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Panel C, Batch C1

DEC Lab Assignment-6

Aim - Apply a-priori algorithm to find frequently occurring items from given data and generate strong association rules.

Objective:

- Learn frequent itemsets, closed itemsets, Market basket analysis.
- Calculate support, Confidence and lift
- To generate the association rules.

Theory:-

- A-priori Algorithm.

F_k - Frequent k itemsets

L_k - Candidate k itemsets.

Let $k=1$

Generate $F_1 = \{ \text{Frequent 1-itemsets} \}$

Repeat until F_k is empty

- Candidate Generation (Join Step): generate L_{k+1} from F_k
- Support Counting Count the support of each candidate in L_{k+1} by scanning the DB

- Candidate pruning: Prime Candidate itemsets L_{k+1} containing Subsets of length k that are infrequent. Eliminate candidates in L_{k+1} that are infrequent leaving only those that are frequent $\Rightarrow F_{k+1}$

Rule. Merge two Frequent itemsets with Size n if their first $n-1$ items are identical

- $F_3 = \{ABC, ABD, ABE, ACD, BCD, BDE, CDE\}$

Merge $(ABC, ABD) = ABCD$

Merge $(ABC, ABE) = ABCE$

Merge $(ABD, ABE) = ABDE$

We can't merge (ABD, ACD) because they share only prefix of length 1 instead of length 2

- Association Rule Generation

- Given a Frequent itemset L , Find all non-empty subsets $F \subset L$ such that $F \rightarrow L$
- F satisfies the minimum Confidence requirement.

If $\{A, B, C, D\}$ is a Frequent itemset, that Candidate rules:

$ABCD \rightarrow D$ $ABCD \rightarrow C$ $ABCD \rightarrow B$

$A \rightarrow BCD$ $B \rightarrow ACD$ $C \rightarrow ABD$

$AB \rightarrow CD$ $AC \rightarrow BD$ $AD \rightarrow BC$

$BD \rightarrow AC$ $CD \rightarrow AB$

If $|L| = k$ then there are $2^k - 2$ Candidate association rules (ignoring $L \rightarrow \emptyset$ and $\emptyset \rightarrow L$)

Input: Datasets

Output: Generating association Rules

Platform: Windows

Conclusion: Thus, we have learn to perform a-priori Algorithm in python Pandas

FAQs.

1. What is the a-priori principle?

Ans. "If an itemset is frequent, then all of its Subsets must also be frequent."

Apriori principle holds due to the following property of the support measure.

$$\forall X, Y: (X \subseteq Y) \Rightarrow S(X) \geq S(Y)$$

2. What are the different steps involved in the a priori algorithm?

Ans. Itemset generation Create candidate itemsets of $L=1$ by Scanning the database for individual items.

- Frequent itemset generation count the support of these candidate itemsets in the db and retain only those itemsets that meet the min. support

- Join: Generate candidate itemsets of length $(k+1)$ ^{by joining} pairs of Frequent itemsets of $l=k$

- Prune: Eliminate candidate itemsets that contain subsets of length k with low support as they are not frequent.

Repeat: Repeat the process by incrementing the length of itemsets until no more frequent itemsets can be found

3. What is a minimum Support threshold? What is a minimum Confidence threshold?

Ans. Minimum Support Threshold: Sets the minimum occurrence frequency for itemsets to be considered significant.

— Minimum Confidence Threshold: Establishes the minimum confidence level for association rules to be considered interesting.

4. How are frequent itemsets identified using a priori algorithm?

Ans. Frequent itemsets are identified using A-priori algorithm by iteratively generating and checking candidate itemsets of increasing lengths, keeping only those that meet the minimum support threshold. This process continues until no more frequent itemsets can be found.

5. How are association rules generated using a priori algorithm?

Ans. To generate association rules A-priori Algo —

1. Find frequent itemsets in the database.
2. Create candidate rules, from these itemsets.
3. Calculate the confidence for each rule.
4. Keep rules that meet a minimum confidence threshold.

5. The retained rules are the generated association

8. rules, Showing item associations in the database.

6. How can be a-priori algorithm be used to detect credit card fraud in real time?

Ans. Detecting credit card fraud in real time involves

1- Data Collection: Gather transaction data in real time, including transaction amounts, user details.

2- Feature Extraction: Extract relevant features from the transaction data.

3- Model Building: - Utilize machine learning models for rules based systems and current transaction data.

4- Real time monitoring: continuously monitoring incoming transactions and compare them to the established models and rules.

5- Anomaly Detection: - Identify transactions that deviate significantly from established patterns.

6- Alert Generation: Flag suspicious transactions for review.

7- Fraud Prevention: - Take actions like blocking, notifying or enhancing security.

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