### Dept. of Computer Science & Engineering Indian Institute of Technology Kharagpur

 Roll no.:
 Name:

# CS43002 Database Management Systems Class Test 16.03.2018 Duration: 30 Minutes Full Marks: 20

Answer all questions. Write your answers in the box provided.

- 1. a-j are multiple choice type questions. For each question, zero or more options may be correct. 2 marks will be awarded for correct answer, 1 mark will be deducted for wrong answer. A question will be considered to be correctly answered if all the correct and only the correct options are chosen. If you think no option is correct for a question, write NONE. If you leave any answer blank, 5 marks will be deducted.
- a. In a star schema, usually:
  - (i) the fact table is normalized
  - (ii) the dimension tables are normalized
  - (iii) the fact table is de-normalized
  - (iv) the dimension tables are de-normalized
- b. In OLAP operations, roll-up can be achieved by:
  - (i) moving up a dimension hierarchy like month  $\rightarrow$  quarter, etc.
  - (ii) moving down a dimension hierarchy like quarter  $\rightarrow$  month, etc.
  - (iii) adding a new dimension
  - (iv) removing one or more dimensions
- c. It is beneficial and practical to materialize all the views in a data cube when
  - (i) the number of levels in dimensional hierarchies are very large and there are too many dimensions
  - (ii) the speed of retrieval is the primary objective
  - (iii) disk space saving is the primary objective
  - (iv) we can implement a greedy algorithm for selecting the views to be materialized
- d. Data warehouses
  - (i) are built to support long term decision making of top-level managers
  - (ii) are built to support short term decision making of junior managers
  - (iii) usually store data from OLTP as well as from other sources like market data and competitors' data
  - (iv) are usually built as a database separate from OLTP databases
- e. The Slicing OLAP operation causes
  - (i) a reduction in the number of dimensions from an OLAP cube
  - (ii) an increase in the number of dimensions from an OLAP cube
  - (iii) no change in the number of dimensions from an OLAP cube
  - (iv) new facts to be added to an OLAP cube

- f. In the context of Association Rule Mining (considering that the same minimum value of support is maintained at each level):
  - (i) If an itemset is frequent, then all its subsets are also frequent
  - (ii) If an itemset is not frequent, then none of its subsets can be frequent
  - (iii) If an itemset is frequent, then all its supersets are also frequent

# (iv) If an itemset is not frequent, then none of its supersets can be frequent

- g. In the context of Association Rule Mining: NONE
  - (i) the support of an association rule is always greater than its confidence
  - (ii) the support of an association rule is always less than its confidence
  - (iii) the support of an association rule is always equal to its confidence
  - (iv) the support of an association rule can never be equal to its confidence
- h. The *a priori* approach to Association Rule Mining (i.e., the approach proposed by Agrawal and Srikant)
  - (i) starts with 1-itemsets and progressively searches for larger itemsets that are frequent
  - (ii) starts with the largest itemsets and progressively searches for smaller itemsets that are frequent
  - (iii) uses pruning to prevent unnecessary trips to the database by removing itemsets from the candidate set that are surely not frequent
  - (iv) will need as many passes through the database as the size of the largest frequent itemset
- i. In the context of Association Rule (AR) Mining,
  - (i) if there is a meaningful AR (i.e., a rule having support > min\_support and confidence > min\_confidence) of the form  $A \rightarrow B$ , where A and B are itemsets, then  $B \rightarrow A$  is also a meaningful AR.
  - (ii) if an itemset of size L is found to be frequent, then the maximum number of possible meaningful association rules derived from it is also L
  - (iii) if an itemset of size L is found to be not-frequent, then the minimum number of possible meaningful association rules derived from it is also L
  - (iv) the upward and downward closure properties may not hold if we change the values of minsupport at various levels (i.e., we vary it with the length of the itemset)
- j. Results of Association Rule Mining can be used to determine
  - (i) the layout of a store
  - (ii) which items to be put on offer jointly
  - (iii) which items are purchased jointly more often
  - (iv) purchase of which item(s) causes purchase of which other item(s)

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- a. In a star schema, usually:
  - (i) the fact table is normalized
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  - (iii) the dimension tables are de-normalized
  - (iv) the dimension tables are normalized
- b. In OLAP operations, roll-up can be achieved by:
  - (i) moving down a dimension hierarchy like quarter  $\rightarrow$  month, etc.
  - (ii) moving up a dimension hierarchy like month  $\rightarrow$  quarter, etc.
  - (iii) removing one or more dimensions
  - (iv) adding a new dimension
- c. It is not beneficial and practical to materialize all the views in a data cube when
  - (i) the number of levels in dimensional hierarchies are very large
  - (ii) the speed of retrieval is not the primary objective
  - (iii) disk space saving is not the primary objective
  - (iv) there are too many dimensions
- d. Data warehouses
  - (i) are built to support short term decision making of junior managers
  - (ii) are built to support long term decision making of top-level managers
  - (iii) are usually built as a database separate from OLTP databases
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- f. In the context of Association Rule Mining (considering that the same minimum value of support is maintained at each level):
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  - (ii) If an itemset is frequent, then all its supersets are also frequent
  - (iii) If an itemset is not frequent, then none of its subsets can be frequent
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- g. In the context of Association Rule Mining: NONE
  - (i) the support of an association rule can never be equal to its confidence
  - (ii) the support of an association rule is always equal to its confidence
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  - (ii) if an itemset of size L is found to be frequent, then the maximum number of possible meaningful association rules derived from it must be greater than L
  - (iii) the upward and downward closure properties may not hold if we change the values of minsupport at various levels (i.e., we vary it with the length of the itemset)
  - (iv) if an itemset of size L is found to be not-frequent, then the minimum number of possible meaningful association rules derived from it is also L
- j. Results of Association Rule Mining cannot be used to determine NONE
  - (i) the layout of a store
  - (ii) which items to be put on offer jointly
  - (iii) which items are purchased jointly more often
  - (iv) purchase of which item(s) causes purchase of which other item(s)