

1. (True/False) k-means always converges to a local optimum. 1 point
- ☒ True
☐ False
2. (True/False) The clustering objective is non-increasing throughout a run of k-means. 1 point
- ☒ True
☐ False
3. (True/False) Running k-means with a larger value of k always enables a lower possible final objective value than running k-means with smaller k. 1 point
- ☒ True
☐ False
4. (True/False) Any initialization of the centroids in k-means is just as good as any other. 1 point
- ☐ True
☒ False
5. (True/False) Initializing centroids using k-means++ guarantees convergence to a global optimum. 1 point
- ☐ True
☒ False
6. (True/False) Initializing centroids using k-means++ costs more than random initialization in the beginning, but can pay off eventually by speeding up convergence. 1 point
- ☒ True
☐ False

7. (True/False) Using k-means++ can only influence the number of iterations to convergence, not the quality of the final assignments (i.e., objective value at convergence). 1 point

☐ True
☒ False

8. Consider the following dataset: 4 points

	X1	X2
Data point 1	-1.88	2.05
Data point 2	-0.71	0.42
Data point 3	2.41	-0.67
Data point 4	1.85	-3.80
Data point 5	-3.69	-1.33

Perform k-means with $k=2$ until the cluster assignment does not change between successive iterations. Use the following initialization for the centroids:

	X1	X2
Cluster 1	2.00	2.00
Cluster 2	-2.00	-2.00

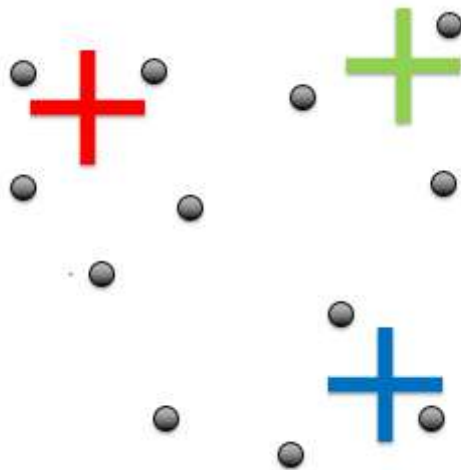
Which of the five data points changed its cluster assignment most often during the k-means run?

☐ Data point 1
☒ Data point 2
☐ Data point 3
☐ Data point 4
☐ Data point 5

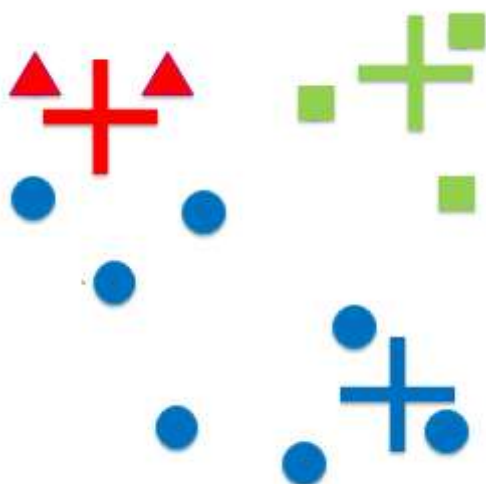
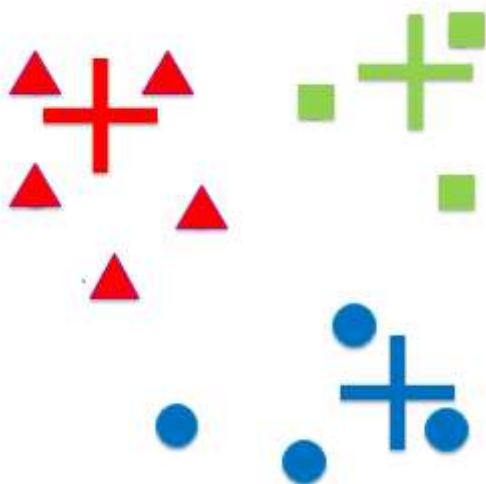
9.

1 point

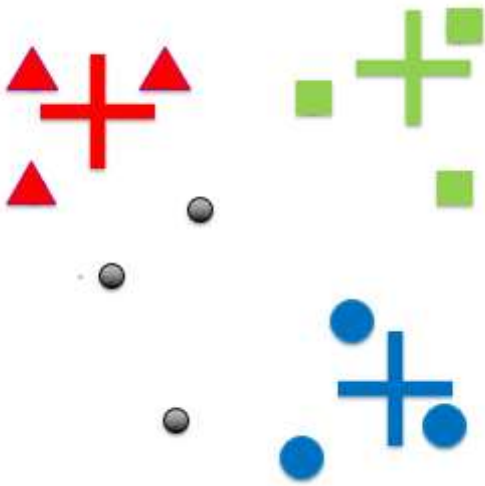
Suppose we initialize k-means with the following centroids



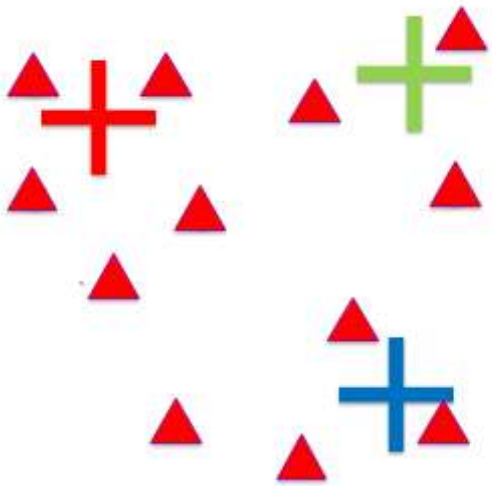
Which of the following best describes the cluster assignment in the first iteration of k-means?



○



○



○

