Notes of Machine Learning Unsupervised Learning

Xiangli Chen

Computer Science Department University of Illinois at Chicago

July 16, 2017

Outline

Association Rule

2 Density Estimation

- Cluster
 - K-Means

Problem Formulation

K-means clustering problem: partition n observations into K sets $(K \le n)$ where $S = \{S_1, \ldots, S_k\}$ such that

$$\arg\min_{S} \sum_{k=1}^{K} \sum_{c_{i} \in S_{k}} ||x_{i} - \mu_{k}||_{2}^{2}$$

where μ_k is the mean of points in set S_k . Remark

- NP hard problem
 - Heurictic algorithm: K-means algorithm

K-means Algorithm

- Initialize K centers $\hat{\mu}^0 = (\hat{\mu}^0_1, \dots, \mu^0_K)$
- Do
 - Classify (fix μ , optimize C): at iteration t, assign each point $i \in \{1, ..., n\}$ nearest center

$$\forall i \quad \hat{C}^t(i) = \arg\min_{k} ||x_i - \hat{\mu}_k^t||_2^2$$

• Recenter (fix C, optimize μ)

$$\forall k \quad \hat{\mu}_k^{t+1} = \arg\min_{\mu} \sum_{i \in \{\hat{C}^t(i) = k\}} ||x_i - \mu||_2^2$$

Termination: none of n objects change membership

Complexity

Note that

$$\forall k \quad \frac{\partial \sum_{i \in \{\hat{C}^{t}(i) = k\}} ||x_{i} - \mu||_{2}^{2}}{\partial \mu} = 0 \Rightarrow$$

$$\hat{\mu}_{k}^{t+1} = \frac{\sum_{i \in \{\hat{C}^{t}(i) = k\}} x_{i}}{|\hat{C}^{t}(i) = k|} = \bar{x}_{i} \in \{\hat{C}^{t}(i) = k\}$$

Time complexity

- clarification O(Kn)
- recentering (compute the new mean) O(n)

Assume I iteractions, the time complexity is O(IKn).

Number of Clusters

The dissimlarity

$$W_k = \sum_{k=1}^K \sum_{i \in \{\hat{C}^t(i) = k\}} ||x_i - \hat{\mu}_k||_2^2$$

In general, $\{W_1, W_2, \ldots, W_{K_{\text{max}}}\}$ decreases with increasing K. Intuition: spliting a natural group reduces the criterion $W_k - W_{k+1}$ less than partitioning the union of two well-separated groups into their proper constituents.

Assume there are actually K^* distinct groupings, there will be a sharp decrease of $W_k - W_{k+1}$ at $k = k^*$, i.e.,

$$\{W_k - W_{k+1}|k < k^*\} >> \{W_k - W_{k+1}|k \ge k^*\}$$

Gap Statistics

To be continued

Seed Choice

Multiple random selections Forward stepsize assignment (choose datapoints to be centers)

- initialize x_{i1}
- do choose new center x_{i_m} for $i_m \in \{1, \dots, n\} \{i_1, \dots, i_{m-1}\}$ $\forall i \quad C(i) = \arg\min_m ||x_i x_{i_m}||_2^2$ $W_{i_m} = \sum_{k=1}^m \sum_{i \in \{C(i) = m\}} ||x_i x_{i_k}||_2^2$ $i_m = \arg\min_{i_m} W_{i_m}$
- *m* = *K*

Relation to Mixture Model and EM

To be continued

Hierarchical Clustering

To be continued