#### Introduction

- Goal is
  - to find associations between different objects
  - to find patterns in objects
  - to derive strategies (store layout, catalogue design, customizing emails, recommender engines, ...)
- Applicable for
  - Marketing
  - Clustering
  - Classification
  - Retailing

#### Principle

Example: Supermarket

Product Portfolio → Item Set



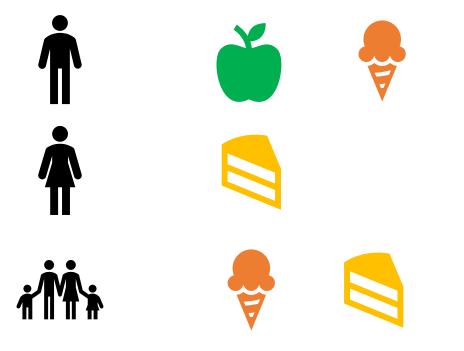




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### Principle

Input: Transactions





Transaction 2

Transaction 3

### Principle

#### Transactions

Transaction	Items
1	{apple, ice}
2	{cake}
3	{ice, cake}
•••	<b>{}</b>
N	<b>{}</b>

### Principle

Output: Rules

[LHS]	[RHS]
{ice}	{cake}
{apple, cake}	{ice}

Key Parameters: Support

■ 
$$Support(X) = \frac{Number\ Transactions\ including\ X}{Total\ Transactions\ Number} = \frac{count(X)}{N}$$
■  $Support(\{X\} \to \{Y\}) = \frac{Transactions\ containing\ X\ AND\ Y}{Total\ Transaction\ Number}$ 

Support threshold can be defined to exclude infrequent transactions

Key Parameters: Support - Example

- Calculate Support(Apple → Pie)!
- Minimum Support = 0.4
- Count(apple&pie) = 3
- Total Number= 5
- Min Support excludes infrequent itemsets from further analysis

Transact. ID			
1	Č		
2		<b>Č</b>	
3	<b>*</b>	Č	
4	Ď		
5		-	Ď

Key Parameters: Confidence

• 
$$Confidence(X \to Y) = \frac{Support(X,Y)}{Support(X)}$$

How often is Y bought, after X was bought?

- Problem: only considers popularity of X (not of Y)
  - If X and Y are popular, buying X and Y together is likely. So high confidence occurs by chance.

Key Parameters: Lift

■ 
$$Lift(X \to Y) = \frac{Support(X,Y)}{Support(X) * Support(Y)}$$

- How often is Y bought, after X was bought?
- This takes popularity of Y into account.