

# Experiment 7 apriori algorithm

Data Mining Techniques (Vignan's Foundation for Science, Technology and Research)



Scan to open on Studocu

## Experiment 7. Implementation of apriori algorithm

Aim:To Perform the implementation of Apriori algorithm

Resources: Weka

#### Algorithm:

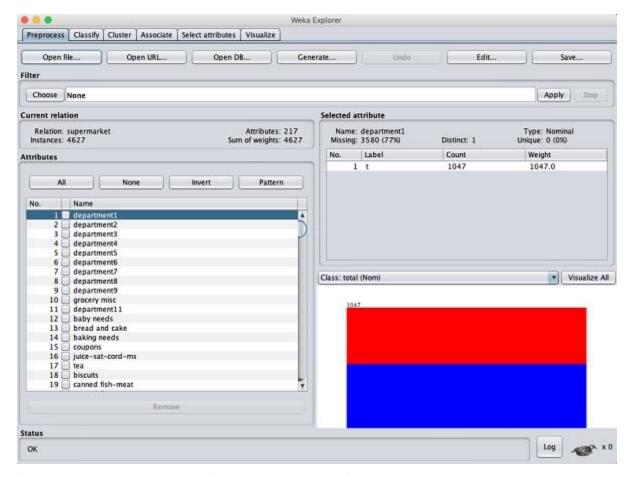
```
Algorithm: Apriori. Find frequent itemsets using an iterative level-wise approach based on candidate
   generation.
  D, a database of transactions;
   min_sup, the minimum support count threshold.
Output: L, frequent itemsets in D.
         L_1 = find\_frequent\_1-itemsets(D);
        for (k = 2; L_{k-1} \neq \emptyset; k++) {
C_k = \operatorname{apriori-gen}(L_{k-1});
(3)
            for each transaction t \in D { // scan D for counts
(4)
                  C_t = \text{subset}(C_k, t); // get the subsets of t that are candidates
(6)
                 for each candidate c \in C_t
(7)
                      c.count++;
(9)
            L_k = \{c \in C_k | c.count \ge min\_sup\}
(10)
(11)
        return L = \bigcup_k L_k;
procedure apriori_gen(L_{k-1}:frequent (k-1)-itemsets)
       for each itemset l_1 \in L_{k-1}
            for each itemset l_2 \in L_{k-1} if (l_1[1] = l_2[1]) \wedge (l_1[2] = l_2[2]) \wedge ... \wedge (l_1[k-2] = l_2[k-2]) \wedge (l_1[k-1] < l_2[k-1]) then { c = l_1 \bowtie l_2; // join step: generate candidates
(2)
(3)
(4)
(5)
                      if has_infrequent_subset(c, L_{k-1}) then
                           delete c; // prune step: remove unfruitful candidate
(6)
(7)
                      else add c to C_k;
(8)
        return Cb;
procedure has_infrequent_subset(c: candidate k-itemset;
            L_{k-1}: frequent (k-1)-itemsets); // use prior knowledge
         for each (k-1)-subset s of c
          if s \not\in L_{k-1} then
(2)
                 return TRUE:
(3)
       return FALSE;
(4)
```

#### Procedure:

The **Apriori** algorithm is one such algorithm in ML that finds out the probable associations and creates association rules. WEKA provides the implementation of the Apriori algorithm. You can define the minimum support and an acceptable confidence level while computing these rules. You will apply the **Apriori** algorithm to the **supermarket** data provided in the WEKA installation.

### **Step 1:Loading Data**

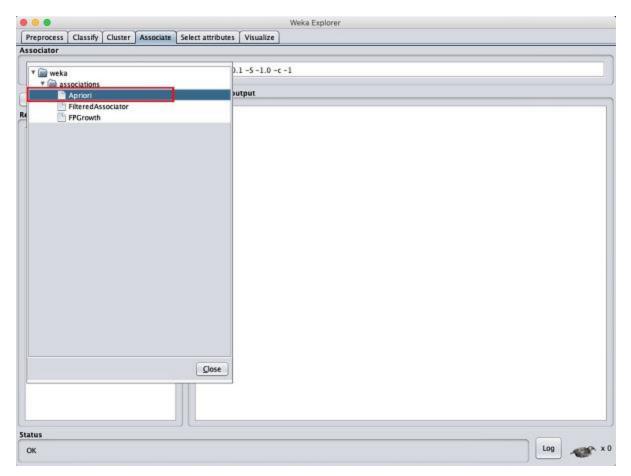
In the WEKA explorer, open the **Preprocess** tab, click on the **Open file** ... button and select **supermarket.arff** database from the installation folder. After the data is loaded you will see the following screen –



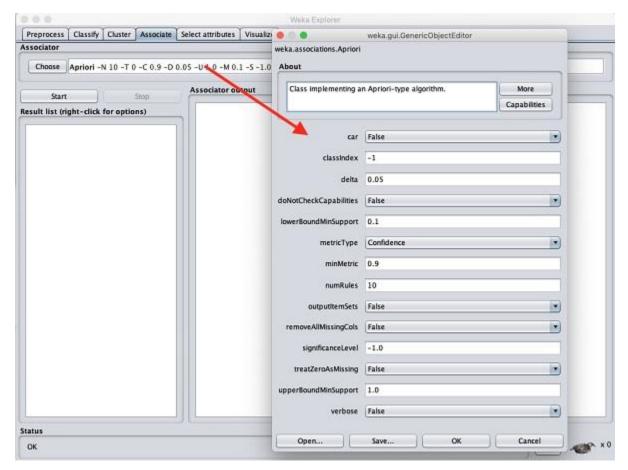
The database contains 4627 instances and 217 attributes. You can easily understand how difficult it would be to detect the association between such a large number of attributes. Fortunately, this task is automated with the help of Apriori algorithm.

## **Step 2:Associator**

Click on the **Associate** TAB and click on the **Choose** button. Select the **Apriori** association as shown in the screenshot –

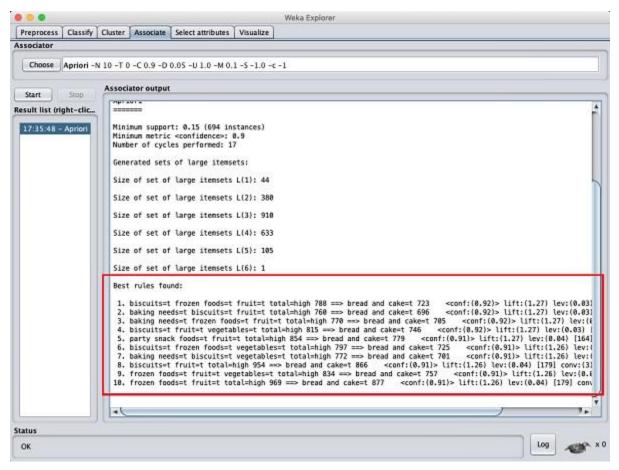


To set the parameters for the Apriori algorithm, click on its name, a window will pop up as shown below that allows you to set the parameters –



After you set the parameters, click the **Start** button. After a while you will see the results as shown in the screenshot below –

#### Output:



At the bottom, you will find the detected best rules of associations. This will help the supermarket in stocking their products in appropriate shelves.