

Clustering

CSCI 347,
Data Mining

K-means Clustering

1. Input k to indicate how many clusters are wanted
2. K points are randomly chosen within the space. These serve as the cluster centers
3. Loop while cluster centers are changing
 - a. All instances are assigned to their closest cluster center
 - b. Calculate the mean point of each cluster

Classical k-means clustering

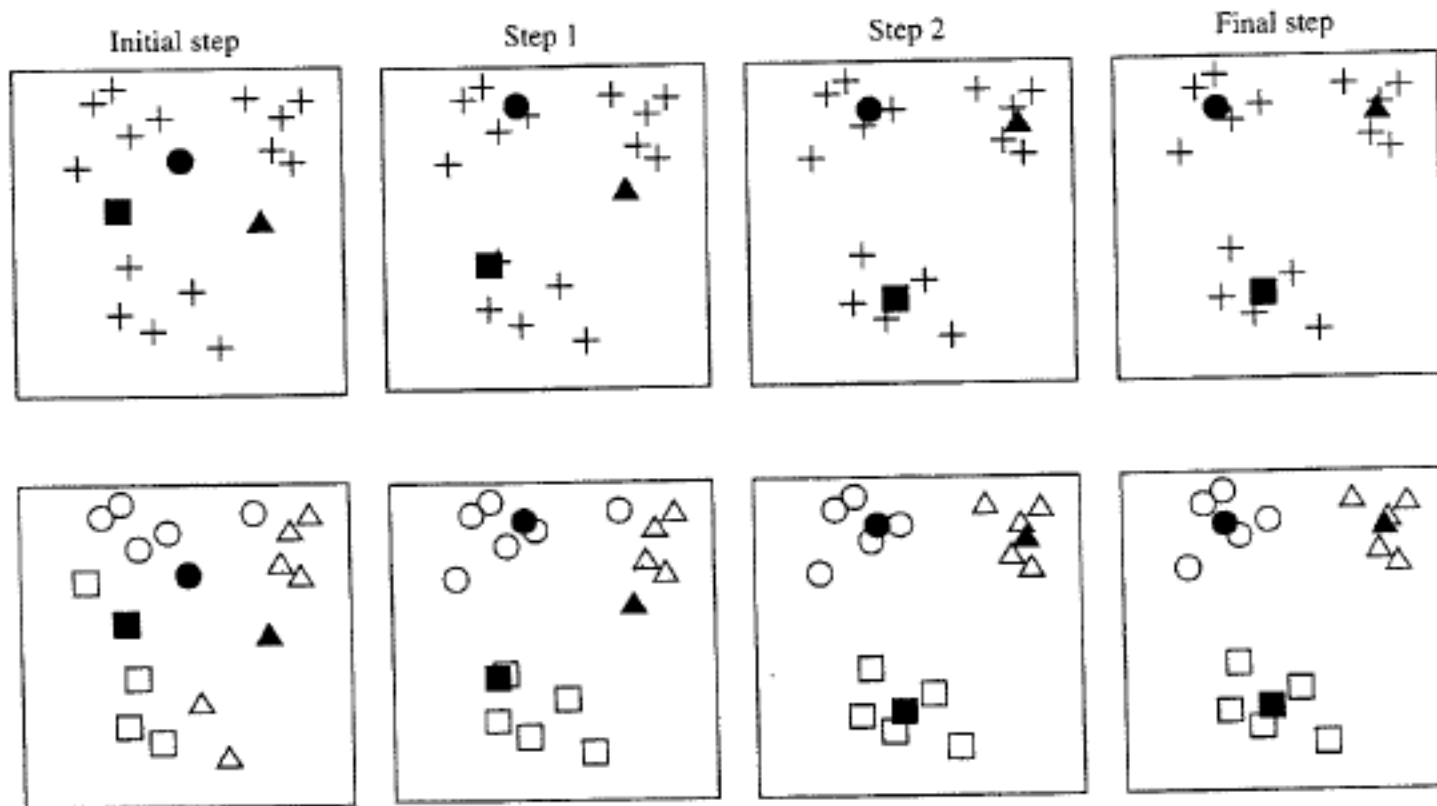
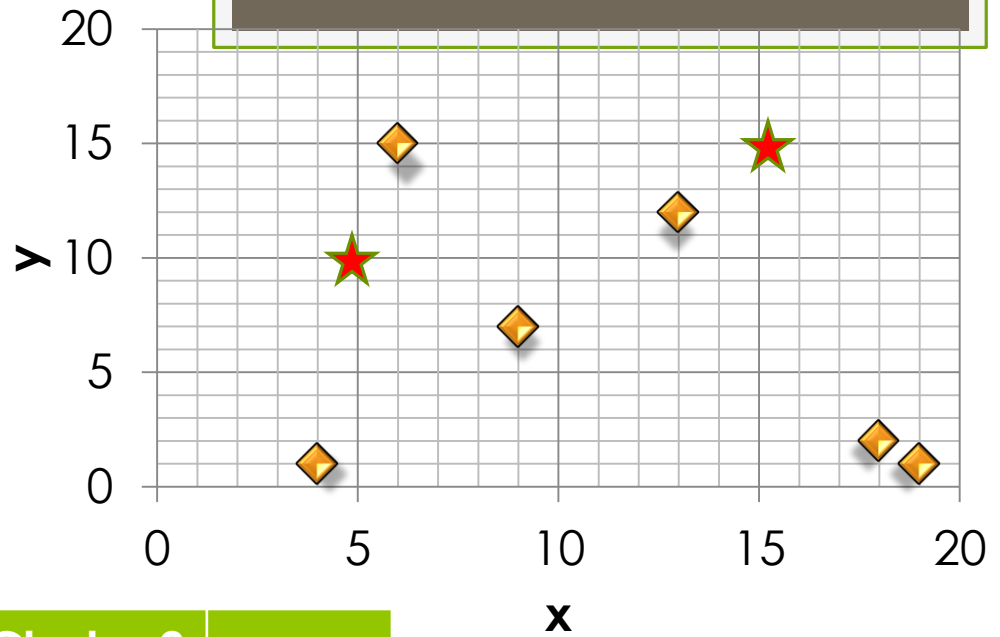


FIGURE 4.17

Iterative distance-based clustering.

EXAMPLE



Data		Cluster 1		Cluster 2	
X	Y	X=5	Y=10	X=15	Y=15
19	1				
13	12				
9	7				
6	15				
18	2				
4	1				

$$\sqrt{(a_1^{(1)} - a_1^{(2)})^2 + (a_2^{(1)} - a_2^{(2)})^2 + \dots + (a_k^{(1)} - a_k^{(2)})^2}$$

EXAMPLE

Data		Cluster 1		Cluster 2	
X	Y	X=5	Y=10	X=15	Y=15
19	1	16.64		14.56	
13	12	8.25		3.61	
9	7	5.00		10.00	
6	15	5.10		9.00	
18	2	15.26		13.34	
4	1	9.06		17.80	

$$d(1) = \sqrt{(19 - 5)^2 + (1 - 10)^2} = 16.64$$

$$d(1) = \sqrt{(19 - 15)^2 + (1 - 15)^2} = 14.56$$

EXAMPLE

Data		Cluster 1		Cluster 2	
X	Y	X=5	Y=10	X=15	Y=15
19	1	16.64		14.56	
13	12	8.25		3.61	
9	7	5.00		10.00	
6	15	5.10		9.00	
18	2	15.26		13.34	
4	1	9.06		17.80	

Now we assign each instance to the cluster which it's closest to (highlighted in the table.)

EXAMPLE

Data		Cluste r 1		Cluste r 2	
X	Y	X=5	Y=10	X=15	Y=15
19	1	16.64		14.56	
13	12	8.25		3.61	
9	7	5.00		10.00	
6	15	5.10		9.00	
18	2	15.26		13.34	
4	1	9.06		17.80	

Then we adjust the cluster centers to be the average of all of the instances

assigned to them. (This is called the centroid.)

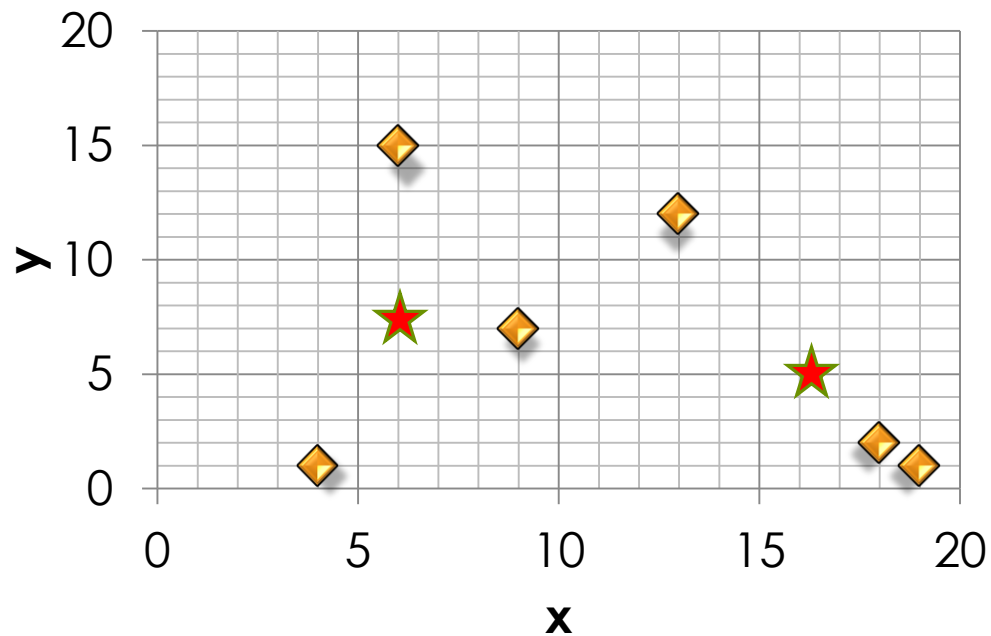
$$\text{Cluster Center 1, } X = (9+6+4)/3 = 6.33$$

$$Y = (7+15+1)/3 = 7.67$$

$$\text{Cluster Center 2, } X = (19+13+18)/3 = 16.67$$

$$Y = (1+12+2)/3 = 5$$

EXAMPLE



We place the new cluster centers and do the entire process again. We repeat this until no changes happen on an iteration.

Classical k-means clustering

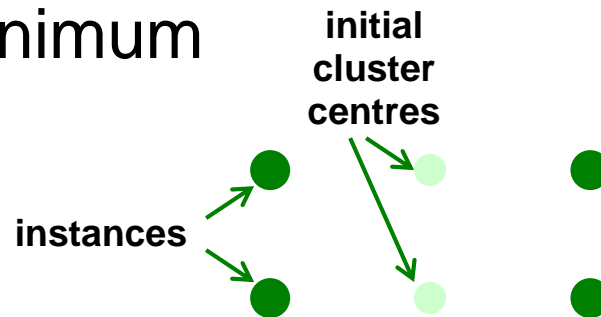
Algorithm minimizes squared distance to cluster centers

Result can vary significantly

based on initial choice of seeds

Can get trapped in local minimum

Example:



To increase chance of finding global optimum: restart with different random seeds

Can be applied recursively with $k = 2$