

| Timestamp | Email Address | Score | Using the given dataset, c | Using the given dataset, c | An engineering student w | An engineering student w | An engineering student w | An engineering student w | An engineering student w | An engineering student w | Consider a dataset in a h | In such a scenario, which Naive Bayes is called na | Which type of distance sh | In an advertising firm , wh | In a decision making proo | As you end binge watchin What are the Steps for K-Means Clustering. (1) Select k random points or centroids (if already) other web-series that migl (2) Calculate variance and repeat (3) assign each datapoint to their closest centroid | | | |
|--------------------|--------------------------|---------|----------------------------|----------------------------|--------------------------|--------------------------|--------------------------|--------------------------|--------------------------|--------------------------|--|--|---------------------------|-----------------------------|---------------------------|---|--|--|--|
| 4/21/2023 11:01:23 | vanah.kapila.ug20@nsut.i | 14 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Hamming Distance | k-Means Clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |
| 5/29/2023 11:08:30 | chanakya.ug20@nsut.ac.i | 12 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Euclidean Distance | hierarchical clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |
| 5/29/2023 11:09:04 | rahulkumar.ug20@nsut.ai | 11 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Hamming Distance | hierarchical clustering | 0.265 | Bayesian Classification | 1 -> 3 -> 2 | | | |
| 5/29/2023 11:09:49 | dipesh.ug20@nsut.ac.in | 12 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Euclidean Distance | hierarchical clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |
| 5/29/2023 11:15:23 | ayank.ug20@nsut.ac.in | 10 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc Naive Bayes assumes th | Euclidean Distance | hierarchical clustering | 0.265 | K-Means clustering | 1 -> 3 -> 2 | | | |
| 5/29/2023 11:24:14 | spansh.ug20@nsut.ac.in | 12 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Euclidean Distance | hierarchical clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |
| 5/29/2023 12:29:58 | ankush.goel.ug20@nsut.i | 12 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Euclidean Distance | hierarchical clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |
| 5/29/2023 12:31:35 | harsh.priye.ug20@nsut.ai | 12 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Euclidean Distance | hierarchical clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |
| 5/29/2023 12:33:28 | dharamveer.ug20@nsut.a | 4 / 14 | 0.037 | 0.037 | NO | 1 | 2 | | 1 | 2 | 3 Multinomial Naive Bayes Naive Bayes assumes th | Hamming Distance | k-Means Clustering | 0.47 | K-Nearest Neighbour | 3 -> 2 -> 1 | | | |
| 5/29/2023 13:00:34 | harsh.dabas.ug20@nsut.i | 3 / 14 | 0.037 | 0.049 | YES | 3 | 4 | | 2 | 3 | 3 Multinomial Naive Bayes By doing so, the joint distr | Euclidean Distance | hierarchical clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |
| 5/29/2023 13:00:46 | harsh.tomar.ug20@nsut.a | 11 / 14 | 0.069 | 0.037 | NO | 5 | 1 | | 3 | 4 | 2 Bernoulli Naive Bayes (Fc By doing so, the joint distr | Euclidean Distance | hierarchical clustering | 0.265 | K-Nearest Neighbour | 1 -> 3 -> 2 | | | |