## Frequent Pattern Analysis

Definition

A pattern that occurs frequently in a data set

Importance

Disclose an intrinisic and important property of data sets

### Types of data or knowledge

- associative pattern
- sequential pattern
- Sub-Grapgh pattern
- Iceberg cube

## Main operations

• read / write / point

#### Other Methods

- closed / max pattern
- compression method
- pruning method
- constraints

### Association Rule

## 概念

• 规则 (Rule):

$$\{x_1,x_2,x_3,\ldots,x_n\} o Y$$

- 可信度 (Confidence) 和最小可信度
  - $\circ$  购买 $x_1, x_2, \ldots x_n$ 的情况下购买Y的可能性,条件概率
  - $\circ$  Confidence $(A \rightarrow B) = P(B|A)$
- 支持度 (Support) 和最小支持度
  - $\circ$  同时购买 $x_1, x_2, \dots x_n$ 和Y的可能性

$$\circ \ Support(A \rightarrow B) = P(A \cup B)$$

● 频繁项目集

满足最小支持度的项目集

### Example for calculate Support & Confidence

### $\{ABC, ACD, BCD, ADE, BCE\}$

| Rule       | Support | Confidence |
|------------|---------|------------|
| A -> D     | 2/5     | 2/3        |
| C -> A     | 2/5     | 2/4        |
| A -> C     | 2/5     | 2/3        |
| B & C -> D | 1/5     | 1/3        |

PS: 注意因果关系

### Evolution of AR (Association Rule)

- 1. AR Model
- 2. Apriori (层次算法产生候选集)
- 3. FP-Growth

### Sub Problems of AR

- 1. 依据最小支持度,寻找频繁项目集
- 2. 依据最小可信度,产生关联规则

### 重要公理

如果一个项目集S是频繁的(项目集S的出现频度大于最小支持度),那么S的任意子集是频繁的 Eg.  $\{a,b,c\}$  其子集  $\{a,b\}$ 

#### 其逆否命题

如果一个项目集合S是不频繁的,那么它的任何超集是不频繁的

Eg. {a} 其超集 {a,b}

# 算法

- 分层挖掘(每一层需要对数据做一次扫描)
  我们只需将精力放在大小为2的频繁项目集上
- 2. 对数据做1、2次扫描就找出频繁项目集(利用公理)

#### **Apriori**

- 1. self-joining  $oldsymbol{L_k}$
- 2. pruning

# Apriori Example (找出频繁集,建立关联规则)

$$\{ABC,AC,BCD,DE,ABCD\}$$
  $(Min_s=2[$ 常忽略分母 $],Min_c=80\%$  $)$ 

 $C_1$ 

| item | Freq |
|------|------|
| A    | 3    |
| В    | 3    |
| С    | 4    |
| D    | 3    |
| E    | 1    |

### $L_1$

| item | Freq |
|------|------|
| A    | 3    |
| В    | 3    |
| С    | 4    |
| D    | 3    |

 $C_2$ 

item Freq

| AB | 2 |
|----|---|
| AC | 3 |
| AD | 1 |
| BC | 3 |
| BD | 2 |
| CD | 2 |

## $L_2$

| item | Freq |
|------|------|
| AB   | 2    |
| AC   | 3    |
| BC   | 3    |
| BD   | 2    |
| CD   | 2    |

# $C_3$

| item | Freq |
|------|------|
| ABC  | 2    |
| BCD  | 2    |

# $L_3$

| item | Freq |
|------|------|
| ABC  | 2    |
| BCD  | 2    |

# Based on $L_2$ We can get

| Rule | Confidence |  |
|------|------------|--|
|      |            |  |

| A -> B | 2/3 |
|--------|-----|
| B -> A | 2/3 |
| A -> C | 1   |
| C -> A | 3/4 |
| ···    | ··· |

## Based on $L_3$ We can get

| Rule    | Confidence |
|---------|------------|
| A -> BC | 2/3        |
| B-> AC  | 2/3        |
| C -> AB | 2/4        |
| AB -> C | 1          |
|         |            |

#### 算法缺陷:候选集的生成耗费太大

## 算法改进

- 基于Hash
- 基于Partition
- 基于Sample

# Graph Mining

A (sub)graph is *frequent* if its support in a given dataset is no less than a minimum support threshold

Subgraph Explosion Problem