## Bottleneck of Frequent-pattern Mining

- Multiple database scans are costly
- Mining long patterns needs many passes of scanning and generates lots of candidates
  - To find frequent itemset  $i_1i_2...i_{100}$ 
    - # of scans: 100
    - # of Candidates:  $\binom{100}{100} + \binom{100}{100} + \dots + \binom{100}{100} = 2^{100} 1 = 1.27*10^{30}!$
- Bottleneck: candidate-generation-and-test
- Can we avoid candidate generation?

## FP- Tree

- Definition: A frequent pattern (FP) tree is a tree structure consisting of an item-prefix-tree and a frequent item header table
- Item-prefix-tree
  - Root node (label-null)
  - 3 fields for each non root node
    - Item name
    - Support count
    - Node link (with link to node with same item name)

- Frequent item header table
  - Item name
  - Support count
  - Head of node link which points to the first node in the FP tree carrying the item name
- FP-tree features
  - Dependent on support threshold  $\sigma$ . (i.e., for different values of  $\sigma$ , the trees are different)
  - Depends on ordering of items (decreasing order of support count)

## FP-growth

- Adopts divide and conquer strategy
- First, compresses the database representing frequent items into a frequent pattern tree (FPtree) which retains itemset association information
- Divides compressed database into a set of conditional databases each associated with one frequent item or pattern segment

## **Benefits**

## Completeness

 Preserve complete information for frequent pattern mining

### Compactness

- Reduce irrelevant info—infrequent items are gone
- Items in frequency descending order: the more frequently occurring, the more likely to be shared
- Never be larger than the original database

- 1. Scan DB once, find frequent 1-itemset (single item pattern)
- 2. Sort frequent items in frequency descending order, L-list
- 3. Scan DB again, construct FP-tree based on L-list
- 4. Mine frequent item sets using conditional pattern base, derived from FP-tree

#### NOTE:

The transformed prefix path of a node 'a' from a truncated database of patterns which co-occur with a is called a conditional pattern base

## Find Patterns Having P From P-conditional Database

- Starting at the frequent item header table in the FP-tree
- Traverse the FP-tree by following the link of each frequent item p
- Accumulate all of transformed prefix paths of item p to form p's conditional pattern base

### From Conditional Pattern-bases to Conditional FP-trees

- For each pattern-base
  - Accumulate the count for each item in the base
  - Construct the FP-tree for the frequent items of the pattern base

### Recursion: Mine Each Conditional FP-tree

## **Example: FP-growth**

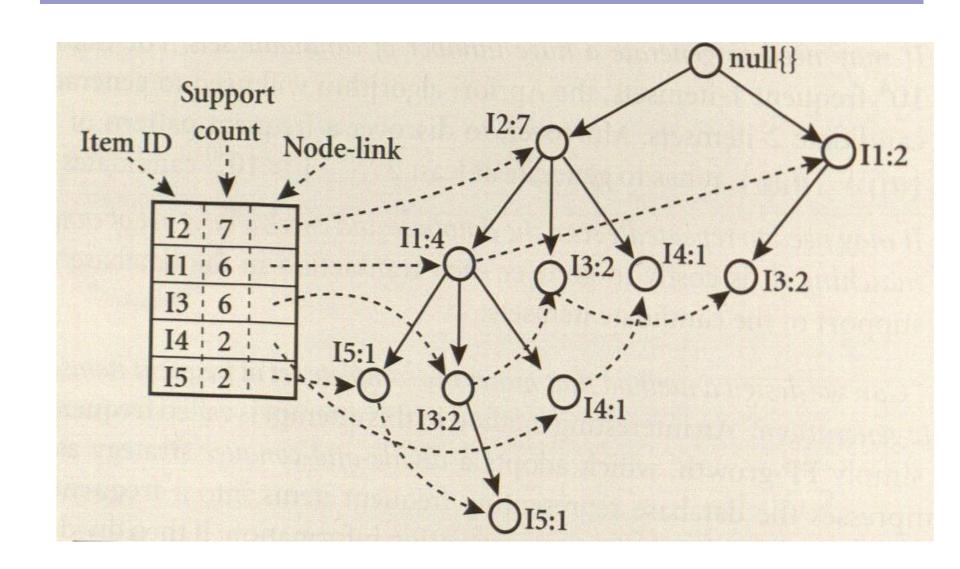
- The first scan of data is the same as Apriori
- Derive the set of frequent 1itemsets
- Let min-sup=2
- Generate a set of ordered items

Item ID	Support count
12	7
11	6
13	6
14	2
15	2

#### **Transactional Database**

TID	List of item IDS
T100	11,12,15
T200	12,14
T300	12,13
T400	11,12,14
T500	11,13
T600	12,13
T700	11,13
T800	11,12,13,15
T900	11,12,13

# Example:



Item	Conditional Pattern Base	Conditional FP-tree	Frequent Pattern Generated
<b>I</b> 5	{{I2, I1:1}, {I2, I1, I3:1}}	(I2:2, I1:2)	{I2, I5:2}, {I1, I5:2}, {I2, I1, I5:2}
14	{{I2, I1:2}, {I2:1}}	(I2:2)	{I2, I4:2}
13	{{I2, I1:2}, {I2:2}, {I1:2}}	(I2:4, I1:2), (I1:2)	{I2, I3:4}, {I1, I3:4}, {I2, I1, I3:2}
<b>I</b> 1	{{I2:4}}	(12:4)	{I2, I1:4}

Т

Т

<u>TID</u>	Items bought	(ordered) frequent items
100	$\{f, a, c, d, g, i, m\}$	$\{p,p\}$
<b>200</b>	$\{a, b, c, f, l, m, o\}$	
<b>300</b>	$\{b, f, h, j, o, w\}$	
<b>400</b>	$\{b, c, k, s, p\}$	
<b>500</b>	$\{a, f, c, e, \overline{l}, p, m\}$	, n

 $min\_support = 3$ 

# Pattern Count Tree (PC-Tree)

- PC-Tree is a data structure used to store all the patterns occurring in the tuples of a transaction database
- Node structure

	item_name	count	child_pointer(c)	sibling_pointer(s)
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- Construction of the PC-tree
  - Scan DB only once
  - Preprocess each transaction to put items in lexicographical order
  - For each transaction
    - If subpattern/prefix does not exist
      - Create a new branch
    - Else
      - Increment count

- As the data base is scanned for constructing pc tree, maintain a table with items & count initialised to zero. As transactions are read increment the count of respective items in the transaction
- Use this table to create a list L1 based on decreasing order of frequency.
- Read the pc tree from the root, and insert the transaction in fp-tree; decrement the count in pc tree as the transaction is inserted in fp tree
- First traverse child links, then sibling links
- At the end count of all items become zero in pctree

## PC Tree - properties

- PC-tree is a complete representation of the database
- PC-tree is a compact representation of the database
- PC-tree corresponding to a given database is unique