

- => Agglomerative dustering Divisive dustering
 - 1 Bottom up approach (1) Top down approach
 - @ Combines the closest @ splits the most dissimilar cluster at each step cluster at each step
 - (3) Each data point is a (3) All data points are single cluster in a single cluster
 - (4) More commonly used (4) Used less often
 - E) Low computational (5) High computational cost
- 6 Easy to implement 6 Hard to implement
- 3 Sensitive to noisy A Sensitive to initial data and outliers split criteria
- (8) Stops when all data (8) stops when each data points are in one point is in its own cluster.
- (9) Builds the hierarchy (9) Builds the hierarchy from bottom to top from top to bottom (merging) (splitting)

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Date K-means clustering => O K- means clustering is an unsupervised machine learning algorithm used for dustoring tasks. 1) In k-means, unlabeled data is grouped into k-distinct clusters based on similarity. 6) Each cluster is defined by its centroid and the goal is to minimize the distance between data points and their respective cluster centraids. (4) Steps in K-means (i) Step !: Initialize K Choose the number of dusters (K) (ii) Step 2: Select initial centroids Randomly initialize K points as controid (iii) Step 3: Assign data points to nearest centroid For each date point, compute its distance from all K centeroids and assign it to the unster with nearest centroid. (i) Step 4: Recalculate centroids Compute the mean of all data points in each cluster and update the centroid positions (1) Step 5: Repeat 12 sound to rodown Reassign all data points to the nearest centroid and recalculate the centroids (vi) Step 6: Convergence check Stop ij: O No date points change chisters (2) The centroids do not move significantly. 3 The max no of iterations is reached Step 7: Output The algorithm outputs k clusters and their corresponding centroi'ds.

