

The background features a complex geometric pattern of thin, light-colored lines forming a network of triangles and polygons. Overlaid on this are several semi-transparent rectangular panels. The top-left panel shows a grid of small, colorful dots (green, blue, orange) connected by lines. The bottom-left panel displays a dense cluster of orange and red dots. The right side of the image has a vertical strip of white lines. A large, white, angular shape serves as a backdrop for the central text.

# **FPGrowth: A Pattern Growth Approach**

# FPGrowth: Mining Frequent Patterns by Pattern Growth

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- ❑ Idea: Frequent pattern growth (FPGrowth)
  - ❑ Find frequent single items and partition the database based on each such item
  - ❑ Recursively grow frequent patterns by doing the above for each partitioned database (also called *conditional database*)
  - ❑ To facilitate efficient processing, an efficient data structure, FP-tree, can be constructed
- ❑ Mining becomes
  - ❑ Recursively construct and mine (conditional) FP-trees
  - ❑ Until the resulting FP-tree is empty, or until it contains only one path—single path will generate all the combinations of its sub-paths, each of which is a frequent pattern

# Example: Construct FP-tree from a Transactional DB

TID	Items in the Transaction	Ordered, frequent items
100	{f, a, c, d, g, i, m, p}	{f, c, a, m, p}
200	{a, b, c, f, l, m, o}	{f, c, a, b, m}
300	{b, f, h, j, o, w}	{f, b}
400	{b, c, k, s, p}	{c, b, p}
500	{a, f, c, e, l, p, m, n}	{f, c, a, m, p}

1. Scan DB once, find single item frequent pattern: **Let min\_support = 3**

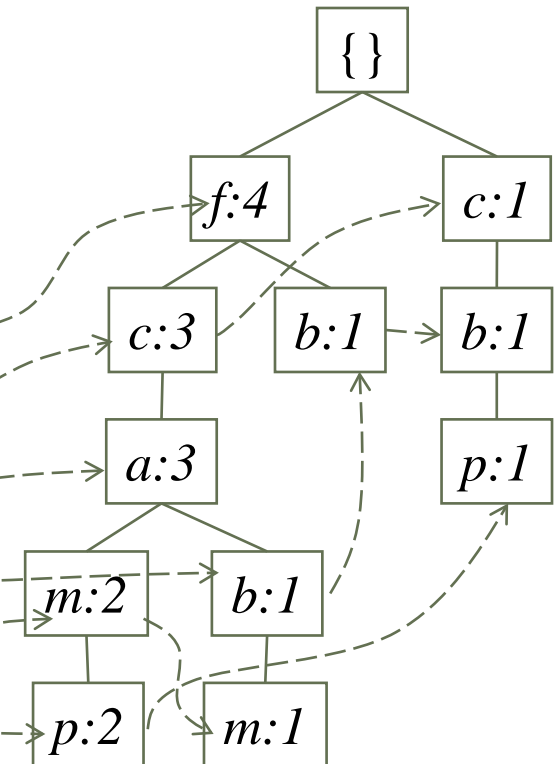
f:4, a:3, c:4, b:3, m:3, p:3

2. Sort frequent items in frequency descending order, f-list

**F-list** = f-c-a-b-m-p

3. Scan DB again, construct FP-tree

Header Table		
Item	Frequency	header
f	4	
c	4	
a	3	
b	3	
m	3	
p	3	

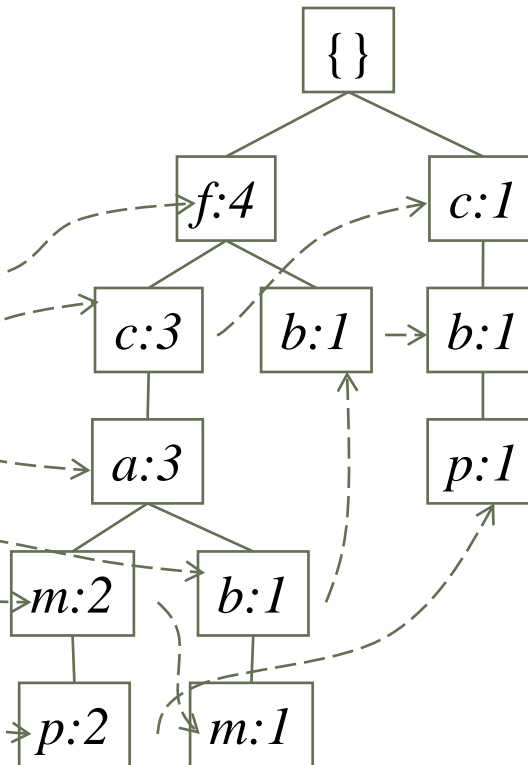


# Divide and Conquer Based on Patterns and Data

- Pattern mining can be partitioned according to current patterns
  - Patterns containing p: p's conditional database:  $fcam:2, cb:1$
  - Patterns having m but no p: m's conditional database:  $fca:2, fcab:1$
  - ..... .....
- p's conditional pattern base: *transformed prefix paths* of item p

min\_support = 3

Item	Frequency	Header
f	4	
c	4	
a	3	
b	3	
m	3	
p	3	



## Conditional pattern bases

Item	Conditional pattern base
c	$f:3$
a	$fc:3$
b	$fca:1, f:1, c:1$
m	$fca:2, fcab:1$
p	$fcam:2, cb:1$

# Mine Each Conditional Pattern-Base Recursively

## Conditional pattern bases

<i>item</i>	<i>cond. pattern base</i>
<i>c</i>	<i>f:3</i> <b>min_support = 3</b>
<i>a</i>	<i>fc:3</i>
<i>b</i>	<i>fca:1, f:1, c:1</i>
<i>m</i>	<i>fca:2, fcab:1</i>
<i>p</i>	<i>fcam:2, cb:1</i>

□ For each conditional pattern-base

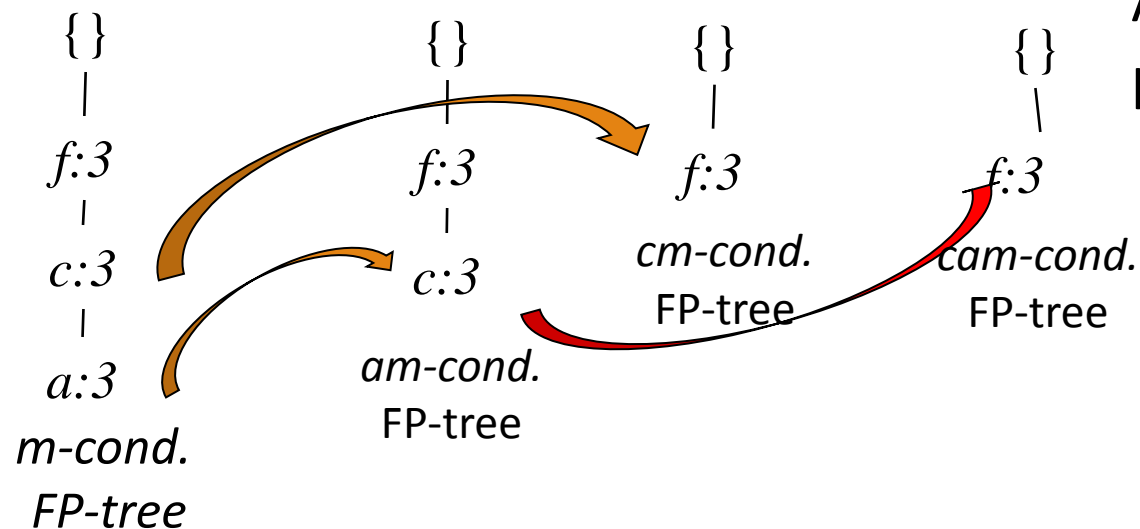
□ Mine single-item patterns

□ Construct its FP-tree & mine it

*p*-conditional PB: *fcam:2, cb:1* → *c: 3*

*m*-conditional PB: *fca:2, fcab:1* → *fca: 3*

*b*-conditional PB: *fca:1, f:1, c:1* →  $\phi$



Actually, for single branch FP-tree, all frequent patterns can be generated in one shot

*m: 3*

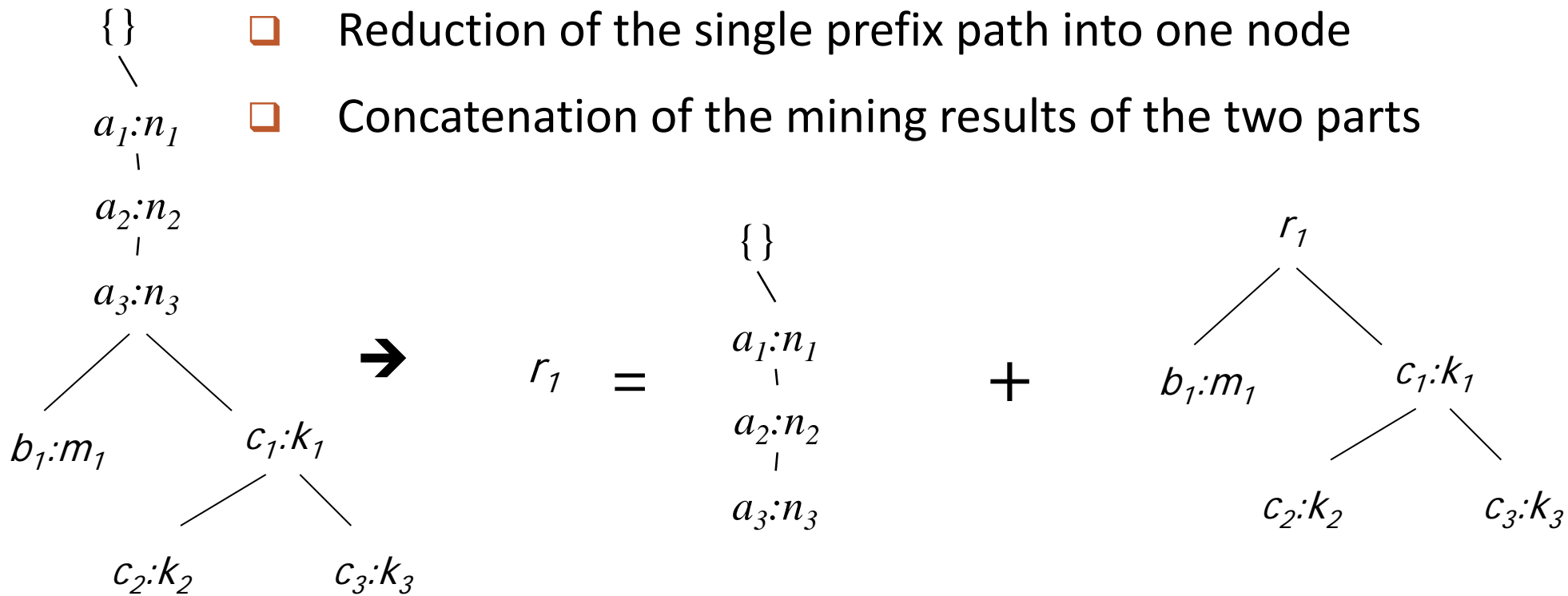
*fm: 3, cm: 3, am: 3*

*fcm: 3, fam:3, cam: 3*

*fcam: 3*

# A Special Case: Single Prefix Path in FP-tree

- Suppose a (conditional) FP-tree  $T$  has a shared single prefix-path  $P$
- Mining can be decomposed into two parts



# Scaling FP-growth by Database Projection

- ❑ What if FP-tree cannot fit in memory? — DB projection
  - ❑ Project the DB based on patterns
  - ❑ Construct & mine FP-tree for each projected DB
- ❑ **Parallel projection** vs. **partition projection**
  - ❑ Parallel projection: Project the DB on each frequent item
    - ❑ Space costly, all partitions can be processed in parallel
  - ❑ Partition projection: Partition the DB in order
    - ❑ Passing the unprocessed parts to subsequent partitions

