

**CSE 5800 Mining/Learning and the  
Internet—HW3**  
**Due Oct 21, Wed, 6:30pm**  
**Submit Server: course= ml-internet ,  
project=hw3**

1. Implement these clustering algorithms, each outputs  $K$  clusters:
  - (a) K-means
  - (b) Bisecting K-means with largest cluster to split
  - (c) Bisecting K-means with least overall similarity to split
  - (d) Agglomerative Hierarchical Clustering with Intra-Cluster Similarity technique (IST)
  - (e) Agglomerative Hierarchical Clustering with Centroid Similarity technique (CST)
  - (f) Agglomerative Hierarchical Clustering with UPGMA
  - (g) Agglomerative Hierarchical Clustering with UPGMA to seed K-means
2. Each document is represented by a TF-IDF unit vector, each component is:  $tf_i \times idf_i$ , where:
  - $tf_i$  is the frequency of term  $i$  in the document divided by the total number of terms in the document and
  - $idf_i = \log(D/df_i)$ , where  $df_i$  is the number of documents that contain term  $i$  and  $D$  is the total number of documents
  - to get a unit vector, divide each component by the magnitude of the vector
3. Allow these parameters:
  - (a) number of (final) clusters ( $K$  in the paper)
  - (b) number of iterations ( $ITER$  in the paper) for Bisecting K-means
4. Measure performance of final clusters using:
  - (a) Entropy
  - (b) F-measure
  - (c) Overall Similarity
  - (d) Silhouette Coefficient
5. Three data sets:
  - (a) toy data set on the course web site
  - (b) news data set on the course web site
  - (c) your own data set
6. A report (in pdf) that discusses the following:
  - (a) Sensitivity analysis of parameters: for the second data set,
    - i. for bisecting k-means, vary  $ITER$  from 2 to 10 with increment of 2 (with  $K=6$ )
    - ii. for each algorithm, vary  $K$  (keep  $ITER$  constant for bisecting k-means based on the previous experiment) [assuming the desired number of clusters is not known]
    - iii. calculate each performance measurement,
    - iv. plot performance vs. value of a parameter,
    - v. discuss the value for each parameter that seems to achieve the highest performance and possible reasons.
    - vi. discuss if any of the performance measurements can help determine the value of  $K$  (which is usually not known in advance)
  - (b) Compare the clustering algorithms: for the second data set,
    - i. use the plot(s) for performance vs. number of clusters for different algorithms
    - ii. discuss the relative performance of different algorithms
7. Implementation:
  - (a) preferably use one of these programming languages: C, C++, Java, Python, or LISP.
  - (b) input files: a file for the topic names; each topic has a file, which has multiple documents, each document starts with --DocID--
  - (c) three modules:
    - i. Preprocess: input the documents, output TF-IDF vectors
    - ii. Cluster: input the TF-IDF vectors; for each cluster, output DocID's in the cluster and the top 3 words in the centroid
    - iii. Evaluate: input DocID's, their class labels and cluster membership; output performance
8. Submission:
  - (a) source code
  - (b) your data set
  - (c) report in pdf
  - (d) README.txt (how to compile and run your program/experiments on code.fit.edu or hopper.cs.fit.edu)