

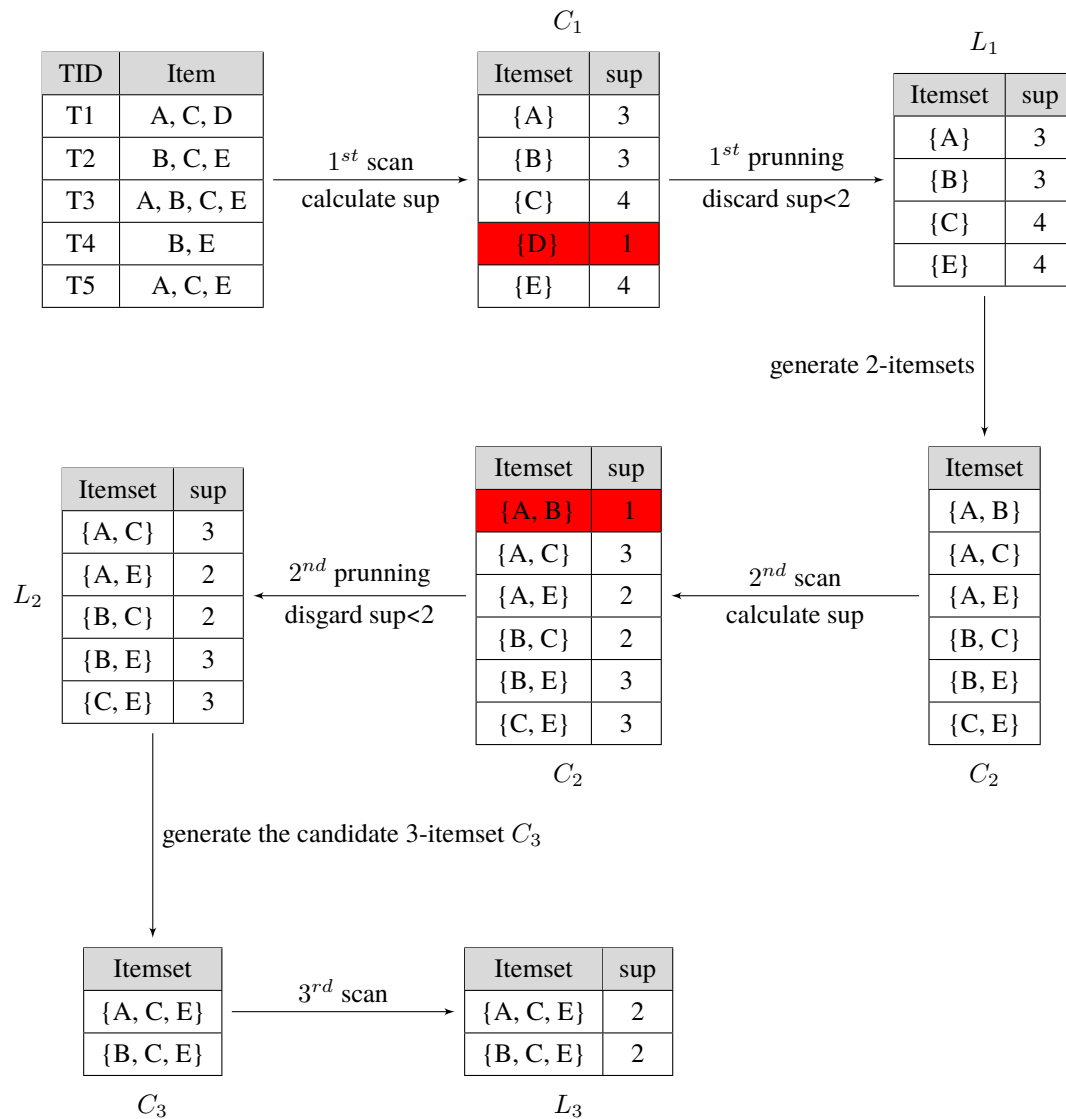
DSAA 5002 - HW1

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Q1 [15 Marks]

Given the transaction database below, set the minimum support count to 2 and the minimum confidence level to 60% to find the strong association rule. Generate the set C_3 of the candidate 3-itemset, using pruning on Apriori principle.

Solution:



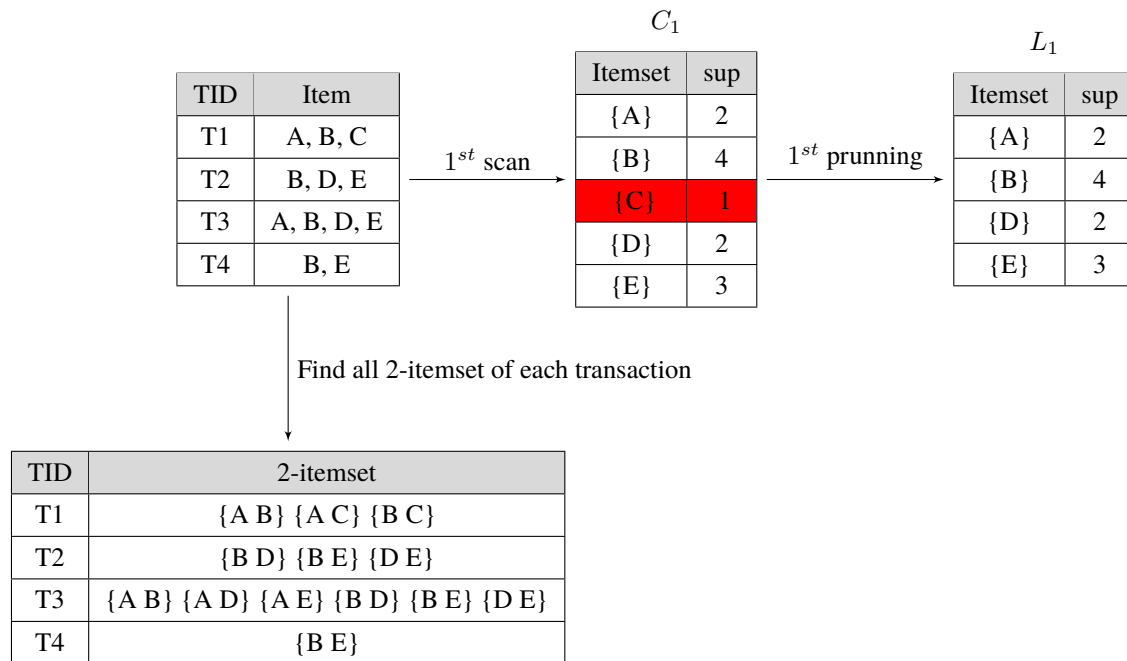
Thus we get the candidate 3-itemset C_3 .

Q2 [15 Marks]

Reducing the transactions using dynamic hashing and pruning(DHP) algorithm. Set the minimum support count to 2.

Hash function bucket # = $h(\{xy\}) = ((\text{order of } x) * 10 + (\text{order of } y)) \% 7$

Solution



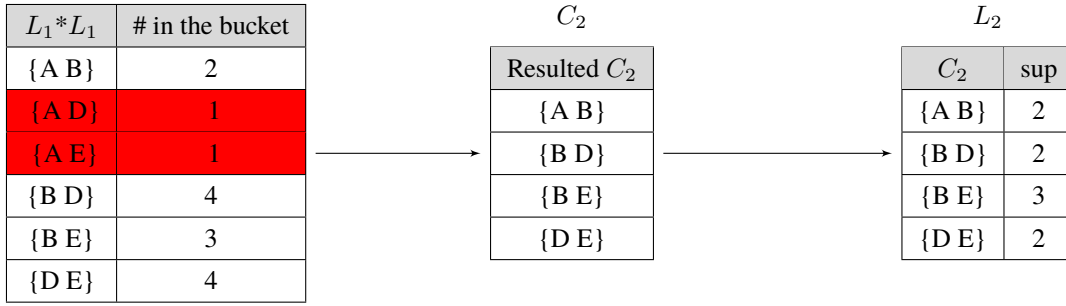
Because we have:

- Items = A, B, C, D, E,
- Order = 1, 2, 3, 4, 5
- Hash function : $h(\{xy\}) = ((\text{order of } x) * 10 + (\text{order of } y)) \% 7$

Thus we have the hash table below:

bucket	0	1	2	3	4	5	6
count	1	1	1	4	3	2	1
2-itemset	{A D}	{A E}	{B C}	{B D} {D E} {B D} {D E}	{B E} {B E} {B E}	{A B} {A B}	{A C}

Let $L_1 * L_1$ to generate a 2-itemset table, and choose the itemsets where the number of content in its bucket is above the minimum support.



Because if an item occurs in a frequent $(k+1)$ -itemset, it must occur in at least k candidate k -itemsets.

TID	Item	2-itemset occurs	
T1	A, B, C	{A B}	Disgard
T2	B, D, E	{B D} {B E} {D E}	Keep {B D E}
T3	A, B, D, E	{A B} {B D} {B E} {D E}	Keep {B D E}
T4	B, E	{E E}	Disacrd

\longrightarrow

TID	Item
T2	B, D, E
T3	B, D, E

Thus we have reduced the transactions.

Q3 [35 Marks]

An itemset X is said to be a frequent itemset if the frequency count of X is at least a given support threshold.

An itemset Y is a proper super-itemset of X if $X \subset Y$ and $X \neq Y$.

An itemset X is said to be a closed frequent itemset if (1) X is frequent and (2) there exists no proper super-itemset Y of X such that Y is frequent and Y has the same frequency count as X .

An itemset X is said to be a maximal frequent itemset if (1) X is frequent and (2) there exists no proper super-itemset Y of X such that Y is frequent.

Let F_c be the set of (traditional) frequent itemsets each of which is associated with a frequency in the dataset.

For example, if there are three frequent itemsets, $\{I_1\}$ with frequency 4, $\{I_2\}$ with frequency 5, and $\{I_1, I_2\}$ with frequency 3, $F = \{\{I_1\}, \{I_2\}, \{I_1, I_2\}\}$ and $F_c = \{<\{I_1\}, 4>, <\{I_2\}, 5>, <\{I_1, I_2\}, 3>\}$.

Similarly, let C be the set of closed frequent itemsets without specifying the frequency of itemsets.

Let C_c be the set of closed frequent itemsets each of which is associated with a frequency of itemsets.

Let M be the set of maximal frequent itemsets without specifying the frequency of itemsets.

Ler M_c be the set of maximal frequent itemsets each of which is associated with a frequency in the dataset.

The following shows six transactions with four items. Each row corresponds to a transaction where 1 corresponds to a presence of an item and 0 corresponds to an absence.

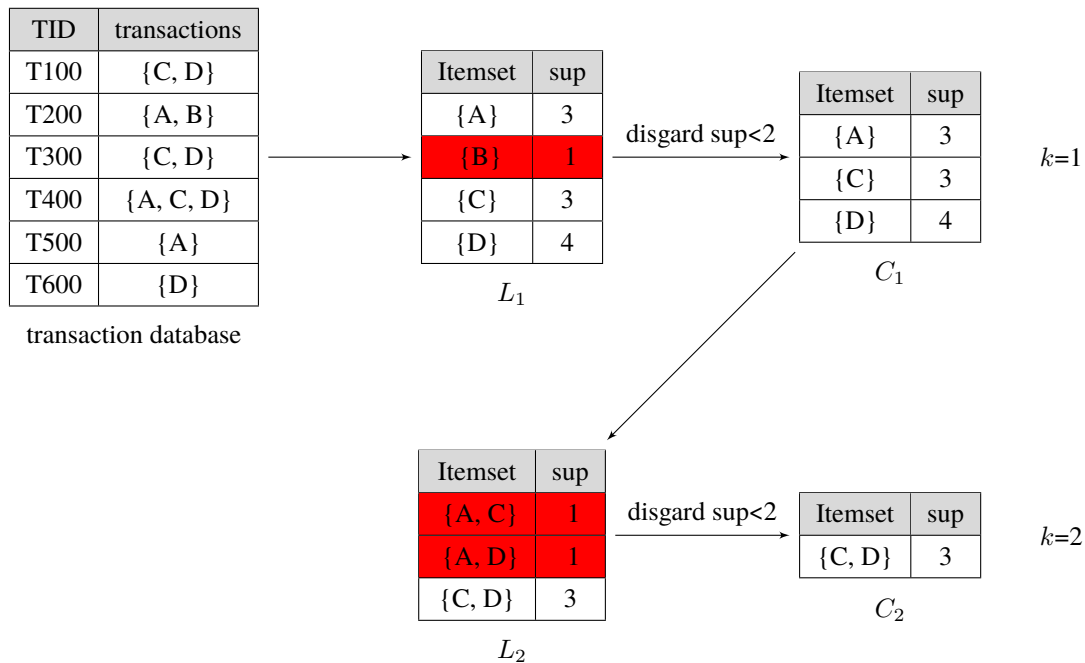
A	B	C	D
0	0	1	1
1	1	0	0
0	0	1	1
1	0	1	1
1	0	0	0
0	0	0	1

Suppose that the support threshold is 2.

- (a) (i) What is F_c ? (ii) What is C_c ? (iii) What is M_c ? (5 Marks)
- (b) (i) What are the advantages and the disadvantages of using closed frequent itemsets compared with traditional frequent itemsets? (5 Marks)
- (ii) What are the advantages and the disadvantages of using closed frequent itemsets compared with maximal frequent itemsets? (5 Marks)
- (c) Please adapt algorithm FP-growth with the use of the FP-tree to find all closed frequent item set. Please write down how to adapt algorithm FP-growth and illustrate the adapted algorithm with the above example. (20 Marks)

Solution

- (a) According to the topic, we have the following transaction database. And we generate all the k -itemsets which might be frequent itemsets.



i: We have $F_c = \{ \langle \{A\}, 3 \rangle, \langle \{C\}, 3 \rangle, \langle \{D\}, 4 \rangle, \langle \{C, D\}, 3 \rangle \}$

ii: We have $C_c = \{ \langle \{A\}, 3 \rangle, \langle \{C, D\}, 3 \rangle \}$

iii: We have $M_c = \{ \langle \{A\}, 3 \rangle, \langle \{C, D\}, 3 \rangle \}$

(b)

(c)

Q4 [35 Marks]

A GSP example: Suppose now we have 5 events: 'Upload Songs', 'Add Tags', 'Share', 'Listen' and 'Comment'. Let min-support be 40%. The sequence database of a Music Platform is shown in following table:

Object	Sequence
A	<{'Upload Songs', 'Add Tags'}>
B	<{'Upload Songs', 'Share'}>
C	<{'Upload Songs'}, {'Share', 'Listen'}>
D	<{'Upload Songs'}, {'Upload Songs', 'Add Tags'}, {'Listen'}>
E	<{'Listen'}, {'Add Tags', 'Comment'}, {'Share', 'Listen'}>

Please answer the following questions:

(a) Make the first pass over the sequence database to yield all the 1-element **frequent** sequences and What is the corresponding support? **5 Marks**

(b) Based on (a), do the 2-sequences Candidate Generation and Candidate Pruning. **10 Marks**

(c) What is the **frequent** 2-sequences based on the result of (b)? **5 Marks**

(d) Based on (c), do the 3-sequences Candidate Generation and Candidate Pruning. When a sequence should be pruned, you need to explain why. **10 Marks**

(e) What is the frequent 3-sequences based on the result of (d)? Please calculate the support. **5 Marks**

Remember: For frequent k-sequences, the support \geq min-support

Solution