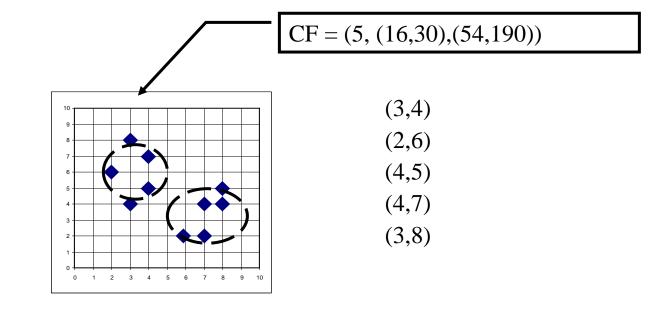


BIRCH (Balanced Iterative Reducing and Clustering Using Hierarchies)

- □ A multiphase clustering algorithm (Zhang, Ramakrishnan & Livny, SIGMOD'96)
- □ Incrementally construct a CF (Clustering Feature) tree, a hierarchical data structure for multiphase clustering
 - □ Phase 1: Scan DB to build an initial in-memory CF tree (a multi-level compression of the data that tries to preserve the inherent clustering structure of the data)
 - □ Phase 2: Use an arbitrary clustering algorithm to cluster the leaf nodes of the CFtree
- □ Key idea: Multi-level clustering
 - Low-level micro-clustering: Reduce complexity and increase scalability
 - ☐ High-level macro-clustering: Leave enough flexibility for high-level clustering
- □ Scales linearly: Find a good clustering with a single scan and improve the quality with a few additional scans

Clustering Feature Vector in BIRCH

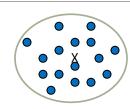
- □ Clustering Feature (CF): *CF* = (*N*, *LS*, *SS*)
 - N: Number of data points
 - LS: linear sum of N points: $\sum_{i=1}^{N} X_i$ SS: square sum of N points: $\sum_{i=1}^{N} X_i^2$



- Clustering feature:
 - □ Summary of the statistics for a given sub-cluster: the 0-th, 1st, and 2nd moments of the sub-cluster from the statistical point of view
 - Registers crucial measurements for computing cluster and utilizes storage efficiently

Measures of Cluster: Centroid, Radius and Diameter

- \Box Centroid: \vec{x}_0
 - the "middle" of a cluster
 - n: number of points in a cluster
 - $\overrightarrow{x_i}$ is the *i*-th point in the cluster
- □ Radius: R
 - Average distance from member objects to the centroid
 - ☐ The square root of average distance from any point of the cluster to its centroid
- Diameter: D
 - Average pairwise distance within a cluster
 - □ The square root of average mean squared distance between all pairs of points in the cluster



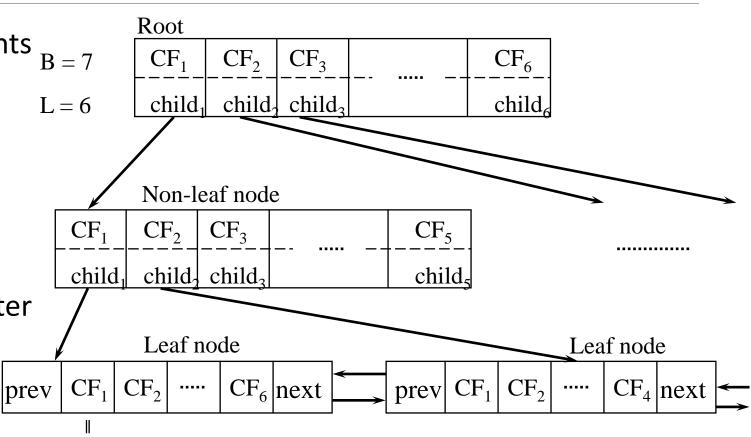
$$\vec{x}_0 = \frac{\sum_{i}^{n} \vec{x}_i}{n}$$

$$R = \sqrt{\frac{\sum_{i}^{n} (\vec{x}_i - \vec{x}_0)^2}{n}}$$

$$D = \sqrt{\frac{\sum_{i=1}^{n} \sum_{j=1}^{n} (\overrightarrow{x_i} - \overrightarrow{x_j})^2}{n(n-1)}}$$

The CF Tree Structure in BIRCH

- ☐ Incremental insertion of new points (similar to B+-tree)
- ☐ For each point in the input
 - □ Find closest leaf entry
 - Add point to leaf entry and update CF
 - If entry diameter > max_diameter
 - split leaf, and possibly parents
- ☐ A CF tree has two parameters
 - Branching factor: Maximum number of children
 - Maximum diameter of subclusters stored at the leaf nodes



- □ A CF tree: A height-balanced tree that stores the clustering features (CFs)
- ☐ The non-leaf nodes store sums of the CFs of their children

BIRCH: A Scalable and Flexible Clustering Method

- ☐ An integration of agglomerative clustering with other (flexible) clustering methods
 - Low-level micro-clustering
 - Exploring CP-feature and BIRCH tree structure
 - Preserving the inherent clustering structure of the data
 - Higher-level macro-clustering
 - □ Provide sufficient flexibility for integration with other clustering methods
- □ Impact to many other clustering methods and applications
- Concerns
 - Sensitive to insertion order of data points
 - Due to the fixed size of leaf nodes, clusters may not be so natural
 - Clusters tend to be spherical given the radius and diameter measures