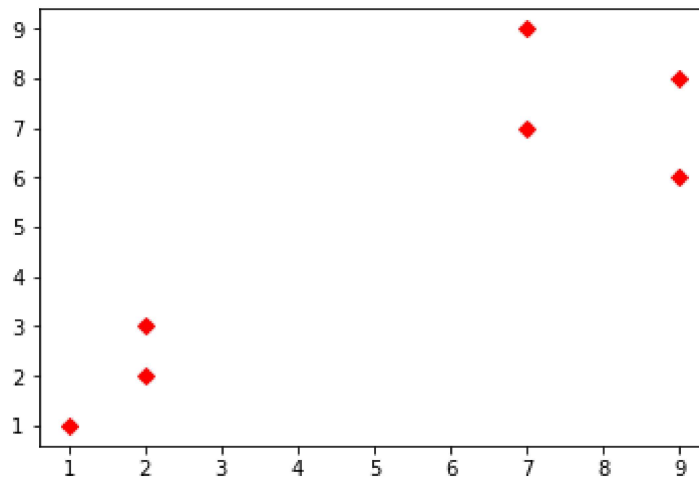


Import the libraries

```
In [7]: import numpy as np
        from scipy import linalg
        import math
        from sklearn.cluster import KMeans
        import matplotlib.pyplot as plt
```

(1) plot the data

```
In [8]: A = np.array([[1.0,1.0],[2.0,2.0],[2.0,3.0],[7.0,9.0],[7.0,7.0],[9.0,8.0],[9.0,6.0]])
        for point in A:
            plt.plot(point[0],point[1],'rD-')
```



Euclidean distance function

```
In [9]: def dist(vec1,vec2):
        # computing the distance, vec1 and vec2 should have the same size; c is the centroid; Ai is the element in A
        d=np.linalg.norm(vec1-vec2)
        return d
```

```
In [10]: a = dist(A[0],A[1])
        a
```

```
Out[10]: 1.4142135623730951
```

e-neighborhood function

```
In [11]: def epsilon(vi,vj,e):
          s=dist(vi,vj)
          if s<0.5:
              return 0
          else:
              return s
```

Gaussian kernel distance function

```
In [20]: d = matrixM(A)
          #for i in range(7):
          #print(d[i][i])
          dd = matrixMM(A)
          d==dd
```

```
Out[20]: array([[ True,  True,  True,  True,  True,  True,  True],
                [ True,  True,  True,  True,  True,  True,  True],
                [ True,  True,  True,  True,  True,  True,  True],
                [ True,  True,  True,  True,  True,  True,  True],
                [ True,  True,  True,  True,  True,  True,  True],
                [ True,  True,  True,  True,  True,  True,  True],
                [ True,  True,  True,  True,  True,  True,  True]])
```

```
In [19]: def matrixMM(A):
          (m,n)=np.shape(A)
          b=np.zeros((m,m))
          for i in range(m):
              for j in range(m):
                  b[i][j]=np.linalg.norm(A[i]-A[j])
          return b
```

```
In [12]: def matrixM(A):
          (m,n)=np.shape(A)
          b=np.zeros((m,m))
          for i in range(m):
              for j in range(m):
                  b[i][j]=dist(A[i],A[j])
          return b
```

```
In [21]: def matrixS(M):
          (m,n)=np.shape(M)
          b=np.zeros((m,m))
          for i in range(m):
              for j in range(m):
                  b[i][j]=math.exp((-np.square(M[i][j]))/(2*np.square(2.1)))
          return b
```

```
In [22]: def matrixW(S):
          (m,n)=np.shape(S)
          b=np.zeros((m,m))
          for i in range(m):
              for j in range(m):
                  if S[i][j]<0.5 or S[i][j]==1:
                      b[i][j]=0
                  else:
                      b[i][j]=S[i][j]
          return b
```

```
In [23]: #diag matr
          def matrixD(A):
              (m,n)=np.shape(A)
              D=np.zeros((m,m))
              for i in range(m):
                  for j in range(m):
                      if A[i][j] != 0:
                          D[i][i]+=A[i][j]
          return D
```

(2) generate distance matrix M

```
In [24]: M=matrixM(A)
```

(3) Print the matrix with at most 2 decimal places

```
In [25]: np.set_printoptions(precision=2) #print("%.2f" % M)
          print(M)
```

```
[[ 0.    1.41  2.24 10.    8.49 10.63  9.43]
 [ 1.41  0.    1.    8.6   7.07  9.22  8.06]
 [ 2.24  1.    0.    7.81  6.4   8.6   7.62]
 [10.    8.6   7.81  0.    2.    2.24  3.61]
 [ 8.49  7.07  6.4   2.    0.    2.24  2.24]
 [10.63  9.22  8.6   2.24  2.24  0.    2.   ]
 [ 9.43  8.06  7.62  3.61  2.24  2.    0.   ]]
```

(4) generate similarity matrix S based on M

```
In [41]: S=matrixS(M)
```

(5) Print the similarity matrix S with at most 2 decimal places

```
In [42]: np.set_printoptions(suppress=True)
np.set_printoptions(precision=2)
print(S)#(5)
```

```
[[1.  0.8 0.57 0.  0.  0.  0. ]
 [0.8 1.  0.89 0.  0.  0.  0. ]
 [0.57 0.89 1.  0.  0.01 0.  0. ]
 [0.  0.  0.  1.  0.64 0.57 0.23]
 [0.  0.  0.01 0.64 1.  0.57 0.57]
 [0.  0.  0.  0.57 0.57 1.  0.64]
 [0.  0.  0.  0.23 0.57 0.64 1.  ]]
```

(6) generate weight matrix W based on S

```
In [43]: W= matrixW(S)
print(W)
```

```
[[0.  0.8 0.57 0.  0.  0.  0. ]
 [0.8 0.  0.89 0.  0.  0.  0. ]
 [0.57 0.89 0.  0.  0.  0.  0. ]
 [0.  0.  0.  0.  0.64 0.57 0. ]
 [0.  0.  0.  0.64 0.  0.57 0.57]
 [0.  0.  0.  0.57 0.57 0.  0.64]
 [0.  0.  0.  0.  0.57 0.64 0.  ]]
```

(7) Plot the graph associated to W

(8) Compute the Laplacian matrix L based on W

```
In [45]: D=matrixD(W)
print(D)
print()
L=D-W
print(L)
```

```
[[1.36 0.  0.  0.  0.  0.  0. ]
 [0.  1.69 0.  0.  0.  0.  0. ]
 [0.  0.  1.46 0.  0.  0.  0. ]
 [0.  0.  0.  1.2  0.  0.  0. ]
 [0.  0.  0.  0.  1.77 0.  0. ]
 [0.  0.  0.  0.  0.  1.77 0. ]
 [0.  0.  0.  0.  0.  0.  1.2  ]]
```



```
[[ 1.36 -0.8 -0.57 0.  0.  0.  0. ]
 [-0.8  1.69 -0.89 0.  0.  0.  0. ]
 [-0.57 -0.89 1.46 0.  0.  0.  0. ]
 [ 0.  0.  0.  1.2 -0.64 -0.57 0. ]
 [ 0.  0.  0. -0.64 1.77 -0.57 -0.57]
 [ 0.  0.  0. -0.57 -0.57 1.77 -0.64]
 [ 0.  0.  0.  0. -0.57 -0.64 1.2  ]]
```

(9) Compute the eigenvalues and eigenvectors of L

```
In [66]: eivalue,eivector=np.linalg.eig(L)
print(eivalue)
print(eivector)
```

```
[-0.      1.97  2.55  0.      1.2   2.41  2.34]
[[ 0.58  0.76  0.3   0.      0.      0.      0. ]
 [ 0.58 -0.12 -0.81  0.      0.      0.      0. ]
 [ 0.58 -0.64  0.51  0.      0.      0.      0. ]
 [ 0.      0.      0.      0.5   0.71 -0.5   -0.04]
 [ 0.      0.      0.      0.5   0.04  0.5    0.71]
 [ 0.      0.      0.      0.5  -0.04  0.5   -0.71]
 [ 0.      0.      0.      0.5  -0.71 -0.5    0.04]]
```

(10) Sort the eigenvalues and eigenvectors in increasing order

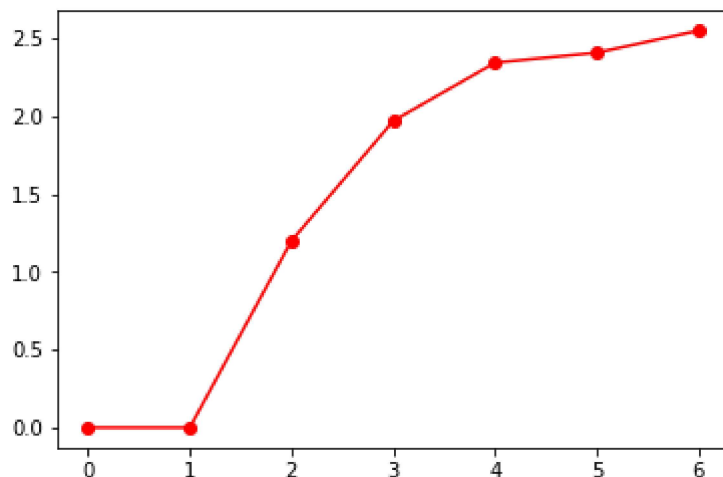
```
In [76]: eivalue_sorted=np.sort(eivalue)
res=sorted(zip(eivalue, eivector))
print(res[1][1])
```

```
[ 0.      0.      0.      0.5   0.71 -0.5   -0.04]
```

(11) Plot the eigenvalues in increasing order

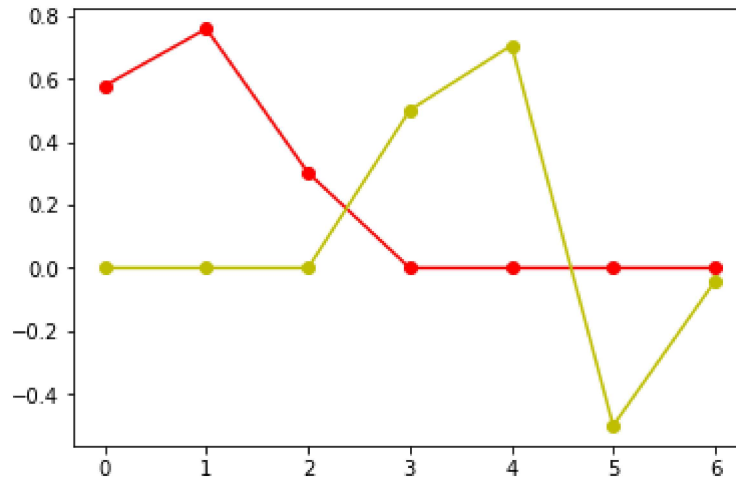
```
In [78]: plt.plot(eivalue_sorted,'ro-')
```

```
Out[78]: [<matplotlib.lines.Line2D at 0x236e409bf98>]
```

**(12) Plot the first eigenvector and the second eigenvector**

```
In [81]: plt.plot(res[0][1], 'ro-')
plt.plot(res[1][1], 'yo-')
```

```
Out[81]: [<matplotlib.lines.Line2D at 0x236e5476630>]
```



(14) Propose a thresholding mechanism based on the mean value of the second eigenvector

```
In [96]: # mean value of the second eigenvector named "vec"
vec=res[1][1]
mean=np.mean(vec)
print(mean)
threshold=mean
print(vec)
group1 = []
group2 = []
for i in range(len(vec)):
    if vec[i]<threshold:
        group1.append(i)
    else:
        group2.append(i)
```

```
0.094803769799918
```

```
[ 0.    0.    0.    0.5   0.71 -0.5  -0.04]
```

(15) Draw the clustering

```
In [99]: for i in group1:
          plt.plot(i,vec[i],'bo')
         for i in group2:
           plt.plot(i,vec[i],'ro')
```

