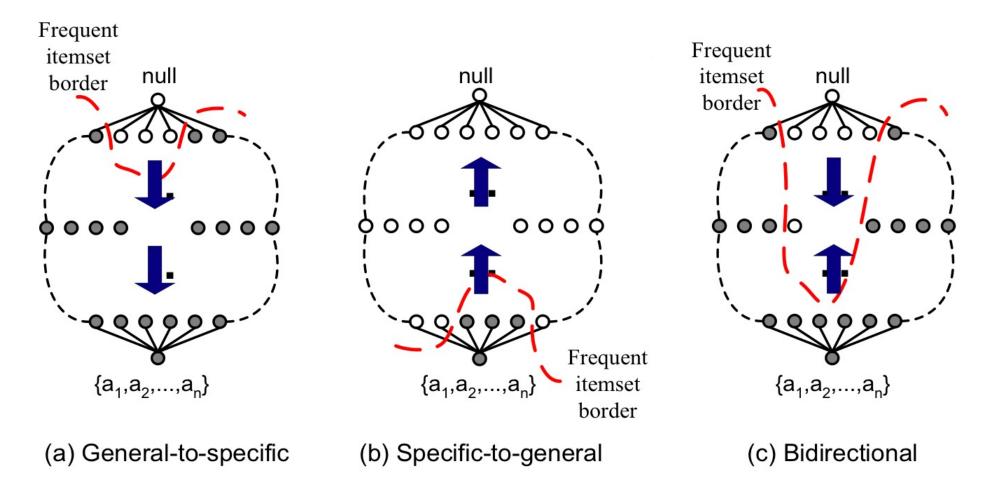
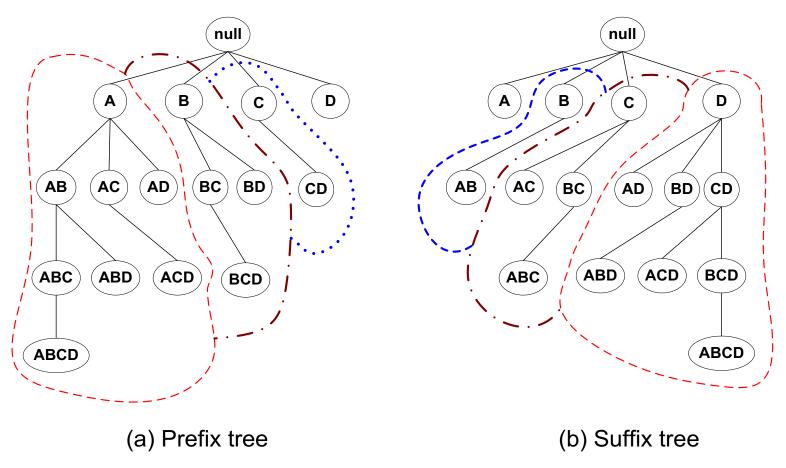
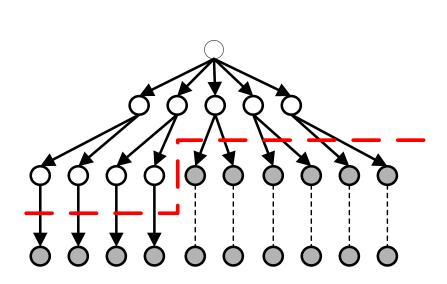
- Traversal of Itemset Lattice
  - General-to-specific vs Specific-to-general



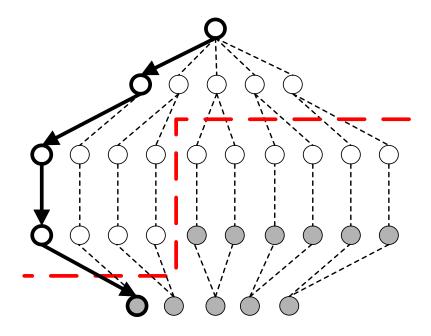
- Traversal of Itemset Lattice
  - Equivalent Classes



- Traversal of Itemset Lattice
  - Breadth-first vs Depth-first



(a) Breadth first



(b) Depth first

- Representation of Database
  - horizontal vs vertical data layout

Horizontal Data Layout

TID	Items
1	A,B,E
2	B,C,D
3	C,E
4	A,C,D
5	A,B,C,D
6	A,E
7	A,B
8	A,B,C
9	A,C,D
10	В

Vertical Data Layout

Α	В	С	D	Е
1	1	2	2	1
4	2	3	4	3 6
4 5 6 7	2 5 7	2 3 4 8 9	2 4 5 9	6
6	7	8	9	
	8	9		
8 9	10			
9				

# **FP-growth Algorithm**

 Use a compressed representation of the database using an FP-tree

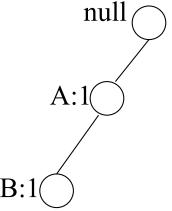
 Once an FP-tree has been constructed, it uses a recursive divide-and-conquer approach to mine the frequent itemsets

## **FP-tree construction**

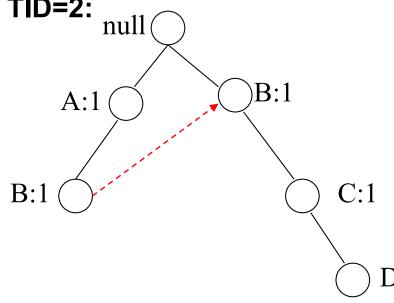
## Maps each transaction onto a path in the FP-tree

TID	Items	
1	{A,B}	
2	{B,C,D}	
3	$\{A,C,D,E\}$	
4	{A,D,E}	
5	$\{A,B,C\}$	
6	$\{A,B,C,D\}$	
7	{B,C}	
8	$\{A,B,C\}$	
9	$\{A,B,D\}$	
10	$\{B,C,E\}$	

After reading TID=1:



**After reading TID=2:** 



Frequent 1-itemsets are sorted in decreasing support counts.

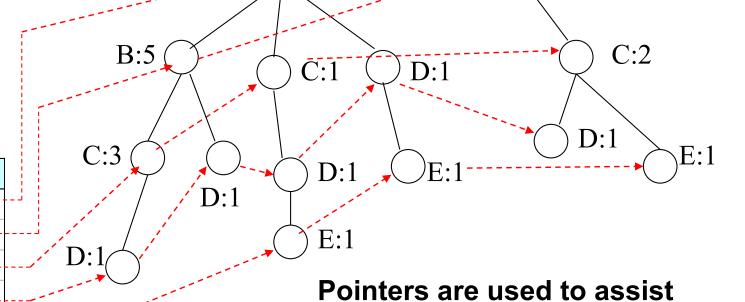
## **FP-Tree Construction**

TID	Items		
1	{A,B}		
2	{B,C,D}		
3	$\{A,C,D,E\}$		
4	$\{A,D,E\}$		
5	$\{A,B,C\}$		
6	{A,B,C,D}		
7	{A}		
8	{A,B,C}		
9	$\{A,B,D\}$		
10	{B,C,E}		

# Transaction Database

#### **Header table**

Item	Pointer
Α	
В	
С	
D	
Ε	



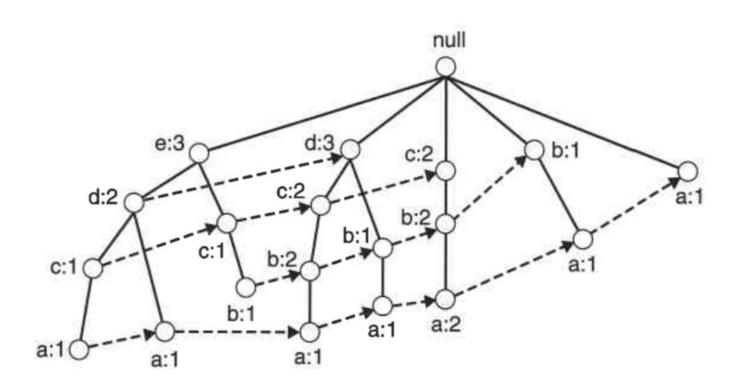
B:2

frequent itemset generation

null

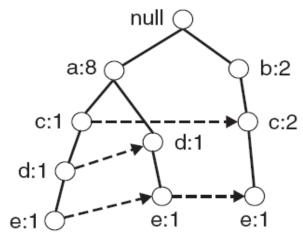
A:8

## **FP-Tree Construction**

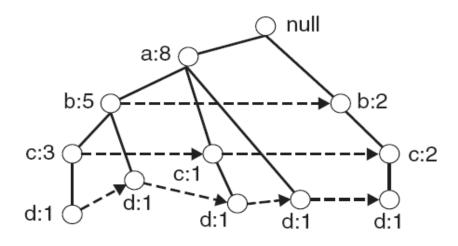


The ordering scheme is from lowest to highest support item.

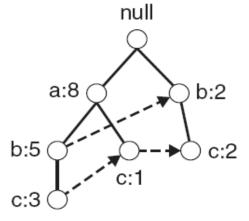
FP-growth generates frequent itemsets from a FP-tree by exploring the tree in a bottom-up fashion.



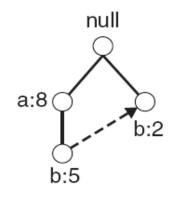
(a) Paths containing node e



(b) Paths containing node d



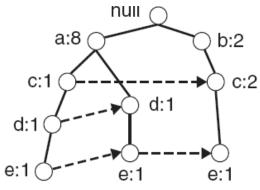
(c) Paths containing node c





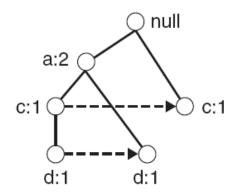
- (d) Paths containing node b
- (e) Paths containing node a

# Finding all the frequent itemsets ending with a particular suffix by employing a divide-and-conquer strategy



(a) Prefix paths ending in e

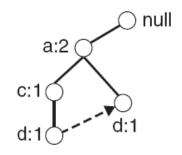
Minimum support count is 2 {e} is a frequent itemset



(b) Conditional FP-tree for e

Conditional FP-tree for e It is used to find frequent itemsets ending in de,ce,ae

- Update the support counts
- Remove e
- Remove infrequent items



(c) Prefix paths ending in de

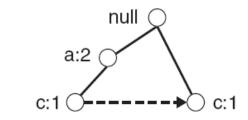
{d,e} is a frequent itemset



(d) Conditional FP-tree for de

Construct a conditional FP-tree for de

{a,d,e} is a frequent itemset



(e) Prefix paths ending in ce

{c,e} is a frequent itemset



(f) Prefix paths ending in ae

{a,e} is a frequent itemset

# **Summary**

- Association rule mining:
  - Frequent itemset generation
    - Evaluation metrics: support, count,...
  - Rule generation
- Frequent itemset generation algorithms:
  - Apriori
    - $F_{k-1}F_1, F_{k-1}F_{k-1}$
  - FP-growth
- Support count:
  - Hash tree, FP-tree
- Frequent itemsets representations:
  - Maximal/closed frequent itemsets