## DATA MINING ASSIGNMENT – 2 Pramod Maharjan (069-BCT-530)

## Singular Value Decomposition

SVD is based on a theorem from linear algebra which says that a rectangular matrix A can be broken down into the product of three matrices - an orthogonal matrix U, a diagonal matrix S, and the transpose of an orthogonal matrix V. The theorem is usually presented something like this:

Amn = Umm Smn V T nn

where UTU = I, VTV = I; the columns of U are orthonormal eigenvectors of AAT, the columns of V are orthonormal eigenvectors of ATA, and S is a diagonal matrix containing the square roots of eigenvalues from U or V in descending order.

The code for SVD is given below:

```
mport math
import numpy as np
from numpy import linalg as LA
def SVD(mat):
        matT = mat.transpose()
        matmatT = mat.dot(matT)
        matTmat = matT.dot(mat)
        egnvalU, egnvecU = LA.eigh(matmatT)
        egnvalV, egnvecV = LA.eigh(matTmat)
        V = np.fliplr(egnvecV)
        VT = V.transpose()
        egnvalV = egnvalV[::-1]
        S = np.zeros(mat.shape)
        for i in range(min(mat.shape)):
                 S[i][i] = math.sqrt(egnvalV[i])
        U = np.dot(np.dot(mat,np.transpose(VT)),LA.pinv(S))
        return U, S, VT
def reduceDim(u,s,v,num):
        for i in range(num):
                 s = np.delete(s, s.shape[1]-1, axis = 1
                 s = np.delete(s, s.shape[1]-1, axis = 1)
v = np.delete(v, v.shape[0]-1, axis = 0)
v = np.delete(v, v.shape[1]-1, axis = 1)
        return u, s, v
def main():
        mat = np.random.rand(5, 5)
        u, s, v = SVD(mat)
        U, S, V = reduceDim(u,s,v, 3)
        red = np.dot(np.dot(U,S),V)
                    ",u,"\n\nS = ",s,"\n\nV =", v, "\n\nReformed =
        print "
if __name__==" main ":
        main()
```

Matrix of size 5\*5 is taken where 5 column represents the five dimension of data and rows represent the data samples and it is decomposed in matrices U S and V as below:

```
pramod@pramod-X555LN:~$ python SVD.py
Original = [[ 0.70377037
                          0.81180308
                                      0.41679331
                                                  0.98820843 0.96367014]
  0.31150079
              0.8514664
                                                  0.2461971
                          0.79478207
                                      0.39829415
  0.5753268
              0.01717992 0.66748664
                                      0.95795825
                                                  0.849805157
  0.04199565 0.64147577 0.49737969 0.83868295
                                                  0.75419044]
                                                  0.77645566]]
  0.9962875
              0.689222
                          0.67109168 0.7432223
    [[-0.53342587 -0.04383926 0.05345115 0.64996678 0.53686229]
 [-0.33508265 0.74467889 -0.05815887 -0.48017405
                                                  0.31499747]
                          0.0170199 -0.57863311
                                                  0.21730547]
 [-0.4306072
              -0.65745709
 [-0.39118225
             0.10053301 0.77788068 0.01910548 -0.48104698]
 [-0.51451954
              0.03427479 -0.62319539 0.10860375 -0.5778647 ]]
 = [[ 3.31810092
                   0.
                                                                 1
                          0.
  0.
              0.83002312
                                      0.
                                                  0.
                          0.6305025
                                                  0.
  0.
              0.
                                      0.
  0.
              0.
                          0.
                                      0.48023182
                                                  0.
                                                  0.04563613]]
  0.
                          0.
                                      0.
V = [[-0.37870006 - 0.40122297 - 0.39659024 - 0.53753078 - 0.49938285]
 [-0.16718558 0.81358887 0.25028977 -0.32137468 -0.37973079]
 [-0.88646971 0.10092826 -0.06963362 0.37300954
                                                  0.244948951
  0.17481668  0.40805268  -0.86328314  -0.0135644
                                                  0.23977147]
              0.01999437 -0.17310281 0.68444055 -0.69923598]]
  0.11066087
pramod@pramod-X555LN:~$
```

The matrix after reduction of dimension in 2 is given below:

(Similarly it cand be done for 100 data samples which need 100\*5 sized matrix )

```
pramod@pramod-X555LN:~$ python SVD.py
After Reduction of Dimension
U = [[-0.5116032  0.06584604  0.67272944  0.34610089  0.40196484]
[-0.4527084 -0.57978669 0.1623556 -0.62016189 -0.2189575 ]
[-0.61195147 -0.10870048 -0.5581509 0.49010458 -0.24892706]
 [-0.19865099  0.64162864  0.26546758  -0.14670516  -0.67591052]
[-0.34550029 0.48580714 -0.3729213 -0.4836202
                                                     0.52121222]]
S = [[2.47387804 0.
                                0.
                                                          0.
                                             0.
                                                                    ]
               0.76159161 0.
[ 0.
                                        0.
                                                     0.
[ 0.
               0.
                           0.43480272 0.
                                                     0.
                           0.
               0.
                                        0.30319771 0.
 [ 0.
               0.
                           0.
                                        0.
                                                     0.13135979]]
V = [[-0.38422109 - 0.55935335 - 0.55524804 - 0.23254048 - 0.42085927]
[ 0.64048035 -0.35222782 -0.48834575  0.12016862  0.46130079]
[ 0.17838653  0.70555258 -0.5376929 -0.38705489 -0.17733768]]
Reformed = [[ 0.51840576  0.69027872]
[ 0.14749569  0.78197514]
 [ 0.52864733  0.87596061]
[ 0.50179744  0.10276838]
[ 0.56537277  0.3477740<u>3</u>]]
pramod@pramod-X555LN:~$
```