file/after/tracklang.stypackage/after/bidipackage/after/luabidi

## **Mining Frequent Patterns, Associations, and Correlations**

When you have a set of transactions in a dataset, some of them can happen frequently. For a dataset D, there are many combinations of the items, so how do we find the ones that happen frequently?

A **frequent itemset** is one whose **support** and **confidence** are above a certain threshold. The support for a rule  $A \implies B$ ) is the percentage of all transactions that contain  $A \cup B$ , or their intersection. The confidence for a frequent itemset is the percentage of all transactions containing A that also contain B, i.e. P(B|A).

A **closed itemset** is one where for all the itemsets X there is no proper super-itemset  $Y(X \boxtimes Y)$  such that X and Y both have the same support count.

## **Apriori Algorithm**

This algorithm uses knowledge of previous frequent itemsets to calculate the current one, e.g. it uses  $L_1$  to calculate  $L_2$ , the set of frequent 2-itemsets. The idea is that we first find all the 1-itemsets and use that to find the 2-itemsets. Since all the 1-itemsets are proper subsets of 2-itemsets, we could build up our  $L_k$  this way. However, for every level of itemsets we need to scan the entire database :(

The central idea of the algorithm is that an itemset I is not frequent, then all of its supersets will not be frequent either.  $P(I \cup A)$  cannot be more frequent than P(a)I when we add an item A.

The first step in the algorithm is the join step. For two subsets of  $L_{k-1}$ ,  $l_1$  and  $l_2$ , we join them if their first k-2 elements are equal, and  $l_1[k-1] < l_2[k-1]$ . This way we produce a subset that is still lexicographically ordered and contains one more element in it than before.

The second step is the prune step. We generated a candidate set  $C_k$ , and maybe not every itemset here will be frequent. If any k-1 itemset in  $C_k$  is not in  $L_{k-1}$ , then we know it can't be frequent thus it can be removed from  $C_k$ . (Here is where we can use a **hash tree** for quick searching of frequent itemsets)