# **APRIORI ALGORITHM**







# International School of Engineering

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## **DEFINITION OF APRIORI ALGORITHM**

- The Apriori Algorithm is an influential algorithm for mining frequent itemsets for boolean association rules.
- Apriori uses a "bottom up" approach, where frequent subsets are extended one item at a time (a step known as *candidate generation*, and groups of candidates are tested against the data.
- Apriori is designed to operate on database containing transactions (for example, collections of items bought by customers, or details of a website frequentation).







## **KEY CONCEPTS**

• Frequent Itemsets: All the sets which contain the item with the minimum support (denoted by  $L_i$  for  $i^{th}$  itemset).





Apriori Property: Any subset of frequent itemset must be frequent.

• Join Operation: To find  $L_k$ , a set of candidate k-itemsets is generated by joining  $L_{k-1}$  with itself.



## STEPS TO PERFORM APRIORI ALGORITHM

#### STEP 1

Scan the transaction data base to get the support of S each 1-itemset, compare S with min\_sup, and get a support of 1-itemsets, L1

#### STEP 2

Use  $L_{k-1}$  join  $L_{k-1}$  to generate a set of candidate k-itemsets. And use Apriori property to prune the unfrequented k-itemsets from this set.

#### STEP 6

For every nonempty subset s of 1, output the rule "s=>(1-s)" if confidence C of the rule "s=>(1-s)" (=support s of 1/support S of s)' min\_conf

#### STEP 3

Scan the transaction database to get the support S of each candidate k-itemset in the find set, compare S with min\_sup, and get a set of frequent k-itemsets  $L_k$ 

#### STEP 4

The candidate set = Null

YES

#### STEP 5

For each frequent itemset 1, generate all nonempty subsets of 1







# APRIORI ALGORITHM EXAMPLE

## Market basket









### MARKET BASKET ANALYSIS

- Provides insight into which products tend to be purchased together and which are most amenable to promotion.
- Actionable rules
- Trivial rules
  - People who buy chalk-piece also buy duster
- Inexplicable
  - People who buy mobile also buy bag







# APRIORI ALGORITHM EXAMPLE

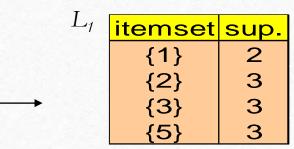
Database D

Minsup = 0.5

Items		
5		

Scan D

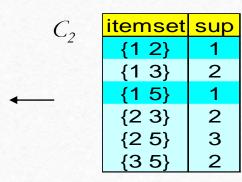
	itemset	sup.
Š	{1}	2
	{2}	3
	{3}	3
	<b>{4</b> }	1
	<b>{5</b> }	3

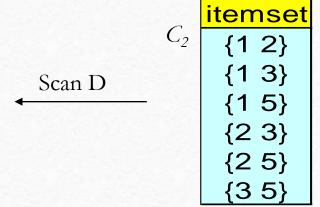






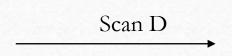
$L_2$	itemset su		
	{1 3}	2	
	{2 3}	2	
	{2 5}	3	
	{3 5}	2	











$L_3$	itemset	sup
	{2 3 5}	2





# The Apriori Algorithm: Pseudo Code

- Join Step:  $C_k$  is generated by joining  $L_{k-1}$  with itself
- Prune Step: Any (k-1)-itemset that is not frequent cannot be a subset of a frequent k-itemset
- Pseudo-code : $C_k$ : Candidate itemset of size k

 $L_k$ : frequent itemset of size k

```
L_1 = {frequent items};

for (k = 1; L_k \mid = \emptyset; k++) do begin

C_{k+1} = candidates generated from L_k;

for each transaction t in database do

increment the count of all candidates in C_{k+1}

that are contained in t

L_{k+1} = candidates in C_{k+1} with min_support

end

return \bigcup_k L_k;
```





## **LIMITATIONS**

• Apriori algorithm can be very slow and the bottleneck is candidate generation.

For example, if the transaction DB has 10<sup>4</sup> frequent 1-itemsets, they will generate 10<sup>7</sup> candidate 2-itemsets even after employing the downward closure.



• To compute those with sup more than min sup, the database need to be scanned at every level. It needs (n + 1) scans, where n is the length of the longest pattern.





## METHODS TO IMPROVE APRIORI'S EFFICIENCY

- Hash-based itemset counting: A k-itemset whose corresponding hashing bucket count is below the threshold cannot be frequent
- Transaction reduction: A transaction that does not contain any frequent k-itemset is useless in subsequent scans
- Partitioning: Any itemset that is potentially frequent in DB must be frequent in at least one of the partitions of DB.
- Sampling: mining on a subset of given data, lower support threshold + a method to determine the completeness
- Dynamic itemset counting: add new candidate itemsets only when all of their subsets are estimated to be frequent







# APRIORI ADVANTAGES/DISADVANTAGES

- Advantages
  - Uses large itemset property
  - Easily parallelized
  - Easy to implement
- Disadvantages
  - Assumes transaction database is memory resident.
  - Requires many database scans

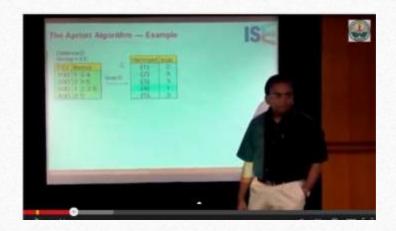






# For Detailed Description of APRIORI ALGORITHM

Check out our video on

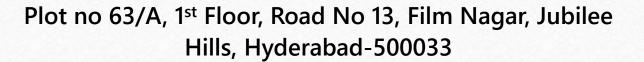






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