Association analysis with Apriori algorithm

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Content

- Motivation
- Relevant definitions
- Apriori Algorithm
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- Code demonstration



Motivation

- Any institution with active participants generates a lot of data.
- Not all of genrated data is useful.
- Trying to find some patterns in this data that leads to a desired outcome for the institution is the point of association analysis.

Motivation, examples

- **Employer**: Freshers and roles.
- Grocery store: Arranging items and profits.
- Voter: Politicians and policies.
- **Blogger**: Blogs and trending keywords.
- Counter-Terrorist: Actions and consequences.
- **Doctor**: Diagnosis and treatment.
- Lover: Actions and consequences.

Definitions

- Frequent itemsets: Collection of items that occur frequently.
- **Association rules**: Heuristic rule that predicts the association between two items.
- **Association analysis**: Finding frequent itemsets or association rules in large datasets .

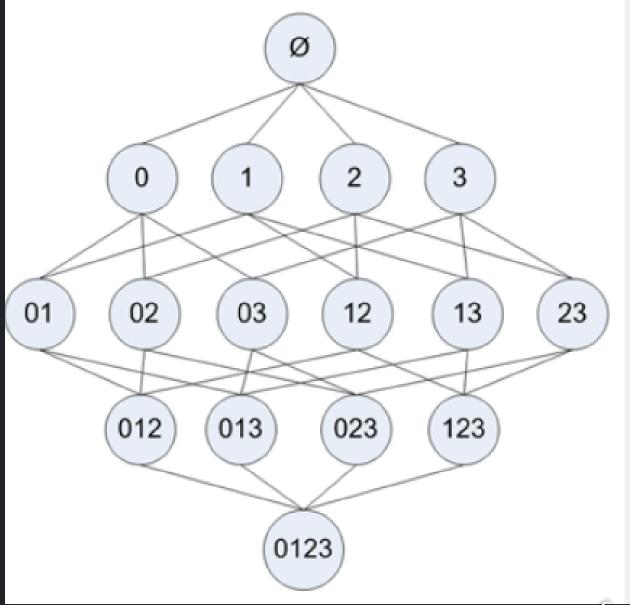
Definitions, continued

Support of CI:

```
rac{\#sets \supseteq CI}{\#Dataset}.
```

Confidence of A ->

B:
$$\frac{support\{A,B\}}{support\{B\}}$$
.



Apriori Algorithm Principle

- Apriori principle: All subsets of a frequent itemset must be frequent.
- Our implementation's principle: If an itemset is infrequent, all its supersets will be infrequent or the contrapositive of the Apriori principle.

Frequent itemsets: Helper functions pseudo code

For each transaction in tran the dataset:

For each candidate itemset, can:

Check to see if can is a subset of tran

If so increment the count of can

For each candidate itemset:

If the support meets the minimum, keep this item

Return list of frequent itemsets

Frequent itemsets: Pseudocode

While the number of items in the set is greater than 0: Create a list of candidate itemsets of length k Scan the dataset to see if each itemset is frequent Keep frequent itemsets to create itemsets of length k+1

Association rules from frequent itemsets

- Step 1: Compute support of each frequent itemset.
- **Step 2**: Set cutoff confidence level.
- **Step 3**: Generate all possible rules from frequent itemsets.
- **Step 4**: Compute confidence of each rule.
- **Step 5**: If rule's confidence \geq cutoff, then add rule to list of rules.

Pros & Cons

• Pros:

- -Easy to code.
- -Works with numeric and nominal values.
- -Determininstic.

• Cons:

- -Time and space complexity: $O(2^n)$.
- -Ancient, introduced in 1994.
- -Too simple.

References:

- Machine Learning in Action by Peter Harrington: http://www2.ift.ulaval.ca/~chaib/IFT-4102-7025/public_html/Fichiers/Machine_Learning_in_Action.pd
- The Numerati by Stephen Baker: https://www.amazon.com/Numerati-Stephen-Baker/dp/B003TO6G20
- Fast Algorithms for mining Association Rules: http://www.vldb.org/conf/1994/P487.PDF