

FP Tree

- It eliminates both the drawbacks of Apriori algo.
- It is used to ~~find~~ find frequent itemsets L_k without generating candidate sets C_k and it requires only two database scans.
- This method ~~transforms~~ transforms the database of transactions into a tree structure called FP (frequent pattern) tree such that association information betⁿ the data items ~~are~~ is preserved.
- The method has following steps.
 - (i) Scan the db once & find set of frequent 1-item sets L_1 .
 - (ii) Arrange the frequent items in descending order of ~~pro~~ support count.
 - (iii) Scan the db again & construct FP tree.
 - (iv) For each node in FP tree, construct conditional pattern base.
 - (v) For each conditional pattern base construct conditional FP tree.
 - (vi) Mine conditional FP trees recursively to grow frequent patterns.

Consider the following set of x'actions.

Tid items

T₁ i₁ i₂ i₅

T₂ i₂ i₄

T₃ i₂ i₃

T₄ i₁ i₂ i₄

T₅ i₁ i₃

T₆ i₂ i₃

T₇ i₁ i₃

T₈ i₁ i₂ i₃ i₅

T₉ i₁ i₂ i₃

let min-support = 2

step 1: Scan the db & find L₁

C₁ = { {i₁} {i₂} {i₃} {i₄} {i₅} }

SC = 6 7 6 2 2

∴ L₁ = { {i₁} {i₂} {i₃} {i₄} {i₅} }

step 2: Arrange the frequent items in descending order of SC

∴ i₂ → 7, i₁ → 6, i₃ → 6, i₄ → 2, i₅ → 2

Now the frequent items in each x'action must be processed in this order only. For this add a column "ordered frequent items" in the set of x'actions.

Tid	items	ordered frequent items
T ₁	i ₁ i ₂ i ₅	i ₂ i ₁ i ₅
T ₂	i ₂ i ₄	i ₂ i ₄
T ₃	i ₂ i ₃	i ₂ i ₃
T ₄	i ₁ i ₂ i ₄	i ₂ i ₁ i ₄
T ₅	i ₁ i ₃	i ₁ i ₃
T ₆	i ₂ i ₃	i ₂ i ₃
T ₇	i ₁ i ₃	i ₁ i ₃
T ₈	i ₁ i ₂ i ₃ i ₅	i ₂ i ₁ i ₃ i ₅
T ₉	i ₁ i ₂ i ₃	i ₂ i ₁ i ₃

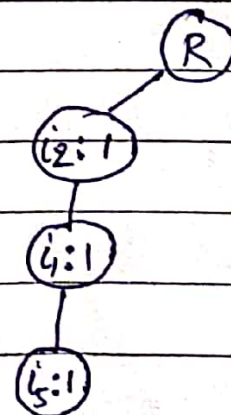
Step III Scan the DB again (the 3rd col) & construct FP tree.

- The root of FP tree is empty. All other nodes contain an item & its occurrence frequency with the items in ~~the~~ predecessor nodes.

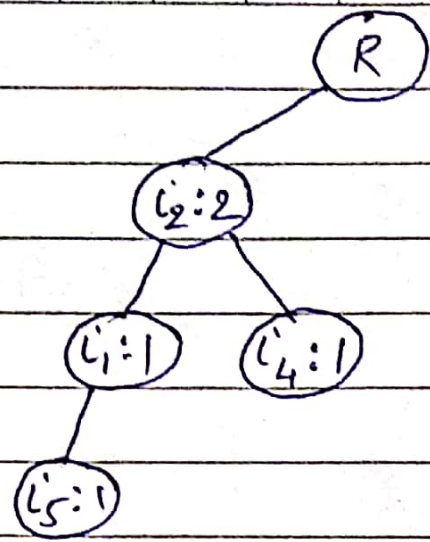
- create empty root node.

(R)

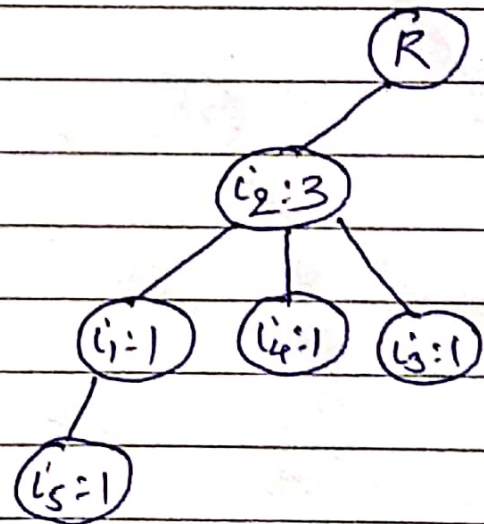
- Read the 1st transaction (3rd col) & create a path in tree



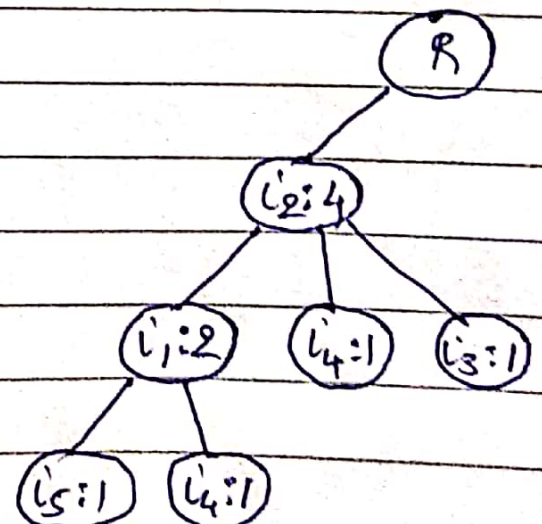
- read the 2nd action $\{i_2, i_4\}$. (3rd col)
 as i_2 occurs under root node, increment its counter to 2,
 it is not followed by i_4 in the tree,
 so create a new path.



- read the 3rd action $\{i_2, i_3\}$.
 as i_2 occurs under root node, increment its counter to 3.
 It is not followed by i_3 in the tree
 so create a new path



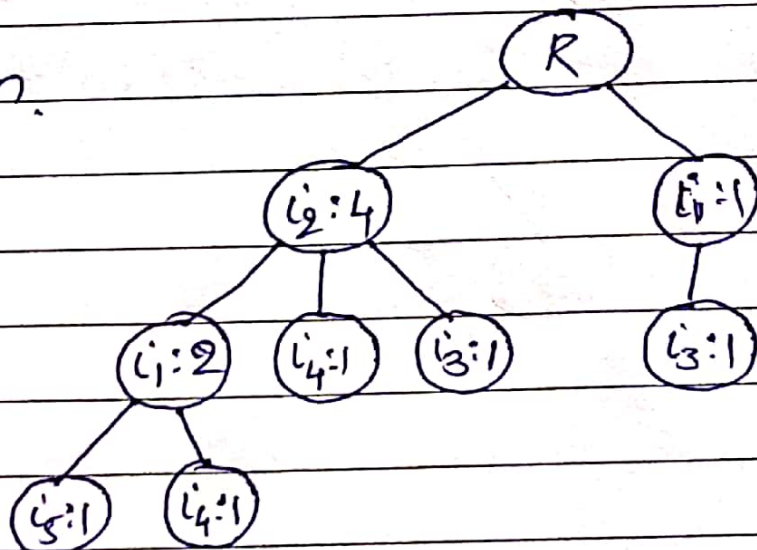
- read the 4th action $\{i_2, i_1, i_4\}$
 i_2 occurs under root node, increment its counter to 4.
 It is followed by i_1 ,
 so increment counter to 2.
 It is not followed by i_4 ,
 so create a new path.



- Read the 5th action.

$\{i_1, i_3\}$

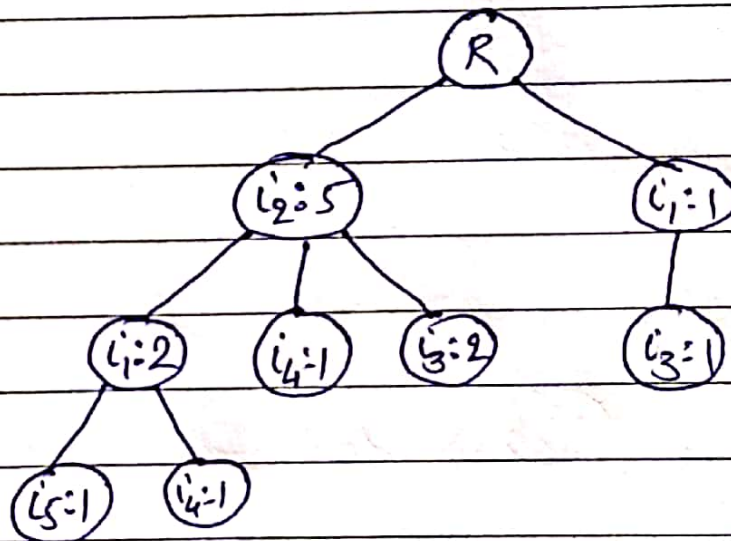
as i_1 doesn't occur under root node, create a new path.



- Read 6th action

$\{i_2, i_3\}$

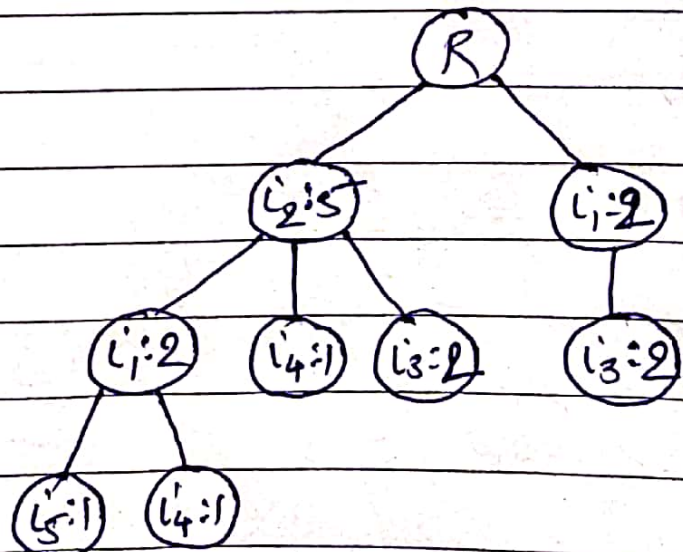
as i_2 occurs under root node & it is followed by i_3 in one of the paths, increment counters in their respective nodes.



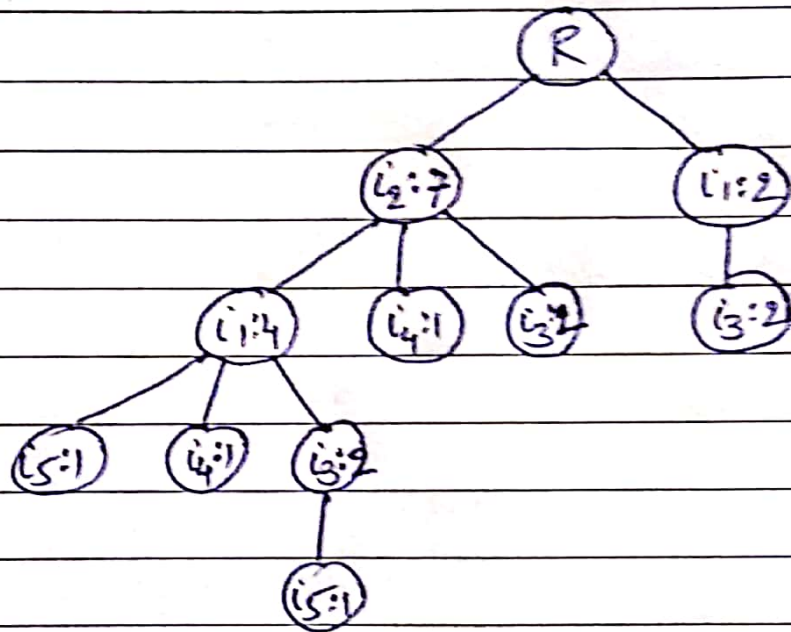
- Read the 7th action

$\{i_1, i_3\}$

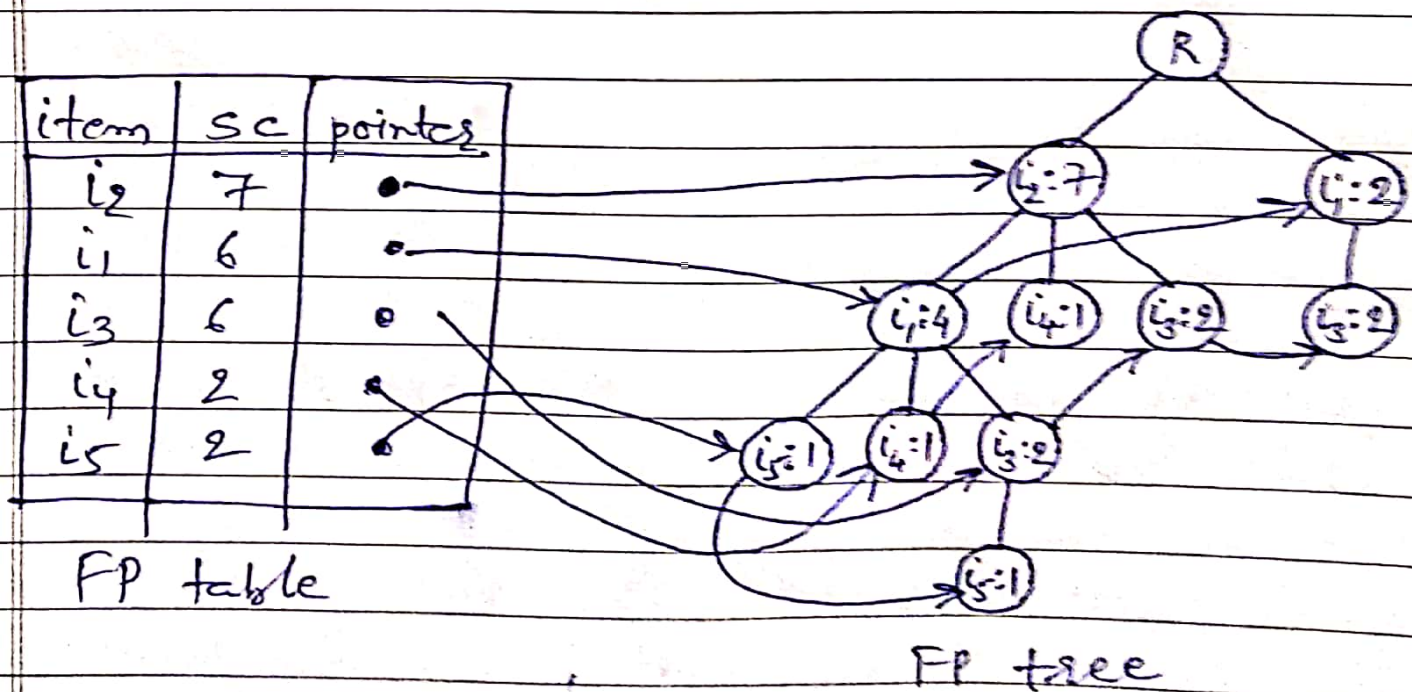
as i_1 occurs under root node & it is followed by i_3 , increment counters in their respective nodes.



- after processing x'actions TS & TR, we get the following tree

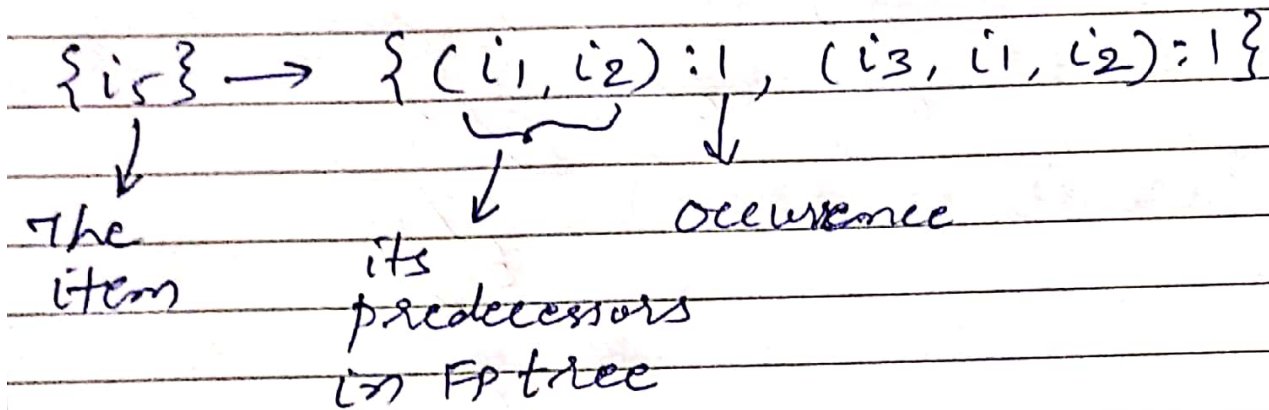


- To process, this tree, an FP table is linked to it.



step IV construct conditional pattern base

- use a pointer from FP tree & write the item's occurrence with its predecessors.



$$\{i_4\} \rightarrow \{(i_1, i_2):1, (i_2):1\}$$

$$\{i_3\} \rightarrow \{(i_1, i_2):2, (i_2):2, (i_1):2\}$$

$$\{i_1\} \rightarrow \{(i_2):4\}$$

$$\{i_2\} \rightarrow \emptyset$$

step V construct conditional FP tree

- It is basically accumulating the number of occurrences of an item with each of its predecessors in the conditional pattern base

eg $\{i_5\}$ has occurred with i_1 twice
once in (i_1, i_2) & once in (i_3, i_1, i_2)
Similarly, it has occurred with i_2 twice,
with (i_1, i_2) also twice & with i_3 once &
 (i_3, i_1, i_2) once.

∴ it can be written as,

$$\{i_5\} \rightarrow \{(i_1:2), (i_2:2), (i_1, i_2:2), \\ (i_3:1), (i_3, i_1:1), (i_3, i_2:1), \\ (i_3, i_1, i_2:1)\}$$

Now discard the occurrences that don't satisfy the min-support (2, as given)

$$\therefore \{i_5\} \rightarrow \{(i_1:2), (i_2:2), (i_1, i_2:2)\}$$

Similarly for other items

$$\{i_4\} \rightarrow \{(i_2:2)\}$$

$$\{i_3\} \rightarrow \{(i_1:4), (i_2:4), (i_1, i_2:2)\}$$

$$\{i_1\} \rightarrow \{(i_2:4)\}$$

$$\{i_2\} \rightarrow \emptyset$$

step VI find frequent-patterns
add the item with occurrences of its predecessors.

$$\therefore \{i_5\} \rightarrow \{(i_1, i_5:2), (i_2, i_5:2), (i_1, i_2, i_5:2)\}$$

↓
member
of L_2

↓
member
of L_2

↓
member
of L_3

Similarly for other items.

$$\{i_4\} \rightarrow \{(i_2, i_4: 2)\}$$

$\hookrightarrow L_2$

$$\{i_3\} \rightarrow \{(i_1, i_3: 4), (i_2, i_3: 4), (i_1, i_2, i_3: 2)\}$$

$\hookrightarrow L_2 \quad \quad \quad \hookrightarrow L_2 \quad \quad \quad \hookrightarrow L_3$

$$\{i_1\} \rightarrow \{(i_2, i_1: 4)\}$$

$\hookrightarrow L_2$

$$\therefore L_2 = \{\{i_1, i_5\}, \{i_2, i_5\}, \{i_2, i_4\}, \{i_1, i_3\}, \{i_2, i_3\}, \{i_2, i_1\}\}$$

$$L_3 = \{\{i_1, i_2, i_5\}, \{i_1, i_2, i_3\}\}$$

- Unlike Apriori, it is not level wise & it doesn't generate C_k .