

[This question paper contains 4 printed pages.]

Your Roll No.....

Sr. No. of Question Paper : 13135

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Unique Paper Code : 32347611

Name of the Paper : Data Mining (CBCS)

Name of the Course : B.Sc. (Hons.) Computer Science

Semester : VI

Duration : 3 Hours

Maximum Marks : 75

**Instructions for Candidates**

1. Write your Roll No. on the top immediately on receipt of this question paper.
2. Question No. 1 (Section A) is compulsory.
3. Attempt any 4 Questions from Nos. 2 to 8 (Section B).
4. Parts of a question must be answered together.

**Section A**

1. (a) Determine the attribute type for each of the following : (2)
  - (i) Movie ratings on a scale of 1 to 5 stars
  - (ii) Temperature measured in degrees Celsius
  - (iii) Postal/ZIP codes
  - (iv) Number of children in a family.
- (b) Find the Euclidean distance between the data points  $X(1, 2, 0, -1)$  and  $Y(0, -1, 2, 1)$ . (2)
- (c) If the recall and precision of a classifier are 0.6 and 0.8 respectively, compute the value of  $F_1$  measure. (2)
- (d) State the *Apriori principle*. If itemset  $\{p, q\}$  is found to be infrequent, what can be said about itemset  $\{p, q, r\}$  ? Explain. (2)
- (e) Briefly describe *two* methods used for handling missing values in a dataset. (3)
- (f) How is an eager learner classifier different from a lazy learner classifier ? Support your answer with one example of each. (3)

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- (g) Consider a categorical attribute Quality with values {very poor, poor, average, good, excellent}. Show the representation of each value when this attribute is converted to binary form using : (3)
- (i) 3 bits
  - (ii) 5 asymmetric binary attributes.
- (h) Explain the following terms with reference to the DBSCAN clustering algorithm : (2)
- (i) Core point
  - (ii) Border point
- (i) Normalize the following set of values using min-max normalization : {15, 19, 22, 28}. (3)
- (j) The confusion matrix for a 2-class problem is given below. Calculate the *Accuracy*, *Sensitivity*, *Specificity* and *Error rate* : (5)

|              |           | Predicted Class |           |
|--------------|-----------|-----------------|-----------|
|              |           | Class = 1       | Class = 0 |
| Actual Class | Class = 1 | 350             | 150       |
|              | Class = 0 | 100             | 400       |

- (k) What is the difference between noise and outliers ? Are noise objects always outliers? Are outliers always noise objects ? Justify with examples. (4)
- (l) What is k-fold cross-validation ? How does it differ from the holdout method ? List two advantages of k-fold cross-validation over the holdout method. (4)

### Section B

2. Consider the following dataset for a 2-class classification problem : (10)

| A | B | Class |
|---|---|-------|
| T | F | +     |
| T | T | +     |
| T | T | -     |
| T | F | -     |
| T | T | +     |
| F | F | -     |
| F | F | -     |
| F | T | +     |
| F | F | -     |
| T | T | +     |

- (i) Calculate the Gini index when splitting on attributes A and B.
- (ii) Which attribute would the decision tree induction algorithm choose to split on?
- (iii) Draw the resulting decision tree showing the number of instances of each class at the leaf nodes.
- (iv) How many instances are misclassified by the resulting decision tree?

3. Consider the following transaction dataset. Assume the minimum support is 40% and the minimum confidence is 60%. (10)

| TID | Items        |
|-----|--------------|
| T1  | {a, b, c}    |
| T2  | {a, b, d}    |
| T3  | {b, c, d}    |
| T4  | {a, c, d}    |
| T5  | {a, b, c, d} |
| T6  | {b, c}       |
| T7  | {a, b, d}    |
| T8  | {b, c, d}    |
| T9  | {a, b, c, d} |
| T10 | {a, c}       |

- (i) Show clearly the steps to find frequent itemsets of all sizes using the Apriori algorithm.
- (ii) Generate all the strong association rules from the frequent 3-itemsets.
4. Given the following data points : 3, 6, 8, 14, 16, 22, 28, 32. Assume  $k = 2$ , and the initial cluster centroids  $\mu_1 = 3$  and  $\mu_2 = 14$ . Show clearly the clusters obtained and the new cluster centres after each iteration of the K-means algorithm for two iterations. (10)
5. The DM Cafe sells cookies with optional toppings : chocolate-chips, walnuts, and oats. Suppose, you have tried five cookies (C1 to C5) and kept a record of which ones you liked : (10)

|    | Chocolate-chips | Walnuts | Oats  | Liked |
|----|-----------------|---------|-------|-------|
| C1 | True            | True    | False | True  |
| C2 | False           | True    | True  | False |
| C3 | True            | False   | True  | True  |
| C4 | True            | True    | True  | False |
| C5 | False           | False   | True  | True  |

Show binarization of the above data and use it to calculate Euclidean distances, to demonstrate how the k-Nearest-Neighbor (k-NN) classifier with majority voting would classify a tuple <True, False, True>, for  $k = 1$  and  $k = 3$  respectively.

6. Consider the dataset shown below :

(10)

| Outlook  | Temperature | Humidity | Windy | Play Golf |
|----------|-------------|----------|-------|-----------|
| Rainy    | Hot         | High     | False | No        |
| Rainy    | Hot         | High     | True  | No        |
| Overcast | Hot         | High     | False | Yes       |
| Sunny    | Mild        | High     | False | Yes       |
| Sunny    | Cool        | Normal   | False | Yes       |
| Sunny    | Cool        | Normal   | True  | No        |
| Overcast | Cool        | Normal   | True  | Yes       |
| Rainy    | Mild        | High     | False | No        |
| Rainy    | Cool        | Normal   | False | Yes       |
| Sunny    | Mild        | Normal   | False | Yes       |
| Rainy    | Mild        | Normal   | True  | Yes       |
| Overcast | Mild        | High     | True  | Yes       |
| Overcast | Hot         | Normal   | False | Yes       |
| Sunny    | Mild        | High     | True  | No        |

- (i) Estimate the conditional probabilities  $P(\text{Outlook}|\text{Yes})$ ,  $P(\text{Temperature}|\text{Yes})$ ,  $P(\text{Humidity}|\text{Yes})$ ,  $P(\text{Windy}|\text{Yes})$ ,  $P(\text{Outlook}|\text{No})$ ,  $P(\text{Temperature}|\text{No})$ ,  $P(\text{Humidity}|\text{No})$ , and  $P(\text{Windy}|\text{No})$ .
- (ii) Use these estimates of conditional probabilities to predict the class label (Play Golf) for a test sample (Outlook = Rainy, Temperature = Cool, Humidity = High, Windy = True) using the naive Bayes approach.

7. Use the distance matrix given below to perform hierarchical clustering using the *single link* approach. Show the dendrogram clearly indicating the merging height at each step.

(10)

|    | P1   | P2   | P3   | P4   | P5 |
|----|------|------|------|------|----|
| P1 | 0    |      |      |      |    |
| P2 | 0.30 | 0    |      |      |    |
| P3 | 0.25 | 0.18 | 0    |      |    |
| P4 | 0.40 | 0.32 | 0.20 | 0    |    |
| P5 | 0.35 | 0.15 | 0.30 | 0.28 | 0  |

8. Consider the following rule set for animal classification :

(10)

R1 : (Habitat = land)  $\wedge$  (Has\_Fur = yes)  $\rightarrow$  Mammals  
 R2 : (Has\_Wings = yes)  $\wedge$  (Has\_Fur = no)  $\rightarrow$  Birds  
 R3 : (Has\_Scales = yes)  $\wedge$  (Habitat = water)  $\rightarrow$  Fish  
 R4 : (Habitat = land)  $\wedge$  (Has\_Fur = no)  $\wedge$  (Has\_Wings = no)  $\rightarrow$  Reptiles  
 R5 : (Habitat = water)  $\wedge$  (Has\_Scales = no)  $\rightarrow$  Amphibians

Consider the following test instances :

| Name  | Habitat | Has_Fur | Has_Wings | Has_Scales | Class |
|-------|---------|---------|-----------|------------|-------|
| eagle | land    | no      | yes       | no         | ?     |
| dog   | land    | yes     | no        | no         | ?     |
| trout | water   | no      | no        | yes        | ?     |
| frog  | water   | no      | no        | no         | ?     |

- (i) Are these rules mutually exclusive ? Justify your answer.  
 (ii) Are these rules exhaustive ? Justify with an example if not.  
 (iii) Identify which rule(s) cover each test instance and predict the class for each.

(200)