

# Association & Correlation in Data Mining

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# What is Association Analysis?

## **Content:**

- Retail product placement
- Recommendation systems (Amazon, Netflix)
- Medical diagnosis patterns
- Fraud detection
- Web usage mining

# Market Basket Analysis

## **Content:**

- A technique used to find associations between products purchased together.
- Helps retailers:
  - Improve product placement
  - Design combo offers
  - Increase sales

## **Example:**

Milk → Bread → Butter

# Key Terms in Association Rules

## Content:

- **Itemset:** Collection of one or more items.
- **Support:** Frequency of itemset in dataset.
- **Confidence:** Likelihood that item B is purchased when item A is purchased.
- **Lift:** Measures strength of association rule.

## Formula:

- $\text{Support}(A) = \text{Transactions containing } A / \text{Total transactions}$
- $\text{Confidence}(A \rightarrow B) = \text{Support}(A \cup B) / \text{Support}(A)$

# Example of Association Rule

## **Content:**

Rule:  $\{\text{Milk}\} \rightarrow \{\text{Bread}\}$

- Support = 40%
- Confidence = 70%
- Lift  $> 1 \rightarrow$  Strong relationship

## **Meaning:**

Customers buying milk are likely to buy bread.



# Apriori Algorithm (Overview)

## Content:

- Used to find frequent itemsets.
- Based on the **Apriori Principle**:  
If an itemset is frequent, all its subsets must also be frequent.

## Steps:

1. Find frequent 1-itemsets.
2. Generate candidate itemsets.
3. Prune infrequent itemsets.
4. Repeat until no new itemsets.

# Apriori Algorithm Example

## Content:

### Transactions:

- T1: Milk, Bread
- T2: Milk, Butter
- T3: Milk, Bread, Butter

### Frequent Itemsets:

- Milk ✓
- Bread ✓
- Milk & Bread ✓

# Mining Without Candidate Generation (FP-Growth)

## Content:

- FP-Growth improves efficiency over Apriori.
- Uses **FP-Tree** structure.
- Avoids generating candidate itemsets.

## Advantages:

- Faster
- Less memory usage
- Suitable for large datasets



# Vertical Data Format Mining

## Content:

- Stores data in vertical format (item  $\rightarrow$  transaction IDs).
- Makes support counting faster.
- Used in algorithms like **ECLAT**.

## Example:

Milk  $\rightarrow$  T1, T2, T3

Bread  $\rightarrow$  T1, T3

# Closed Frequent Itemsets

## **Content:**

- A frequent itemset with no superset having same support.
- Reduces redundancy.
- Helps compress frequent patterns.

## **Benefit:**

Smaller result set, same information.

# Multilevel Association Rules

## **Content:**

- Finds associations at different abstraction levels.

## **Example:**

- Level 1: Dairy  $\rightarrow$  Bakery
- Level 2: Milk  $\rightarrow$  Bread

**Use:** Retail product hierarchy.

# Multidimensional Association Rules

## **Content:**

- Rules involving multiple attributes.

## **Example:**

Age = 20–30 AND Buys = Laptop  $\rightarrow$  Buys = Headphones

## **Used in:**

- Customer profiling
- Targeted marketing

# Correlation Analysis

## Content:

- Measures strength of relationship between variables.
- Helps identify meaningful patterns.

## Types:

- Positive correlation
- Negative correlation
- No correlation

## Example:

Temperature  $\uparrow \rightarrow$  Ice cream sales  $\uparrow$



# Constraint-Based Association Mining

## **Content:**

- Applies constraints to focus on useful rules.
- Reduces number of irrelevant patterns.

## **Types of Constraints:**

- Item constraints
- Rule constraints
- Aggregate constraints

## **Benefit:**

Faster and more relevant results.

# Applications of Association Mining

## **Content:**

- Retail product placement
- Recommendation systems (Amazon, Netflix)
- Medical diagnosis patterns
- Fraud detection
- Web usage mining

Thank You !