Association Rules – FP Trees and FP-Growth Algorithm

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Basic Idea

- Improve association rule mining by maintaining a special tree that tracks frequent patterns => Frequent Pattern (FP) tree.
- Two steps:
 - Construct FP-tree from data
 - Extract association rules.

FP-Tree Node Structure

- Each node has the following fields:
 - Item (item labeling the node. Root is labeled NULL)
 - Count (number of items labeling the node)
 - Next (link to next node with the same item)
 - LCHILD (left child)
 - RCHILD (right child)
- A separate *Itemset* table has schema (*Item,First*)
 where First is a pointer to the first node labeled
 with *Item*.

Running Example

Transaction	Items
1	A,B,D,E,F
2	A,B,E,F,G
3	A,B,C,F,H
4	C,E,D,G,H
5	A,C,E,F
6	A,B,E,H
7	B,C,F,G
8	A,D,E,F,G
9	A,C,D,H
10	B,E,F,G,H

Step 1. Build a table with a COUNT of each item (i.e. how many transactions each item occurs in.

Transaction	Items
1	A,B,D,E,F
2	A,B,E,F,G
3	A,B,C,F,H
4	C,E,D,G,H
5	A,C,E,F
6	A,B,E,H
7	B,C,F,G
8	A,D,E,F,G
9	A,C,D,H
10	B,E,F,G,H

Item	Count
А	7
В	6
С	5
D	4
Е	7
F	7
G	5
Н	5

Step 2. Sort all items in descending order of count.

Transaction	Items
1	A,B,D,E,F
2	A,B,E,F,G
3	A,B,C,F,H
4	C,E,D,G,H
5	A,C,E,F
6	A,B,E,H
7	B,C,F,G
8	A,D,E,F,G
9	A,C,D,H
10	B,E,F,G,H

Item	Count
Α	7
В	6
С	5
D	4
E	7
F	7
G	5
Н	5

SORTED LIST: A,E,F,B,C,G,H,D

Step 3. Sort item lists in transaction table in descending order., ignoring cases where count of an item is below support threshold. E.g. Support Threshold = 5. So D is ignored.

Transaction	Items
1	A,B,D,E,F
2	A,B,E,F,G
3	A,B,C,F,H
4	C,E,D,G,H
5	A,C,E,F
6	A,B,E,H
7	B,C,F,G
8	A,D,E,F,G
9	A,C,D,H
10	B,E,F,G,H

Transaction	Items
1	A,E,F,B,D
2	A,E,F,B,G
3	A,F,B,C,H
4	E,C,G,H,D
5	A,E,F,C
6	A,E,B,H
7	F,B,C,G
8	A,E,F,G, D
9	A,C,H,D
10	E,F,B,G,H

SORTED LIST: A,E,F,B,C,G,H,D

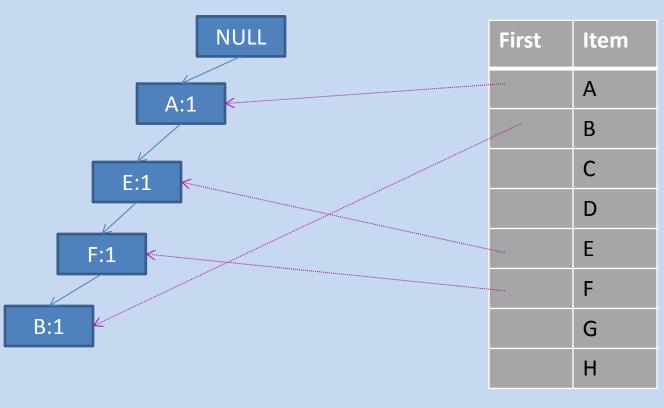
Step 4. Build Tree, processing one tuple at a time

NULL

Item	First
Α	
В	
С	
D	
E	
F	
G	
Н	

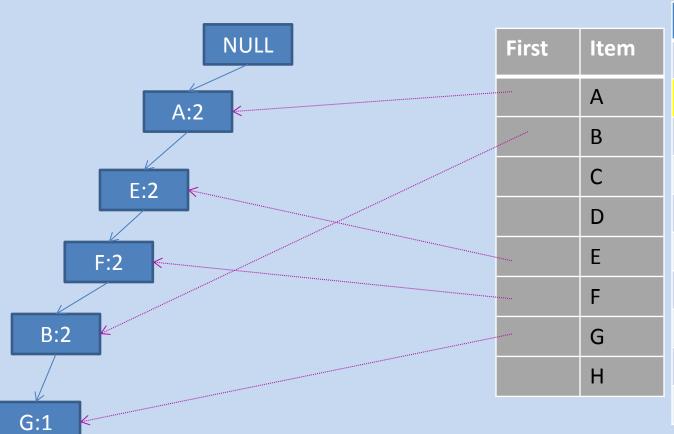
Trans	Items
1	A,E,F,B,D
2	A,E,F,B,G
3	A,F,B,C,H
4	E,C,G,H,D
5	A,E,F,C
6	A,E,B,H
7	F,B,C,G
8	A,E,F,G,D
9	A,C,H,D
10	E,F,B,G,H

Step 4. Process transaction 1



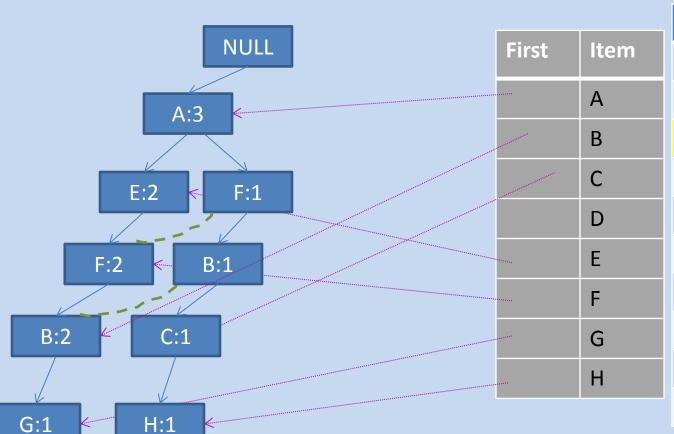
Trans	Items
1	A,E,F,B,D
2	A,E,F,B,G
3	A,F,B,C,H
4	E,C,G,H,D
5	A,E,F,C
6	A,E,B,H
7	F,B,C,G
8	A,E,F,G,D
9	A,C,H,D
10	E,F,B,G,H

Step 4. Process transaction 2



Trans	Items
1	A,E,F,B,D
2	A,E,F,B,G
3	A,F,B,C,H
4	E,C,G,H,D
5	A,E,F,C
6	A,E,B,H
7	F,B,C,G
8	A,E,F,G,D
9	A,C,H,D
10	E,F,B,G,H

Step 4. Process transaction 2



Trans	Items
1	A,E,F,B,D
2	A,E,F,B,G
3	A,F,B,C,H
4	E,C,G,H,D
5	A,E,F,C
6	A,E,B,H
7	F,B,C,G
8	A,E,F,G,D
9	A,C,H,D
10	E,F,B,G,H

Example Continued

Build rest of FP-tree in class on board.

Key Properties of FP-tree

- What is the max height of an FP-tree?
 Bounded above by the max number of items in a transaction that survive the support requirement.
- What is the max number of nodes in an FPtree? Bounded above by the sum of the items in each transaction.

FP-Growth: Finding the Frequent Itemsets

- The algorithm follows three steps.
 - Step 1: Here, we look at the Itemset table. For each item, we construct a conditional pattern base (CPB).
 - Step 2: We build a tree using the CPB above. This
 is called the conditional pattern tree.
 - Step 3: We analyze this tree.

FP-Growth: Step 1

 We create a table, starting from the "bottom" of the *Itemset* table (after sorting in descending order of counts and ignoring any items that don't meet support criteria).

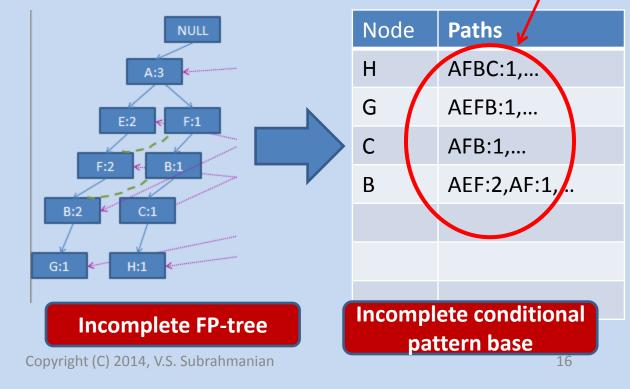
Item	Count		Item	Count
А	7		A	7
В	6		Е	7
С	5		F	7
D	4		В	6
E	7	,	С	5
F	7		G	5
G	5		Н	5
Н	5	pyright (C) 2014, V.S. Sub	₽	4

FP-Growth: Step 1

So in this case, we start with H.

We create a new table which shows all paths from Pattern
 Root to H (not incl. H) and the count of H

Item	Count
A	7
Е	7
F	7
В	6
C G	5
G	5
Н	5
₽	4



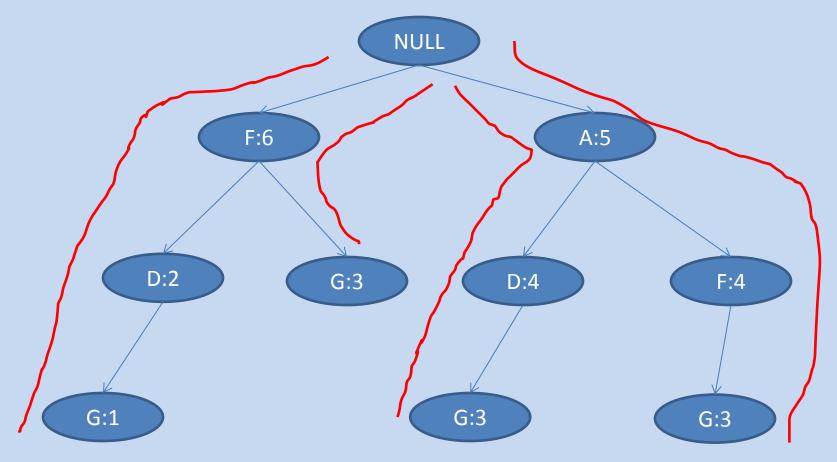
Example Continued

Build rest of Conditional Pattern Base in class on board.

FP-Growth: Step 1

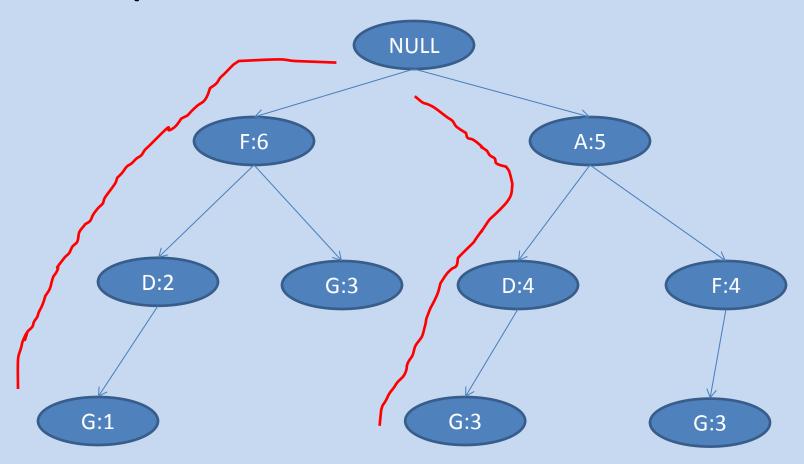
- Extract frequent itemsets, starting from the leaves of the FP-tree and working its way up.
- For each node N:
 - Use Itemsets Table to find places where N occurs.
 Work your way up.
 - Build a "prefix path subtree"
- Later, we can use the prefix path subtree to extract paths ending in suffixes, not just ending in a node.

Example: Prefix Path Subtree for G



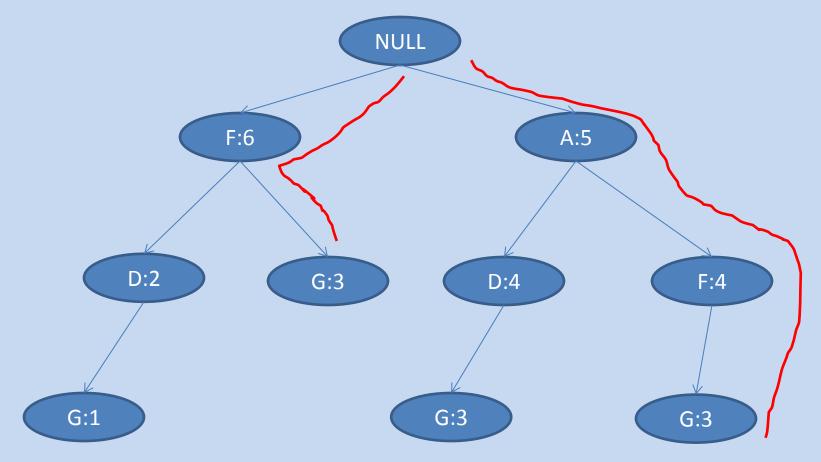
These are all paths ending in G. Clearly, any paths ending in PG must be drawn from these for different pre-fixes I.

Example: Prefix Path Subtree for DG



These are all paths ending in G. Clearly, any paths ending in PG must be drawn from these for different pre-fixes I.

Example: Prefix Path Subtree for FG

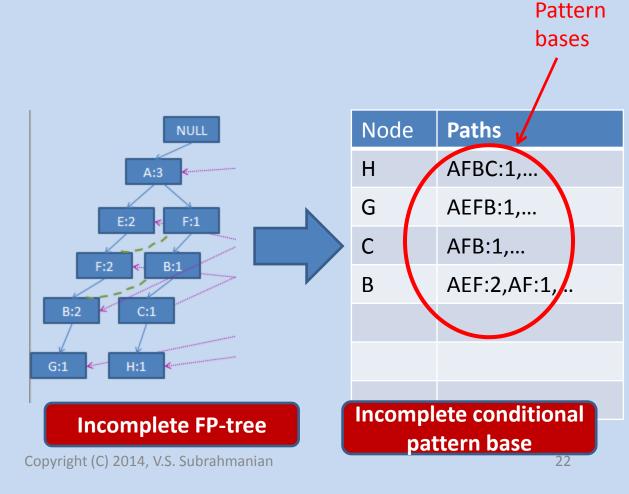


These are all paths ending in G. Clearly, any paths ending in PG must be drawn from these for different pre-fixes I.

Conditional FP-Tree: Step 2

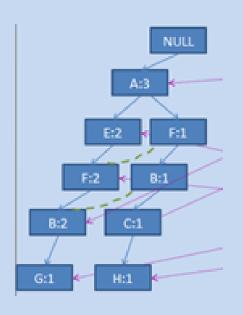
Consider this situation again.

Item	Count
Α	7
E	7
F	7
В	6
С	5
G	5
Н	5
₽	4



Conditional FP-Tree: Step 2

- Condition FP-tree for H:
- A:1->F:1->B:1->C:1



Item	Count	Node	Paths
Α	7	Н	AFBC:1,
E	7	G	AEFB:1,
F	7	С	AFB:1,
В	6	В	AEF:2,AF:1,
С	5		
G	5		
Н	5		

Reference

Also consult slides from: Mining Frequent Patterns without Candidate Generation

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