## Birla Institute of Technology and Science-Pilani, Hyderabad Campus

## **SEMESTER-II 2018-2019**



## Data Mining (CS F415)

## Association Rule Mining

### $\underline{BY}$

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# Dataset

#### **Groceries Market Basket Dataset**

Number of transactions: 9835Number of unique items: 169

# Pre-processing done on data

Groceries.csv file was read transaction by transaction and each transaction was saved as a list. A mapping was created from the unique items in the dataset to integers so that each item corresponded to a unique integer. The entire data was mapped to integers to reduce the storage and computational requirement. A reverse mapping was created from the integers to the items, so that the item names could be written in the final output file.

# Formulas Used

- Confidence  $(X \rightarrow Y) = \text{support } (X \cup Y) / \text{support } (X)$
- Support (X, Y) = support-count (X, Y) / total dataset size

# Results for different for values of support and confidence

Confidence/Support	No. of Frequent Itemsets	No. of rules
High confidence (MIN_CONF=0.5) High support count(MINSUP=60)	725	65
Low confidence (MIN_CONF=0.1) High support count(MINSUP=60)	725	1219
High confidence (MIN_CONF=0.5) Low support count(MINSUP=10)	11390	4484
Low confidence (MIN_CONF=0.1) Low support count(MINSUP=10)	11390	34497

### Redundant rules:

We kept a support count = 10 and confidence = 0.1 to get 12 non-closed frequent itemsets from which we found 49 redundant rules. Some of which are:

[These are in format  $X(X_{sup})-Y(Y_{sup})$ , confidence]

- 1. rice(75)->sugar(333), 0.16
- 2. brown bread, whipped/sour cream(46)-> pip fruit, 744, 0.24
- 3. pip fruit, whipped/sour cream(91)->brown bread(638), 0.12

### Observation:

Apriori takes significantly more time than Fp Tree algorithm as FP tree algorithm uses divide and conquer approach.

Time taken for frequent item generation(Apriori) for Minimum Support=10: 25.04 sec

Time taken for frequent item generation(FP Growth) for Minimum Support=10: 3.74 sec

Most of the rules we generated have a common item (whole milk and other vegetables) on the consequent side. This happens when any item is very frequent in the transactions. This can be avoided by using lift instead of confidence.

Lift  $(X \rightarrow Y) = \text{support } (X \cup Y) / \text{support } (X) * \text{support } (Y)$