```
In [146]: #Multi class classification strategy
           #Comparing original kesler (one vs rest or one vs all)
          #all vs all method (ie nicking anv 2 classes at a time)
In [147]: import os, struct
           import matplotlib as plt
           from array import array as pyarray
           import numpy as np
           import time
           from pylab import *
           import pandas as pd
           from scipy.stats import multivariate_normal
           import sys
In [148]: X = np.array([2,2])
           X = np.row_stack((X,[2,-2]))
           X = np.row_stack((X,[-2,2]))
           C = ['red','blue','black']
           print "X",X
           X [[ 2 2]
           [ 2 -2]
[-2 2]]
In [149]: | fig = plt.figure()
           plt.scatter(X[0,0],X[0,1],s=100,color='red')
           plt.scatter(X[1,0],X[1,1],s=100,color='blue')
           plt.scatter(X[2,0],X[2,1],s=100,color='black')
           plt.ylim([-3,3])
           plt.xlim([-3,3])
          nlt.show()
            2
            1
            0
           -1
           -2
                    -2
                           -1
                                        i
In [150]: T = np.array([1, -1, -1])
           T = np.row_stack((T,[-1,1,-1]))
           T = np.row_stack((T,[-1,-1,1]))
          nrint "T" T
          T [[ 1 -1 -1]
[-1 1 -1]
[-1 -1 1]]
```

```
In [151]: 0 = np.array([1,1,1])
          Xa = np.column_stack((0,X))
          nrint "Xa" Xa
          Xa [[ 1 2 2]
           [ 1 2 -2]
           [ 1 -2 2]]
In [152]: Y = np.linalg.pinv(Xa)
          Wr = np.dot(Y,T[:,0])
          nrint Wr
          [-1.
                 0.5 0.5]
In [153]: xx = np.arange(-3,3,0.1)
          yy = (-Wr[0] - Wr[1]*xx)/Wr[2]
          fig = plt.figure()
          plt.plot(xx,yy)
          plt.scatter(X[0,0],X[0,1],s=100,color='red')
          plt.scatter(X[1,0],X[1,1],s=100,color='grey')
          plt.scatter(X[2,0],X[2,1],s=100,color='grey')
          plt.xlim(-3,3)
          plt.ylim(-3,3)
          nlt show()
            3
            2
            1
            0
           -1
           -2
           −3 <del>+</del>
−3
                   -2
                          -1
In [154]: Y = np.linalg.pinv(Xa)
          Wu = np.dot(Y,T[:,1])
          nrint Wu
          [ 2.77555756e-16 -5.55111512e-17 -5.00000000e-01]
  In [ ]:
```

```
In [155]: xx = np.arange(-3,3,0.1)
          yy = (-Wu[0] - Wu[1]*xx)/Wu[2]
          fig = plt.figure()
          plt.plot(xx,yy)
          plt.scatter(X[0,0],X[0,1],s=100,color='grey')
          plt.scatter(X[1,0],X[1,1],s=100,color='blue')
          plt.scatter(X[2,0],X[2,1],s=100,color='grey')
          plt.xlim(-3,3)
          plt.ylim(-3,3)
          nlt show()
            2
            1
           -1
           -2
           -3
                          -1
                   -2
                                 Ò
In [156]: Y = np.linalg.pinv(Xa)
          Wk = np.dot(Y,T[:,2])
          nrint Wk
          [ 1.11022302e-16 -5.00000000e-01
                                                8.59551441e-17]
In [157]: xx = np.arange(-3,3,0.1)
          yy = (-Wk[0] - Wk[1]*xx)/Wk[2]
          fig = plt.figure()
          plt.plot(xx,yy)
          plt.scatter(X[0,0],X[0,1],s=100,color='grey')
          plt.scatter(X[1,0],X[1,1],s=100,color='grey')
          plt.scatter(X[2,0],X[2,1],s=100,color='black')
          plt.xlim(-3,3)
          plt.ylim(-3,3)
          nlt show()
            2
            1
            0
           -1
           -2
           -3
                          -1
                   -2
```

```
In [158]: xx = np.arange(-3,3,0.1)
          yyk = (-Wk[0] - Wk[1]*xx)/Wk[2]
          yyu = (-Wu[0] - Wu[1]*xx)/Wu[2]
          yyr = (-Wr[0] - Wr[1]*xx)/Wr[2]
          fig = plt.figure()
          plt.plot(xx,yyk,color='black')
          plt.plot(xx,yyu,color='blue')
          plt.plot(xx,yyr,color='red')
          Z = [1, 0.5, 0.5]
          plt.scatter(Z[1],Z[2],s=100,color='grey')
          plt.xlim([-3,3])
          plt.ylim([-3,3])
          plt.show()
            2
            1
                                    •
            0
           -1
           -2
           -3
                   -2
                          -1
  In [ ]:
  In [ ]:
In [159]:
          Tr = np.dot(Z,Wr)
          Tu = np.dot(Z,Wu)
          Tk = np.dot(Z,Wk)
          Tpred=[Tr,Tu,Tk]
          print "Tpred", Tpred
          #first of all, Tr is the most -ve =>not red , it should be the other way round
          #hesides. Tk and Tu have the same value, so inconclusive
          Tpred [-0.50000000000000022, -0.24999999999991, -0.24999999999978]
In [160]: XRU = np.array([2,2])
          XRU = np.row_stack((XRU,[2,-2]))
          TRU = np.array([1,-1])
          TRU = np.row_stack((TRU,[-1,1]))
          print "XRU\n", XRU
print "TRU\n", TRU
          XRU
          [[ 2 2]
           [ 2 -2]]
          TRU
          [[ 1 -1]
           [-1 1]]
```

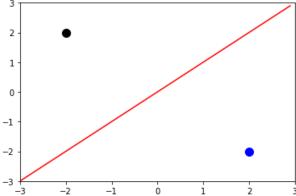
```
In [161]: | XaRU = np.column_stack(([1,1],XRU))
           YRU = np.linalg.pinv(XaRU)
          WRU = np.dot(YRU,TRU[:,0])
          nrint "WRII" WRII
          WRU [ 4.16333634e-17 -2.77555756e-17
                                                      5.0000000e-01]
In [162]: xx = np.arange(-3,3,0.1)
           yy = (-WRU[0] - WRU[1]*xx)/WRU[2]
           fig = plt.figure()
           plt.plot(xx,yy,color='black')
           plt.scatter(2,2,s=100,color='red')
           plt.scatter(2,-2,s=100,color='blue')
           plt.xlim([-3,3])
           plt.ylim([-3,3])
          nlt show()
            2
            1
            0
           -1
           -2
           -3
             -3
                    -2
                          -1
                                  0
                                        1
In [163]:
          XRK = np.array([2,2])
          XRK = np.row_stack((XRK,[-2,2]))
           TRK = np.array([1,-1])
           TRK = np.row_stack((TRK,[-1,1]))
          print "XRK\n", XRK
print "TRK\n" TRK
          XRK
          [[2 2]
           [-2 2]]
          TRK
          [[1 -1]
           [-1 \quad 1]
In [164]: XaRK = np.column stack(([1,1],XRK))
           YRK = np.linalg.pinv(XaRK)
          WRK = np.dot(YRK,TRK[:,0])
          nrint "WRK". WRK
          WRK [ 4.16333634e-17
                                    5.00000000e-01
                                                      2.77555756e-17]
```

```
In [165]: xx = np.arange(-3,3,0.1)
           yy = (-WRK[0] - WRK[1]*xx)/WRK[2]
           fig = plt.figure()
           plt.plot(xx,yy,color='blue')
           plt.scatter(2,2,s=100,color='red')
           plt.scatter(-2,2,s=100,color='black')
           plt.xlim([-3,3])
           plt.ylim([-3,3])
          nlt show()
             2
             1
             0
            -1
            -2
            -3 -
              -3
                    <u>-2</u>
                           -1
In [166]: |XKU = np.array([-2,2])
           XKU = np.row_stack((XKU,[2,-2]))
           TKU = np.array([1,-1])
           TKU = np.row_stack((TKU,[-1,1]))
          print "XKU\n", XKU
print "TKII\n" TKII
          XKU
           [[-2 2]
           [ 2 -2]]
           TKU
           [[ 1 -1]
           [-1 1]]
In [167]: XaKU = np.column_stack(([1,1],XKU))
           YKU = np.linalg.pinv(XaKU)
           WKU = np.dot(YKU,TKU[:,0])
          nrint "WKII" WKII
```

2.50000000e-01]

```
6 of 10 06/11/2017 09:49 PM
```

WKU [-5.55111512e-17 -2.50000000e-01



```
In [169]: Z = [1, 0.5, 0.5]
    TRK = np.dot(Z,WRK)
    print "TRK",TRK
    TKU = np.dot(Z,WKU)
    print "TKU", TKU
    TRU = np.dot(Z,WRU)
    print "TRU", TRU

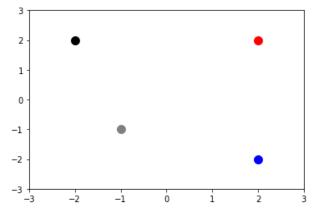
#TPN - P is positive class N is negarive class If TPN > 0, class is P.
#Then take a majority vote

TRK 0.25
    TKU -5.55111512313e-17
    TRU 0.25
```

```
In [170]: Z = [1, -1,-1]
    fig = plt.figure()
    plt.scatter(2,2,s=100,color='red')
    plt.scatter(2,-2,s=100,color='blue')
    plt.scatter(-2,2,s=100,color='black')
    plt.scatter(Z[1],Z[2],s=100,color='grey')

plt.xlim(-3,3)
    plt.ylim(-3,3)

plt.show()
```



```
In [171]: Tpred_onevsall = [np.dot(Z,Wr),np.dot(Z,Wu),np.dot(Z,Wk)]
    print Tpred_onevsall
    #inconclusive, but atleast strongly confirms not red .
#expected since equally close to black and blue but very far from red
[-2.00000000000000004, 0.500000000000044, 0.4999999999999999]
```

```
In [172]: TRK = np.dot(Z,WRK)
   TKU = np.dot(Z,WKU)
   TRU = np.dot(Z,WRU)
   print "TRK", TRK
   print "TKU", TKU
   print "TRU", TRU
# gives similar result
```

TRK -0.5 TKU -5.55111512313e-17 TRU -0.5

```
In [173]: #How about computation time?
    #original method : N points * K kesler outputs = N*K,
    #with pseudo inv matrix based classifier. It is almost close to N*1
    #allvsall method
    #Since this involves KC2 number of classifications performed over N/K points =
    #what is the impact of fewer training points?
```

```
In [182]: n = np.arange(1,8,1)
          time_linear_onevsall = [0.96, 1.04, 0.99, 0.89, 0.91, 1.06, 0.78]
          time_linear_allvsall = [34.08,36.74,39.2,39.25,41.22,40,38.08]
          time_bayes_onevsall = [178.575,195.88,229,225,223,217,229]
          time_bayes_allvsall = [526.8,648.8,663,687,706,687,1020]
          acc_{inear\_onevsall} = [14.31, 22.53, 38.77, 41.84, 52.55, 62.66, 63.07]
          acc linear allvsall = [30.7,47.3,68.3,82.75,83.14,84.33,85.57]
          acc_bayes_onevsall = [15.76,26.3,45.24,57.82,71.53,78.06,79.85]
          acc_bayes_allvsall = [31.33,48.14,68.81,82.86,83.7,85.42,86.11]
          plt.plot(n,acc_linear_onevsall,label="acc_linear_onevsall")
          plt.plot(n,acc linear allvsall,label="acc linear allvsall")
          plt.plot(n,acc bayes onevsall,label="acc bayes onevsall")
          plt.plot(n,acc bayes allvsall,label="acc bayes allvsall")
          plt.ylabel("Accuracy in %")
          plt.xlabel("Number of Principle Component")
          plt.title("accuracy of different classifiers")
          plt.grid("on")
          plt.legend()
          plt.show()
          fig = plt.figure()
          plt.plot(n,time_linear_onevsall,label="time_linear_onevsall")
          plt.plot(n,time_linear_allvsall,label="time_linear_allvsall")
          plt.plot(n,time_bayes_onevsall,label="time_bayes_onevsall")
          plt.plot(n,time_bayes_allvsall,label="time_bayes_allvsall")
          plt.ylabel("Time in seconds")
          plt.xlabel("Number of Principle Component")
          plt.title("run time of different classifiers")
          plt.grid("on")
          plt.legend()
          nlt.show()
```

