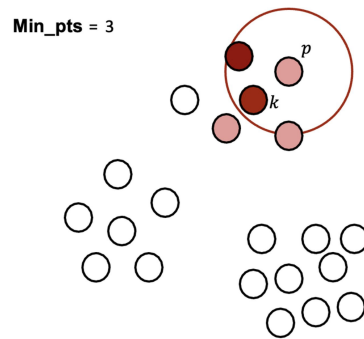


CS 506 Review Slides Questions

February 2026

Questions

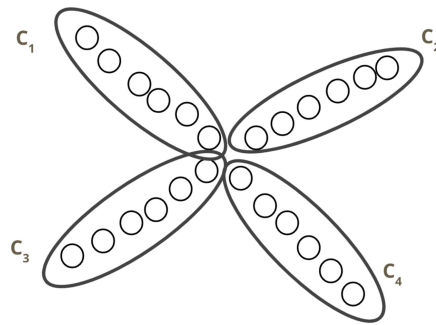
1. How will point p be assigned? (Note: we are assuming that for a point to be considered a part of p 's epsilon neighborhood, it must be entirely in the circle representing p 's neighborhood)



- (a) p is a core point, so we will iterate on its entire epsilon neighborhood
 - (b) p is a border point, so we will iterate on its entire epsilon neighborhood
 - (c) p is a border point, so we will assign it to k 's cluster and move onto the other points in k 's neighborhood
 - (d) p is a noise point, so we won't assign it to any cluster
2. Keeping epsilon the same and increasing min_pts in DBScan will likely increase the number of noise points
 - (a) True
 - (b) False
 3. Keeping epsilon the same and increasing min_pts in DBScan will likely increase the number of core points

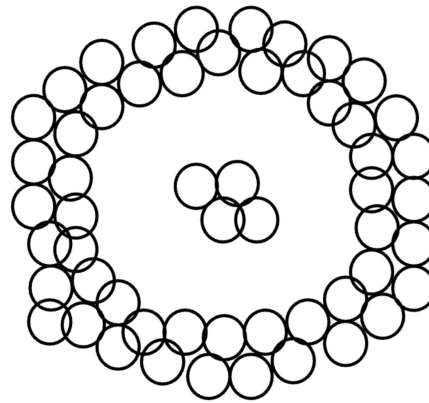
- (a) True
- (b) False

4. Is the following a possible output of DBScan?



- (a) Yes
- (b) No

5. Is DBScan well suited for the following dataset?

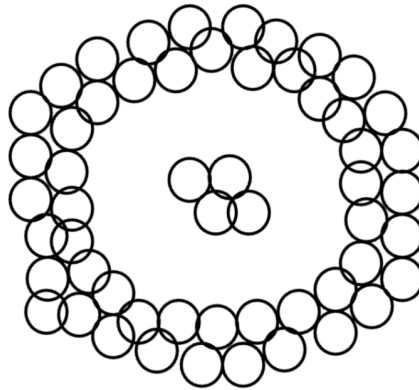


- (a) Yes
- (b) No

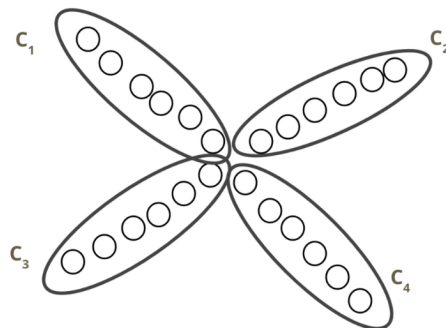
6. DBScan, like K-means, requires you to specify the number of clusters

- (a) Yes
- (b) No

7. Using cosine distance to calculate epsilon neighborhoods would likely enable us to detect the two natural clusters here.

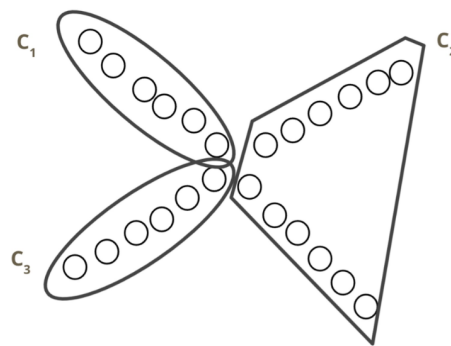


- (a) True
 - (b) False
8. One way to initialize GMM clustering is to use K-means since it also performs well when clusters are Gaussian
- (a) True
 - (b) False
9. In GMM, what is the likelihood of the data (with N data points) when $k=N$
- (a) ∞
 - (b) $-\infty$
 - (c) 1
 - (d) 0
10. Is the following a possible output from K-means++ clustering?



- (a) Yes
- (b) No

11. Is the following a possible output of K-means clustering?



- (a) Yes
- (b) No

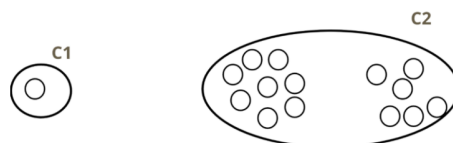
12. The difference between K-means and K-means++ is that K-means++ is faster, hence the ++

- (a) True
- (b) False

13. Consider the 1-dimensional dataset $\{4, 5, x\}$. With $k = 2$ and $x > 5$, what is the smallest value of x BEYOND which x will always be in its own cluster?

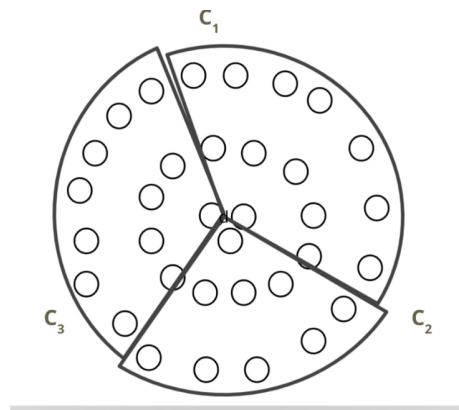
- (a) 7
- (b) 9
- (c) 6
- (d) 8

14. Is the following a possible output from k-means++ clustering?



- (a) Yes
- (b) No

15. Is the following a possible output of k-means clustering?



- (a) Yes
- (b) No

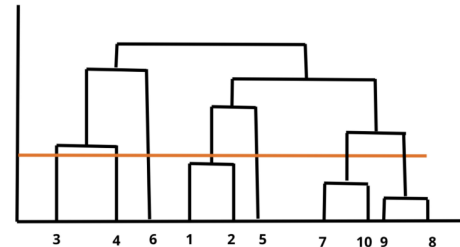
16. One way to divide clusters in Hierarchical Clustering's Divisive Algorithm (as opposed to agglomerative) is to use a partitional clustering method like K-means.

- (a) True
- (b) False

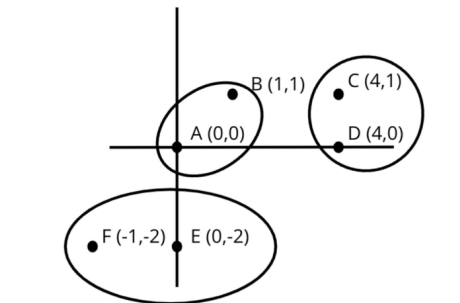
17. Using Euclidean and complete-link distance, what is the merging order for $A = (0, 0)$, $B = (1, 1)$, $C = (3, 0)$, $D = (0, -2)$?

- (a) $AB \rightarrow ABD \rightarrow ABCD$
- (b) $AB \rightarrow CD \rightarrow ABCD$
- (c) $AB \rightarrow ABC \rightarrow ABCD$

18. How many clusters would be created by cutting the dendrogram below as shown?



- (a) 5
 (b) 4
 (c) 6
 (d) 7
19. Which clusters get merged next if using Euclidean and complete-link distances?



- (a) AB and EF
 (b) AB and CD
 (c) CD and EF
20. The Manhattan distance is always less than or equal to the Euclidean distance between two points in a 2D space.
- (a) True
 (b) False

21. In n -dimensional space the Manhattan distance equals the Euclidean distance for
 - (a) Points where all the coordinates are the same except for one
 - (b) Points where all the coordinates are different except for one
 - (c) Points where all the coordinates are zero except for one
 - (d) all of the above
22. Consider arbitrary points A and B in \mathbb{R}^n s.t. $A \neq B$. For which distance function d , is $d(A, B)$ invariant to scaling? i.e., $c * d(A, B) = d(A, B)$ for any constant c
 - (a) Euclidean distance
 - (b) Chebyshev Distance
 - (c) Cosine Distance
 - (d) Manhattan Distance
23. What is the Jaccard distance between two completely different documents (i.e. no words in common at all)?
 - (a) -1
 - (b) 1
 - (c) 0
 - (d) Depends on the size of the documents
24. How many parameters does GMM need to estimate for k clusters and 1D data?
 - (a) $3k$
 - (b) k
 - (c) $4k$
 - (d) \sqrt{k}
25. Since outliers most likely don't belong to any cluster, the probabilities of belonging to each cluster, assigned by a GMM, will all be small
 - (a) True
 - (b) False
26. What type of distribution does each cluster follow in a GMM?
 - (a) Normal
 - (b) Geometric
 - (c) Hypergeometric
 - (d) Poisson

Solutions

1. (c) p is a border point, so we will assign it to k 's cluster and move onto the other points in k 's neighborhood
2. (a) True
3. (b) False
4. (b) No
5. (a) Yes
6. (b) No
7. (b) False
8. (a) True
9. (a) ∞
10. (a) Yes
11. (a) Yes
12. (b) False
13. (a) 7
14. (a) Yes
15. (a) Yes
16. (a) True
17. (c) $AB \rightarrow ABC \rightarrow ABCD$
18. (d) 7
19. (a) AB and EF
20. (b) False
21. (a) Points where all the coordinates are the same except for one
22. (c) Cosine Distance
23. (b) 1
24. (a)
25. False
26. (a) Normal