# Mining Frequent Itemsets 

CPS 296.1
Topics in Database Systems

## Mining frequent itemsets

- Given: a large database of transactions, each containing a set of items
- Example: market baskets
- Find all frequent itemsets
- A set of items $X$ is frequent if no less than $s_{\min } \%$ of all transactions contain $X$

- Examples: \{diaper, beer\}, \{scanner, color printer\}


## The Apriori property

- All subsets of a frequent itemset must also be frequent
- Because any transaction that contains $X$ must also contains subsets of $X$

If we have already verified that $X$ is infrequent, there is no need to count $X$ 's supersets because they must be infrequent too

## Data mining

- Data $\rightarrow$ knowledge
- DBMS meets AI and statistics
- Clustering, prediction (classification and regression), association analysis, outlier analysis, evolution analysis, etc.
> Usually complex statistical "queries" that are difficult to answer $\rightarrow$ often specialized algorithms outside DBMS
$>$ We will focus on papers related to association rule/frequent itemset mining


## A naïve algorithm

- First try
- Keep a running count for each possible itemset
- For each transaction $T$, and for each itemset $X$, if $T$ contains $X$ then increment the count for $X$
- Return itemsets with large enough counts
- Problem: The number of itemsets is huge!
$-2^{n}$, where $n$ is the number of items
- Think: How do we prune the search space?


## The Apriori algorithm

Agrawal and Srikant. "Fast Algorithms for Mining Association Rules." VLDB 1994

- Multiple passes over the transactions
- Pass $k$ finds all frequent $k$-itemsets (itemset of size $k$ )
- Use the set of frequent $(k-1)$-itemsets found in the previous pass to narrow the search for $k$ itemsets


## Pseudo-code for Apriori

Scan the transactions to find $L_{1}$, the set of all frequent 1itemsets, together with their counts;
for ( $k=2 ; L_{k-1} \neq \varnothing ; k++$ ) $\{$
Generate $C_{k}$, the set of candidate $k$-itemsets, from $L_{k-1}$, the set of frequent ( $k-1$ )-itemsets found in the previous step;
Scan the transactions to count the occurrences of itemsets in $C_{k}$;
Find $L_{k}$, a subset of $C_{k}$ containing $k$-itemsets with counts no less than ( $s_{\text {min }} \% \cdot$ total \# of transactions) ; $\}$
Return $L_{1} \cup L_{2} \cup \ldots \cup L_{k}$;

## Candidate generation

From $L_{k-1}$ to $C_{k}$

- Join: combine frequent ( $k-1$ )-itemsets to form candidate $k$-itemsets
- Prune: ensure every size- $(k-1)$ subset of a candidate is frequent



## Example: final answer



## Other tricks and extensions

- Transaction reduction
- If a transaction does not contain any frequent $k$-itemset, remove it from further consideration
» AprioriTid, AprioriHybrid, from the same paper
- Dynamic itemset counting
- Why only introduce candidate itemsets at the end of a scan? Start counting them whenever there is enough support from smaller itemsets
- Fewer passes over data
» Brin et al., SIGMOD 1997
- Parallelization, sampling, incremental mining, etc.

