

Algorithm: FP_growth. Mine frequent itemsets using an FP-tree by pattern fragment growth.

Input:

- D , a transaction database;
- min_sup , the minimum support count threshold.

Output: The complete set of frequent patterns.

Method:

1. The FP-tree is constructed in the following steps:
 - (a) Scan the transaction database D once. Collect F , the set of frequent items, and their support counts. Sort F in support count descending order as L , the list of frequent items.
 - (b) Create the root of an FP-tree, and label it as “null.” For each transaction $Trans$ in D do the following. Select and sort the frequent items in $Trans$ according to the order of L . Let the sorted frequent item list in $Trans$ be $[p|P]$, where p is the first element and P is the remaining list. Call `insert_tree([p|P], T)`, which is performed as follows. If T has a child N such that $N.item_name = p.item_name$, then increment N ’s count by 1; else create a new node N , and let its count be 1, its parent link be linked to T , and its node-link to the nodes with the same *item-name* via the node-link structure. If P is nonempty, call `insert_tree(P, N)` recursively.
2. The FP-tree is mined by calling `FP_growth(FP_tree, null)`, which is implemented as follows.


```

procedure FP_growth(Tree,  $\alpha$ )
(1)  if Tree contains a single path  $P$  then
(2)    for each combination (denoted as  $\beta$ ) of the nodes in the path  $P$ 
(3)      generate pattern  $\beta \cup \alpha$  with support_count = minimum support count of nodes in  $\beta$ ;
(4)  else for each  $a_i$  in the header of Tree {
(5)    generate pattern  $\beta = a_i \cup \alpha$  with support_count =  $a_i.support\_count$ ;
(6)    construct  $\beta$ ’s conditional pattern base and then  $\beta$ ’s conditional FP-tree  $Tree_\beta$ ;
(7)    if  $Tree_\beta \neq \emptyset$  then
(8)      call FP_growth(Tree $_\beta$ ,  $\beta$ ); }
```

Figure 5.9 The FP-growth algorithm for discovering frequent itemsets without candidate generation.