# CS412 "An Introduction to Data Mining" (Fall 2014) Final Exam

(Friday, Dec. 12, 2014, 180 minutes, 150 marks, two sheets of references, brief answers)

| Name: |       | NetID: |       |       |       | Score:      |
|-------|-------|--------|-------|-------|-------|-------------|
| 1[30] | 2[10] | 3 [20] | 4[30] | 5[40] | 6[20] | Total [150] |

#### 1. [30] Short Questions

(a) [3] Name one visualization technique that may help to decide if using PCA on a particular 2D dataset is beneficial or not. Explain how the technique is helpful for the purpose.

(b) [3] Give an example when using Manhattan distance is more suitable than using Euclidean distance.

(c) [4] Suppose Dunkin' Donuts wants to study the correlation between the unit price  $x_{i,t}$  and net profit  $y_{i,t}$  where i denotes the i<sup>th</sup> store, while t denotes the date. A linear regression model is used that  $y = \alpha x + \beta$ . (A) Judge which category  $\alpha$  and  $\beta$  belong to (distributive, algebraic, or holistic), and (B) explain how each can be computed efficiently in multidimensional space.

**Hint:**  $\alpha$  and  $\beta$  can be calculated as follows

$$\alpha = \frac{NT \sum_{t=1}^{T} \sum_{i=1}^{N} x_{i,t} y_{i,t} - \sum_{t=1}^{T} \sum_{i=1}^{N} x_{i,t} \sum_{t=1}^{T} \sum_{i=1}^{N} y_{i,t}}{NT \sum_{t=1}^{T} \sum_{i=1}^{N} x_{i,t}^{2} - (\sum_{t=1}^{T} \sum_{i=1}^{N} x_{i,t})^{2}}$$

$$\beta = \frac{1}{NT} \left( \sum_{t=1}^{T} \sum_{i=1}^{N} y_{i,t} - \alpha \sum_{t=1}^{T} \sum_{i=1}^{N} x_{i,t} \right)$$

- (d) [4] Suppose a base cuboid has D dimensions and contains m (m > 0) nonempty cells. Each dimension i has  $H_i$  levels (not including all), answer the following questions.
  - i. [2] How many aggregate cuboids does this cube contain (not including the base cuboid)?

ii. [2] What is the maximum number of nonempty cells possible in such a materialized cube?

| (e) | [4] Give one example of classification problem where false negative rate is more important than false positive rate, and explain why.   |
|-----|---|
| (f) | [4] Give one real-world application when semi-supervised learning might be a good solution.   |
| (g) | [4] Explain if the following claim is true or not: the result of DBSCAN with $MinPts=2$ will be the same as of hierarchical agglomerative clustering (AGNES) with the single link metric with the dendrogram cut at height $\epsilon$ . |
| (h) | [4] Among DBSCAN ( $MinPts > 2$ ), K-Means, and AGNES, which one is deterministic, i.e. the output does not depend on random numbers used in the algorithm. Explain why.  |
|     |   |

## 2. [10] Data Preprocessing.

Consider 10 data points in 2-D space, as specified in Table 1.

| X1    | -2 | -1 | 0  | 1  | 2  | -2  | -1  | 0   | 1   | 2   |
|-------|----|----|----|----|----|-----|-----|-----|-----|-----|
| X2    | -1 | -1 | -1 | -1 | -1 | 1   | 1   | 1   | 1   | 1   |
| Class | no | no | no | no | no | yes | yes | yes | yes | yes |

Table 1: Data points in 2-D space.

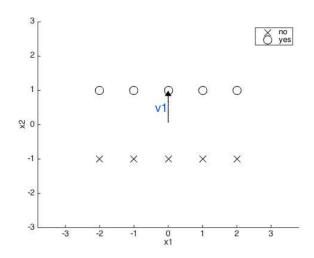


Figure 1: Visualization of data points and principal components

(a) [3] One of the principal component is  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ , which is visualized and labeled as v1 on Figure 1. Please write down the vector representation of the remaining principal component, as well as visualize and label it as v2 on Figure 1.

| (b) | [3] Which one should be the first principal component, i.e., the most important one? explain.  | Briefly |
|-----|--|---------|
| (c) | [2] Use formula to find the projected point for [-2,1] on the given principal component.   | Plassa  |
| (0) | visualize it on Figure 1.  | 1 lease |
| (d) | [2] If we want to use the dataset to train a model to predict the class label, does it makes sense to project data points on the first or the second component? Briefly explain. | æ more  |

- 3. [20] Data Warehousing, OLAP and Data Cube Computation
  - (a) [8] Suppose an online social website consists of five dimensions: user, location, posts information, friends, and education, and one measure: overall post count. Education refers to the schools the user has attended by 2014. Overall post count measures how active the user is.
    - i. [4] Draw a **snowflake** schema diagram (sketch it, do not have to mark every possible level, and make your implicit assumptions on the levels of a dimension when you draw it).

ii. [4] If one would like to start at the Apex cuboid and find top 3 active users in each university within Illinois state at year 2014, what are the specific OLAP operations (e.g., roll-up on which dimension from which level to which level) that one should perform based on your design?

(b) [6] Suppose the base cuboid of a data cube contains only four cells

$$(a_1, a_2, a_3, a_4, a_5, \dots, a_{2k})$$

$$(b_1, a_2, b_3, a_4, b_5, \dots, a_{2k})$$

$$(c_1, a_2, c_3, a_4, c_5, \dots, a_{2k})$$

$$(d_1, d_2, d_3, d_4, d_5, \dots, d_{2k})$$

where  $k \in \mathbb{N}_+$  (k is positive),  $a_i \neq b_i \neq c_i \neq d_i$ ,  $\forall i = 1, \dots 2k$ .

- i. [3] If we set minimum support = 2, how many nonempty aggregate cells are there in the corresponding *iceberg cube*?
- ii. [3] How many closed cells are there in the *iceberg cube*?

Note: Please show essential calculation steps.

- (c) [3] For the following tasks, which cube implementation method is better? multiway array cubing, BUC (bottom-up computation), or **neither**? Briefly explain.
  - i. Considering a data cube about sales in a small store, with five dimensions (customer, time, product, unit, price), fully materialize this data cube.
  - ii. Computing a large iceberg cube of around 830 dimensions.

(d) [3] In the study of TV ratings, as it is impossible to gather the opinion of everyone in the population, the ratings analysis relies on a sample of the data for analysis. The sampling cube is a data cube structure that stores the sample data and their multidimensional aggregates. In the context of sampling cube, what does "confidence interval" mean? What can we do to boost the reliability of query answers?

**Hint:** The confidence interval determines the reliability of query answers. Consider the factors that influence the confidence interval.

## 4. [30] Mining Frequent Patterns

(a) [14] Given a database of 4 transactions, you are to use MaxMiner to mine the frequent max patterns. Here  $min\_sup = 2$ .

| customer_id | shopping items |
|-------------|----------------|
| 1           | abcde          |
| 2           | bde            |
| 3           | aef            |
| 4           | bcde           |

Table 2: Transaction Database to Mine Max Patterns

i. [4] Show the nodes you generate from the root node (abcdef) of your set-enumeration tree.

ii. [4] When do you apply global pruning principle? Show an example from the given database. You need to show what are pruned.

iii. [4] When do you apply local pruning principle? Show an example from the given database. You need to show what are pruned.

iv. [2] Show the max patterns in the given database.

(b) [8] Suppose a sequential database D contains three sequences as follows. Note (bc) means that items b and c are purchased at the same time (i.e., in the same transaction). Let the minimum support be 3. You are going to use PrefixSpan to mine the frequent sequential patterns.

| customer_id | shopping sequence |
|-------------|-------------------|
| 1           | a(bc)(de)f        |
| 2           | bc(ad)ef          |
| 3           | a(bc)d(ab)ef      |

Table 3: Transaction Database to Mine Sequential Patterns

i. [4] Show  $\langle b \rangle$ -projected database.

ii. [4] What frequent patterns will you get from  $\langle b \rangle$ -projected database?

- (c) [8] The price of each item in a store is nonnegative. For each of the following cases, identify the type of constraint they represent and briefly discuss how to mine such association rules efficiently with frequent pattern mining algorithms.
  - i. [4] Containing at least one Xbox game.

ii. [4] Containing items the sum of whose prices is less than \$50.

#### 5. [40] Classification

(a) [7] Suppose we want to build a **Naive Bayes** classifier to filter spam messages. We collect five messages with labels as in **Table 4**. As the first step, you need to do pre-processing on the raw data. Please turn the data into a format upon which you can build a Naive Bayes classifier. Show the pre-processed results. (Hint: Generate a feature vector for each message).

| ID | Message            | Label |
|----|--------------------|-------|
| 1  | save money coupon  | Yes   |
| 2  | coupon money visit | Yes   |
| 3  | pay you money      | No    |
| 4  | save you           | No    |
| 5  | pay you visit      | No    |

Table 4: Training data for Question 5a and 5b

(b) [7] Based on the pre-processed data in **Question 5a**, we want to construct a Naive Bayes classifier. If no smoothing is applied, we'll have a problem in classifying the short message "save you coupon". State the problem via calculation, and outline a solution to the problem.

(c) [6] Suppose we have built a classifier to filter spam messages, and evaluated its performance. We summarize the evaluation results in the form of confusion matrix as in **Table 5**.

| A\P   | yes | no  | Total |
|-------|-----|-----|-------|
| yes   | 180 | 20  | 200   |
| no    | 320 | 480 | 800   |
| Total | 500 | 500 | 1,000 |

Table 5: A confusion matrix for the classes spam = yes and spam = no. 'A' represents "Actual class", and 'P' represents "Predicted class".

Please answer the following questions:

- i. Compute the precision and recall of the classifier.
- ii. Based on the calculation, is it a useful classifier? Explain why or why not for this particular task.

(d) [14] Suppose we have four training points, which are listed in **Table 6**. Each point has two attributes, i.e.,  $x_1$  and  $x_2$ , and a label y. You'll apply Adaboost Algorithm in this question. Suppose, initially, the weight for each point is uniform, i.e.,  $w_1(j) = \frac{1}{4}, j = 1, ..., 4$ . In each round, we sample with replacement according to the weights, and get a training datset  $D_i$ , i = 1, 2, 3. Suppose, based on  $D_i$ , we learn a classifier  $M_i$ , i = 1, 2, 3, which has the following rule:

$$M_1: \ \hat{y} = \begin{cases} +1 & \text{if } x_1 \le -0.5 \\ -1 & \text{if } x_1 > -0.5 \end{cases} \qquad M_2: \ \hat{y} = \begin{cases} +1 & \text{if } x_2 \le 0.5 \\ -1 & \text{if } x_2 > 0.5 \end{cases} \qquad M_3: \ \hat{y} = \begin{cases} +1 & \text{if } x_1 \ge 0.5 \\ -1 & \text{if } x_1 < 0.5 \end{cases}$$

| $x_1$ | $x_2$ | y  |
|-------|-------|----|
| 1     | 0     | +1 |
| -1    | 0     | +1 |
| 0     | 1     | -1 |
| 0     | -1    | -1 |

Table 6: Training data for Question 5d and 5e

#### Please answer:

- i. Compute the (weighted) error  $\epsilon_i$  of model  $M_i$  (i = 1, 2, 3), and the weights for each point after the first and second round, i.e.,  $w_2$  and  $w_3$ .
- ii. Combine the three classifiers  $M_1$  to  $M_3$  based on your calculation in the above step. What's the error of this combined classifier on the training dataset?

- (e) [6] Perceptron algorithm iteratively updates the weights of a linear classifier until certain condition is satisfied. Please answer:
  - i. State this stopping condition.
  - ii. For the training data in **Table 6**, will the Perceptron algorithm terminate? Why or why not? (Hint: plot the points)

## 6. [20] Clustering

Suppose we have 11 data points, which are listed and plotted in Figure 2. The ground truth (the correct cluster output) is also provided.

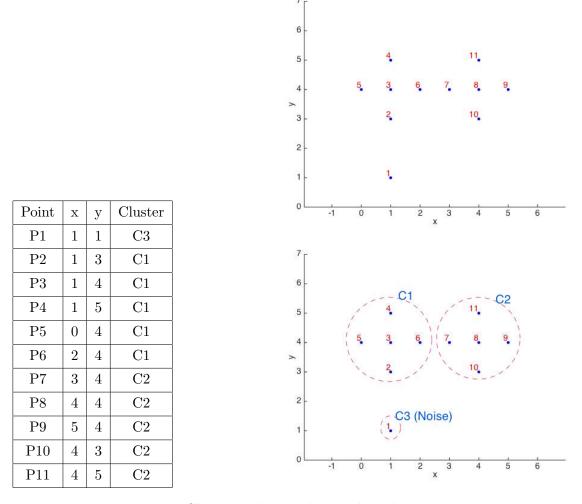


Figure 2: Clustering data and ground truth

| (a) | [3] If we perform AGNES, a hierarchical clustering algorithm on the points above, using single link method and adopting Euclidean distance as the dissimilarity measure, how many levels will the output diagram have? What are the corresponding values? Hint: each level of a dendrogram corresponds to a dissimilarity value, and you do not have to show how you perform AGNES. |
|-----|---|
|     |   |
| (b) | [2] If we don't know the ground truth, and want to cluster the data set into 2 groups, based on the result above, what are the members of the 2 groups?   |
|     |   |
| (c) | [4] Based on the given ground truth, what are the B-Cubed precision and recall of the output? Show your computation.  |
|     |   |

(d) [3] Using the same data as in Figure 2, show the clustering result of K-Means by annotating Figure 3, with K=2, initial means: P3 and P8, and Euclidean distance as the distance function. Note: You do not have to show your computation or to illustrate the means.

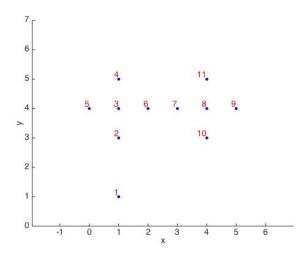


Figure 3: Output of K-Means

(e) [5] Using the same data as in Figure 2, perform DBSCAN, a density-based algorithm, with MinPts = 4,  $\epsilon = 1.1$ , and random points: P1, then P5, then P3, then P8. You need to list all the steps, and the final clusters.

(f) [3] Which clustering result is better, the one from K-Means or the one from DBSCAN? Which characteristics of the dataset cause that difference?