefig(pp, format='pdf')
plt.show()
pp.close()
```