

# Data Mining; Assignemt 6

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**Find step by step procedure of 1-5 in attached excel file.**

1. K-means algorithm

Starting with k=4 clusters. I selected  $(0,7)$ ,  $(4,4)$ ,  $(0,0)$ , and  $(9,9)$  as my initial cluster centers.

$d(x,y)$  = Manhattan distance.

First Iteration:

Cluster centers			
$(0, 7)$ cluster 1	$(4, 4)$ cluster 2	$(0, 0)$ cluster 3	$(9, 9)$ cluster 4
$(0,7)$	$(2,5)$	$(1,1)$	$(7,8)$
$(1,6)$	$(3,6)$	$(3,0)$	$(9,9)$
$(1,8)$	$(5,3)$		
$(2,7)$	$(6,2)$		
$(2,8)$	$(6,4)$		
$(3,7)$	$(7,2)$		
	$(7,3)$		
	$(7,5)$		
	$(8,3)$		
	$(8,4)$		

Second Iteration:

Cluster centers			
$(1.5, 7.16)$ cluster 1	$(5.9, 3.7)$ cluster 2	$(2, 0.5)$ cluster 3	$(8, 8.5)$ cluster 4
$(0,7)$	$(5,3)$	$(1,1)$	$(7,8)$
$(1,6)$	$(6,2)$	$(3,0)$	$(9,9)$
$(1,8)$	$(6,4)$		
$(2,5)$	$(7,2)$		
$(2,7)$	$(7,3)$		
$(2,8)$	$(7,5)$		
$(3,6)$	$(8,3)$		
$(3,7)$	$(8,4)$		

Third Iteration:

Cluster centers			
(1.75, 6.75)	(6.75, 3.25)	(2, 0.5)	(8, 8.5)
cluster 1	cluster 2	cluster 3	cluster 4
(0,7)	(5,3)	(1,1)	(7,8)
(1,6)	(6,2)	(3,0)	(9,9)
(1,8)	(6,4)		
(2,5)	(7,2)		
(2,7)	(7,3)		
(2,8)	(7,5)		
(3,6)	(8,3)		
(3,7)	(8,4)		

Algorithm terminated since the clusters where the same.

2. K-medoid algorithm

Starting with  $k=4$  clusters. I selected  $(1,8)$ ,  $(1,1)$ ,  $(2,5)$ , and  $(7,8)$  as my initial cluster centers.

$d(x,y)$  = Manhattan distance.

First Iteration:

Cluster centers			
(1, 8)	(1, 1)	(2, 5)	(7, 8)
cluster 1	cluster 2	cluster 3	cluster 4
(0,7)	(1,1)	(1,6)	(7,2)
(1,8)	(3,0)	(2,5)	(7,3)
(2,7)	(6,2)	(3,6)	(7,5)
(2,8)		(5,3)	(7,8)
(3,7)		(6,4)	(8,3)
			(8,4)
			(9,9)
<i>Cost = 59</i>			

Second Iteration:

Cluster centers			
(1, 8)	(1, 1)	(5, 3)	(7, 8)
cluster 1	cluster 2	cluster 3	cluster 4
(0,7)	(1,1)	(5,3)	(7,5)
(1,6)	(3,0)	(6,2)	(7,8)
(1,8)		(6,4)	(9,9)
(2,5)		(7,2)	
(2,7)		(7,3)	
(2,8)		(8,3)	
(3,6)		(8,4)	
(3,7)			
<i>Cost = 43</i>			

Third Iteration:

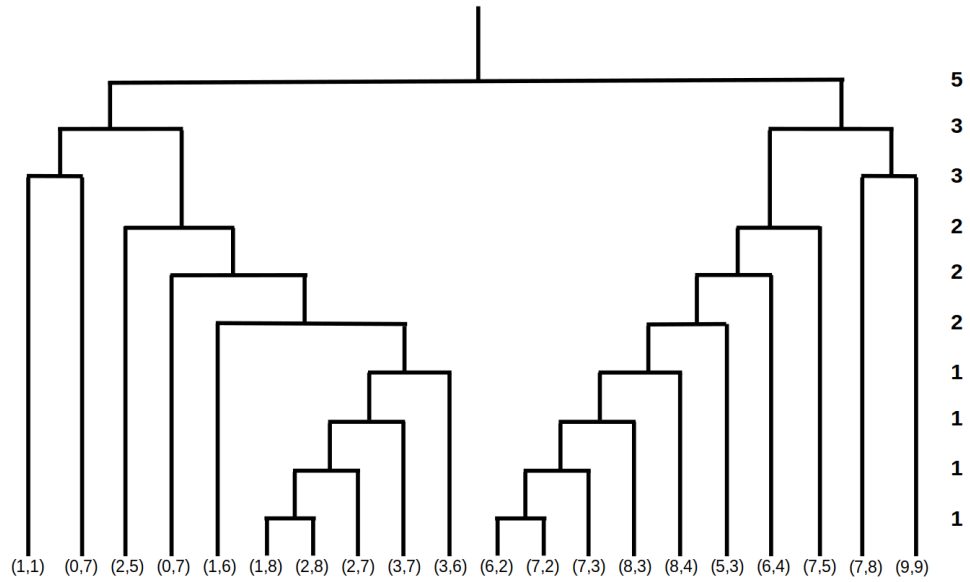
Cluster centers			
(1, 8)	(1, 1)	(7, 3)	(7, 8)
cluster 1	cluster 2	cluster 3	cluster 4
(0,7)	(1,1)	(5,3)	(7,8)
(1,6)	(3,0)	(6,2)	(9,9)
(1,8)		(6,4)	
(2,5)		(7,2)	
(2,7)		(7,3)	
(2,8)		(7,5)	
(3,6)		(8,3)	
(3,7)		(8,4)	
<i>Cost = 36</i>			

Forth Iteration:

Cluster centers			
(1, 8)	(1, 1)	(7, 3)	(7, 8)
cluster 1	cluster 2	cluster 3	cluster 4
(0,7)	(1,1)	(5,3)	(7,8)
(1,6)	(3,0)	(6,2)	(9,9)
(1,8)		(6,4)	
(2,5)		(7,2)	
(2,7)		(7,3)	
(2,8)		(7,5)	
(3,6)		(8,3)	
(3,7)		(8,4)	
<i>Cost = 30</i>			

Algorithm terminated because of the reduced cost and unchanged clusters.

3. AGNES algorithm  
 $d(x,y) = \text{Manhattan distance}.$



Algorithm terminated because every single cluster joined the hierarchical tree.

4. DBSCAN algorithm  
 $d(x,y)$  = Euclidean distance.

Core points					
(1,6)	(1,8)	(2,7)	(2,8)	(3,6)	(3,7)
(6,2)	(6,4)	(7,2)	(7,3)	(8,3)	(8,4)
Border points					
(0,7)	(2,5)	(5,3)	(7,5)		

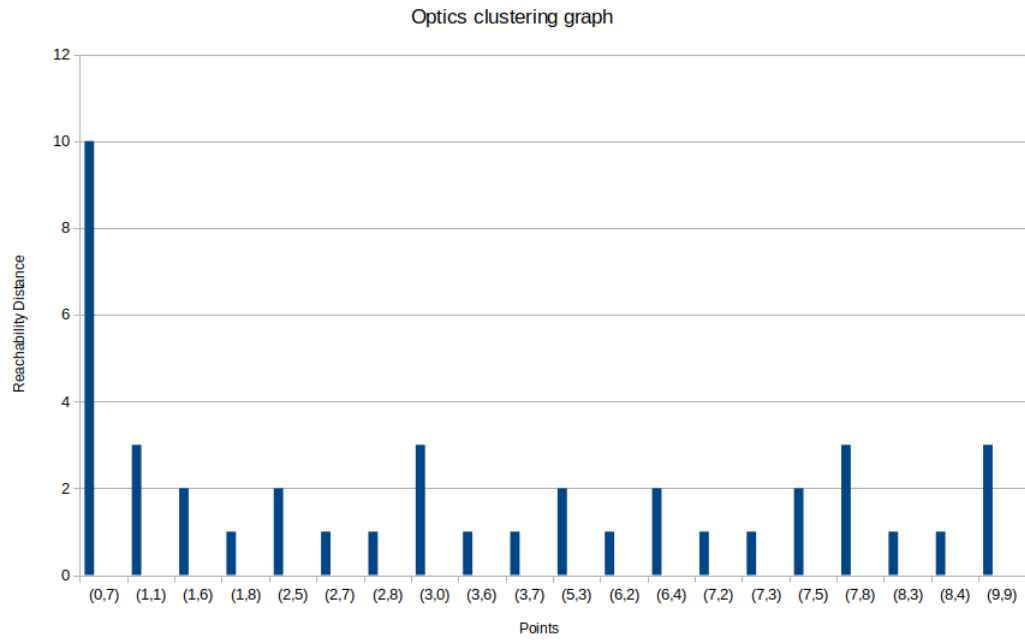
Clustering core points.

Core points						
Cluster 1	(1,6)	(1,8)	(2,7)	(2,8)	(3,6)	(3,7)
Cluster 2	(6,2)	(6,4)	(7,2)	(7,3)	(8,3)	(8,4)

Adding border points to the clusters

Cluster points								
Cluster 1	(0,7)	(1,6)	(1,8)	(2,5)	(2,7)	(2,8)	(3,6)	(3,7)
Cluster 2	(5,3)	(6,2)	(6,4)	(7,2)	(7,3)	(7,5)	(8,3)	(8,4)
Outlier points								
	(1,1)	(3,0)	(7,8)	(9,9)				

5. OPTICS algorithm  
 $d(x,y)$  = Manhattan distance.





6. Spectral clustering algorithm

*Kindly check attached jupyter notebook file.*