Data Mining

Mining Frequent Patterns

Mining Frequent Patterns, Association and Correlations

- Basic concepts and a road map
- Efficient and scalable frequent itemset mining methods
- Mining various kinds of association rules
- ☐ From association mining to correlation analysis
- Constraint-based association mining
- □ Summary

What Is Frequent Pattern Analysis?

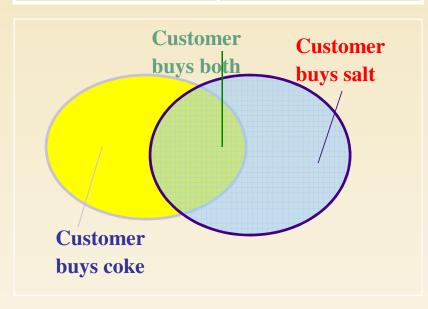
- Frequent pattern: a pattern (a set of items) that occurs frequently in a data set
- □ First proposed by Agrawal, Imielinski, and Swami [AIS93] in the context of frequent itemsets and association rule mining
- Motivation: Finding inherent regularities in data
 - What products were often purchased together?— Sabzi and Masala?!
 - > What are the subsequent purchases after buying a PC?
 - What kinds of DNA are sensitive to this new drug?
- Applications
 - Basket data analysis, cross-marketing, catalog design, sale campaign analysis, Web log (click stream) analysis, and DNA sequence analysis.

Why Is Freq. Pattern Mining Important?

- Discloses an intrinsic and important property of data sets
- Forms the foundation for many essential data mining tasks
 - > Association and correlation
 - Sequential and structural (e.g., sub-graph) patterns
 - Pattern analysis in multimedia, time-series, and stream data

Basic Concepts: Frequent Patterns and Association Rules

Transaction-id	Items bought
10	A, B, D
20	A, C, D
30	A, D, E
40	B, E, F
50	B, C, D, E, F



- $\Box \quad \text{Itemset X} = \{x_1, ..., x_k\}$
- □ Find all the rules $X \rightarrow Y$ with minimum support and confidence
 - ➤ support, s, probability that a transaction contains X ∪ Y
 - confidence, c, conditional probability that a transaction having X also contains Y

Let $sup_{min} = 50\%$, $conf_{min} = 50\%$ Freq. Pat.: {A:3, B:3, D:4, E:3, AD:3} Association rules:

$$A \rightarrow D$$
 (60%, 100%) $D \rightarrow A$ (60%, 75%)

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Scalable Methods for Mining Frequent Patterns

- The downward closure property of frequent patterns
 - Any subset of a frequent itemset must be frequent
 - > If {coke, salt, nuts} is frequent, so is {coke, salt}
 - > i.e., every transaction having {coke, salt, nuts}
 also contains {coke, salt}
- Scalable mining methods: Two major approaches
 - Apriori (Agrawal & Srikant@VLDB'94)
 - Freq. pattern growth (FPgrowth—Han, Pei & Yin @SIGMOD'00)

Apriori: A Candidate Generation-and-Test Approach

Apriori pruning principle: If there is any itemset which is infrequent, its superset should not be generated/tested! (Agrawal & Srikant @VLDB'94, Mannila, et al. @ KDD' 94)

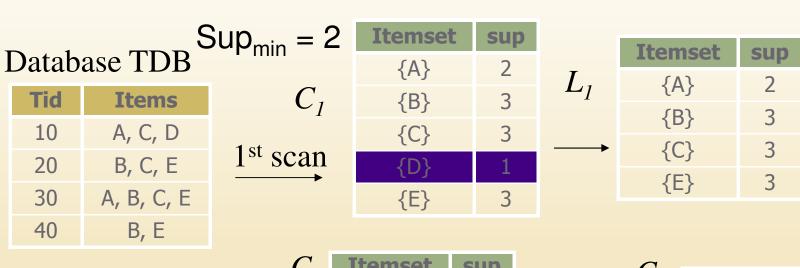
Method:

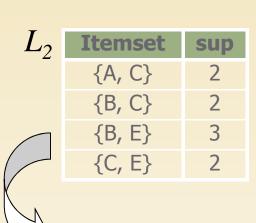
- > Initially, scan DB once to get frequent 1-itemset
- Generate length (k+1) candidate itemsets from length k frequent itemsets
- > Test the candidates against DB
- > Terminate when no frequent or candidate set can be generated

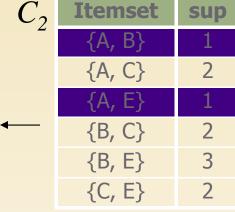
The Apriori Algorithm

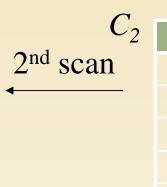
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Pseudo-code:
     C_k: Candidate itemset of size k
     L_{\nu}: frequent itemset of size k
     L_1 = \{ \text{frequent items} \};
     for (k = 1; L_k != \emptyset; k++) do begin
         C_{k+1} = candidates generated from L_k;
        for each transaction t in database do
              increment the count of all candidates in
                                   that are contained in
        C_{k+1}
        L_{k+1} = candidates in C_{k+1} with min_support
        end
     return \cup_k L_k;
```

The Apriori Algorithm—An Example









Itemset
{A, B}
{A, C}
{A, E}
{B, C}
{B, E}
{C, E}

C_3	Itemset	
3	{B, C, E}	

3 rd scan	L_3

Itemset	sup
{B, C, E}	2

Challenges of Frequent Pattern Mining

Challenges

- Multiple scans of transaction database
- > Huge number of candidates
- Tedious workload of support counting for candidates
- Improving Apriori: general ideas
 - > Reduce passes of transaction database scans
 - > Shrink number of candidates
 - > Facilitate support counting of candidates

Bottleneck of Frequent-pattern Mining

- Multiple database scans are costly
- Mining long patterns needs many passes of scanning and generates lots of candidates
 - > To find frequent itemset $i_1i_2...i_{100}$
 - # of scans: 100
 - # of Candidates: $\binom{100}{100} + \binom{100}{100} + \dots + \binom{100}{100} = 2^{100} 1 = 1.27 \times 10^{30}!$
- Bottleneck: candidate-generation-and-test
- Can we avoid candidate generation?

Partition: Scan Database Only Twice

- Any itemset that is potentially frequent in DB must be frequent in at least one of the partitions of DB
 - > Scan 1: partition database and find local frequent patterns
 - Scan 2: consolidate global frequent patterns
- A. Savasere, E. Omiecinski, and S. Navathe. An efficient algorithm for mining association in large databases. In VLDB'95

Sampling for Frequent Patterns

- Select a sample of original database, mine frequent patterns within sample using Apriori
- Scan database once to verify frequent itemsets found in sample, only borders
 of closure of frequent patterns are checked
 - > Example: check *abcd* instead of *ab, ac, ..., etc.*
- Scan database again to find missed frequent patterns
- □ H. Toivonen. Sampling large databases for association rules. In VLDB'96

Mining Frequent Patterns Without Candidate Generation

- Grow long patterns from short ones using local frequent items
 - "abc" is a frequent pattern
 - ➤ Get all transactions having "abc": DB|abc
 - "d" is a local frequent item in DB|abc → abcd is
 a frequent pattern

Closed Patterns and Max-Patterns

- □ A pattern $\{a_1, a_2, ..., a_n\}$ contains 2^n -1 sub-patterns > If n=100, $1.27*10^{30}$ sub-patterns!
- A manager may be only interested in patterns involving some items being managed (not all)
- Solution: Mine closed patterns and max-patterns instead
- □ An itemset X is closed if X is frequent and there exists no super-pattern Y ⊃ X, with the same support as X
- □ An itemset X is a max-pattern if X is frequent and there exists no frequent super-pattern Y > X

Closed Patterns and Max-Patterns

- □ Exercise. DB = { $<a_1, ..., a_{100}>$, $<a_1, ..., a_{50}>$ }

 ➤ Min_sup = 1.
- What is the set of closed itemset?

$$>$$
 < a_1 , ..., $a_{100}>$: 1

$$>$$
 < a_1 , ..., $a_{50}>$: 2

What is the set of max-pattern?

$$>$$
 < a_1 , ..., $a_{100}>: 1$

What is the set of all patterns?

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> !!
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