## **Quiz 7 Thursday - Rubrics**

- [1 point] What are the characteristics of points belonging to a cluster?
   Members of a cluster are close/similar to each other.
   Members of different clusters are far/dissimilar from each.
- 2. [3 points] Hierarchical Clustering: Imagine we are clustering the number of items bought on Amazon for a given user each day. We wish to perform a hierarchical clustering of the number of items bought among all users: 1, 3, 8, 16, and 30. Show what happens at each step until there are two clusters, and give these two clusters. Assume clusters are represented by their centroid (average), and at each step choose to merge two clusters whose resulting cluster has the smallest diameter.

Your answer should be in steps where each step shows the members of the new cluster formed, and its centroid. More specifically, if you are merging a cluster  $C1 = \{x, y, z\}$  of centroid c1 with a cluster  $C2 = \{p, q\}$  of centroid c2, you should report  $\{x, y, z, p, q\}$  in the table, as well as the centroid obtained with these 5 points).

Initial clusters : {1}, {3}, {8}, {16}, {30}

Centroids: 1, 3, 8, 16, 30

First step: {1, 3}, {8}, {16}, {30} [0.5 points]

Centroids: 2, 8, 16, 30 [0.5 points]

Second step: {1, 3, 8}, {16}, {30} [0.5 points]

Centroids: 4, 16, 30 [0.5 points]

Third step: {1, 3, 8}, {16, 30} [0.5 points]

Centroids: 4, 23 [0.5 points]

- 3. [1 point] What are the advantages of Item-based Collaborative filtering to User-based methods? Give with at least 2 advantages.
  - -Match user's rated items to similar items, Rather than matching similar users
  - -In practice, often leads to faster online systems and better recommendations
  - -Similarities between pairs of items i and j are computed off-line
  - -Predict rating of <u>user a on item i</u> with a simple weighted average.
  - Item based doesn't suffer from cold start problems as much as user based does.
  - User based is more prone to shilling attacks than Item based Collaborative filtering.

Rubrics: 0.5 points for each correct advantage.

- 4. [1 point] Select ALL of the statements that are TRUE about Inverse User Frequency.
  - A. Universally liked items are as useful in capturing similarity as less common items.
  - B. Inverse frequency  $f_j = log(n_j/n)$ ,  $n_j$  is number of users who have rated item j, n is total number of users.
  - C. If everyone has rated item j, then fj is zero.

D. When transform ratings by multiplying the original rating by fj, less popular items will have greater effect on prediction

Ans.) C, D

- 5. [1 point] For 2 points A(x1,y1) and B(x2,y2) and the distance between them being **dist** =  $((x1-x2)^2 + (y1-y2)^2)^0.5$ . **dist** could be which of the following distance measures?
  - A. Manhattan Distance
  - B. Euclidean Distance
  - C. Minkowski Distance
  - D. Edit Distance

Answer. B,C

- 6. [2 points]What are the Characteristics and Challenges of Collaborative Filtering with Cold start problems? Name two approaches on how to deal with this problem?
  •Cold start problem
  - •When a new user or item has just entered the system
  - •Hard to find similarities: not enough information to make good recommendations
  - •New item problem: can't be recommended until some users rate it
    - •Also applies to obscure items
    - Also called "first-rater problem"
  - •New users:

not given good recommendations because of lack of rating or purchase history •Approaches:

- Content-based systems do not rely on ratings from other users
- •Hybrid CF (content-boosted CF): external content information can be used to produce predictions for new users or new items
- •Research on effectively selecting items to be rated by a user to rapidly improve recommendation performance

Rubrics: 1 point for challenges and 0.5 points for each approaches

- 7. [1 point] Which of the following Hybrid Recommendation Systems involves the concept of breaking ties using one of its components?
  - A. Feature Augmentation
  - B. Cascade Hybrid
  - C. Meta-Level Hybrid
  - D. Switching Hybrid

Answer, B