

Mining Data for Rules Underlying User Behavior



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Overview

Understand what association rules are

Mine transaction data for association rules using the apriori algorithm

Implement the apriori algorithm on a bakery sales dataset

Recommendation Algorithms

Content Based Filtering



Find products with
“similar” attributes

Collaborative Filtering



Find products liked
by “similar” users

Association Rules Learning



Find “complementary”
products

Association Rules Learning



What items are bought together in a transaction?

What items are bought by a user in a short period of time?

Market basket analysis

Association Rules Learning





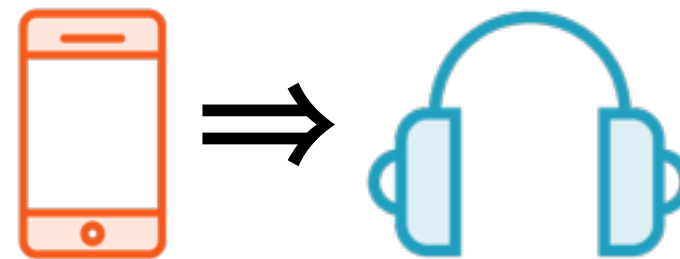
Conditional Probabilities

$$P(\text{headphones} / \text{phone})$$

Association Rules Learning

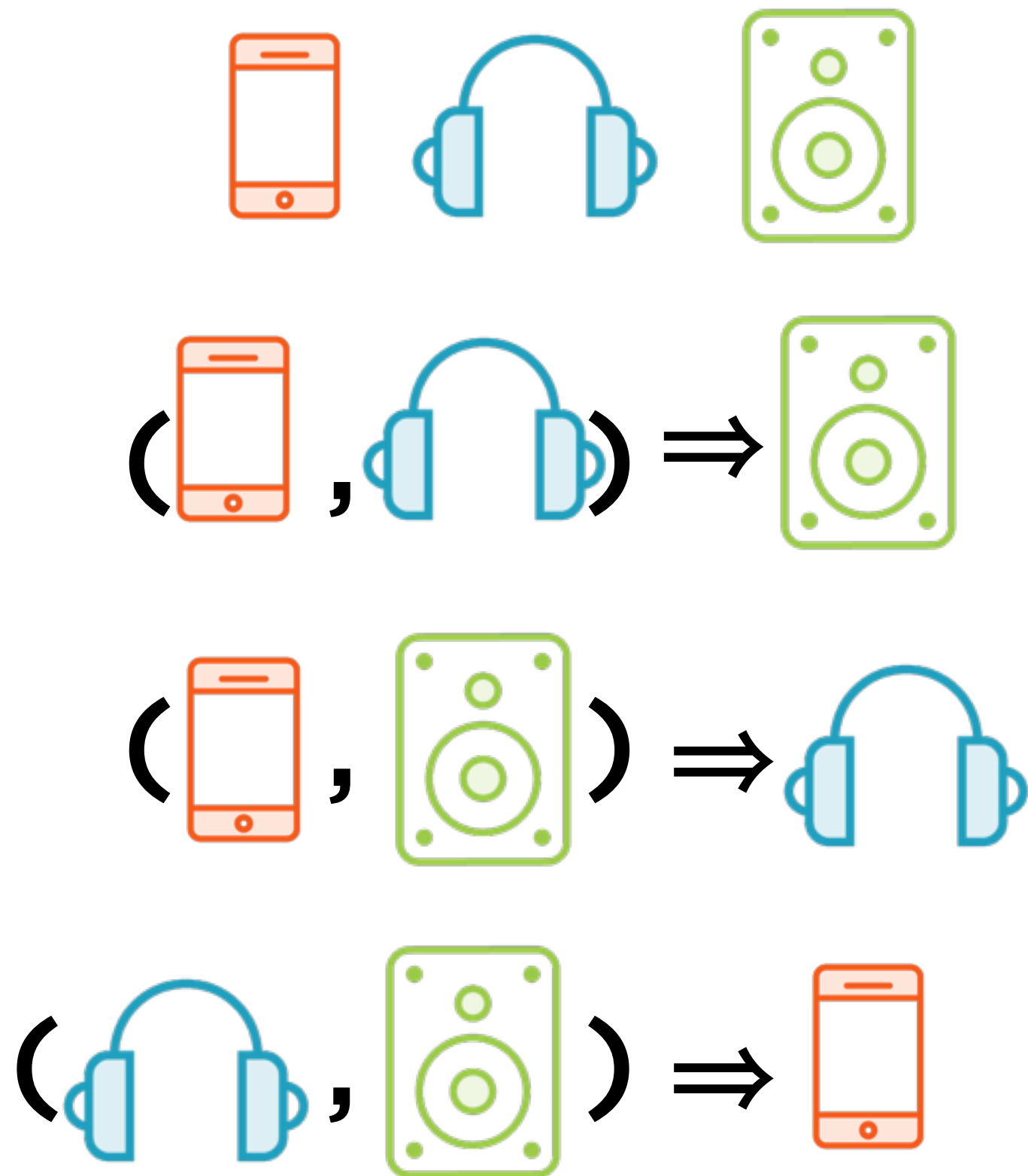


Does a person buying  increase the likelihood of buying ?

















Association rule

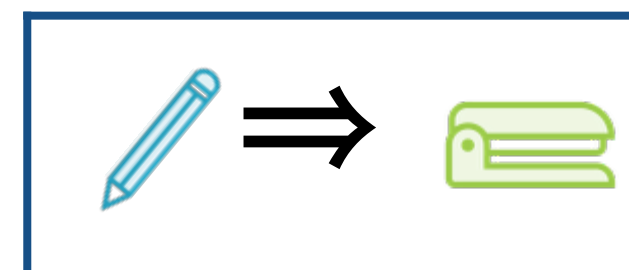
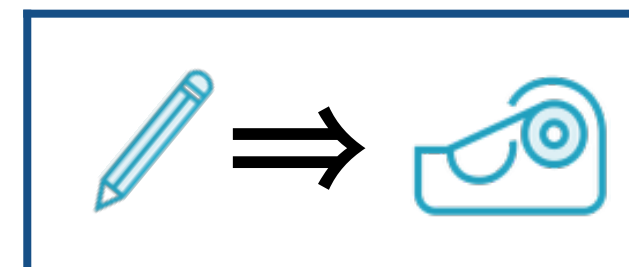
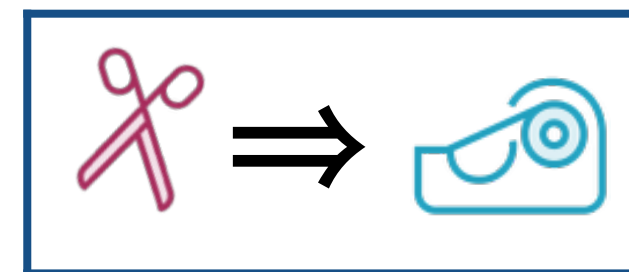
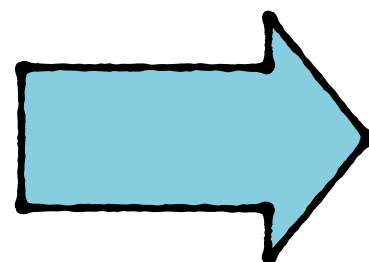
Association Rules Learning



Mining for Association Rules

1	  
2	  
3	   
4	 
5	 

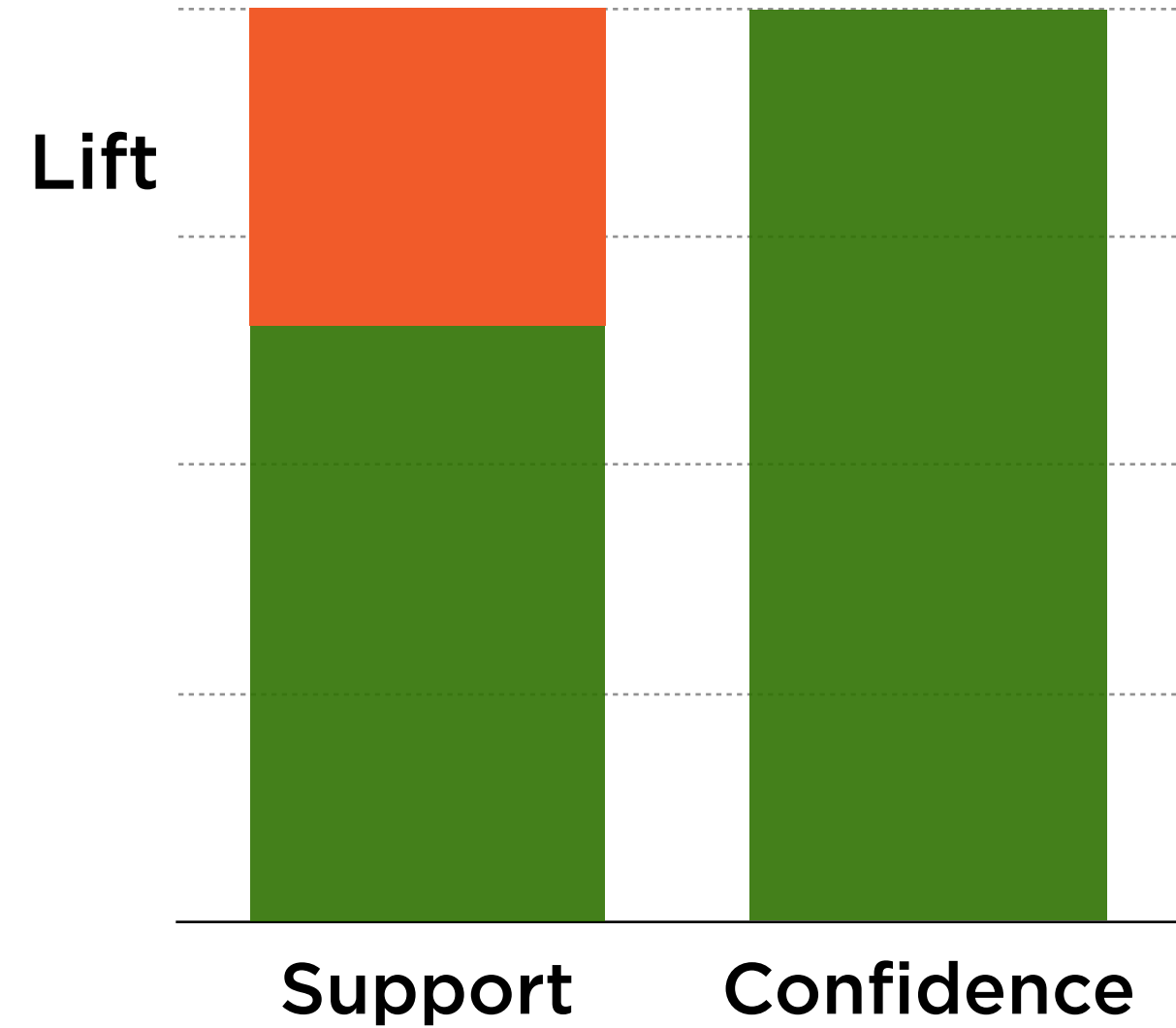
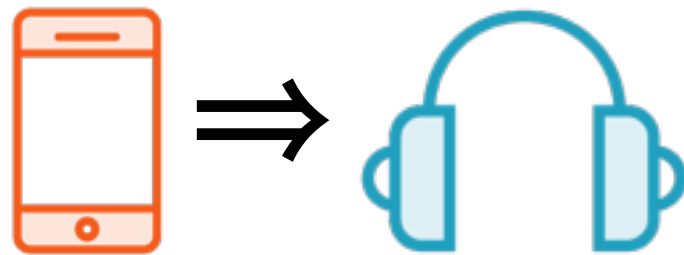
Transactions



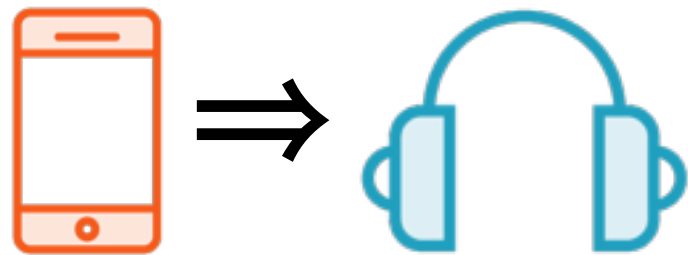
Rules

Measuring the Strength of a Rule

Measuring Rule Strength



Measuring Rule Strength

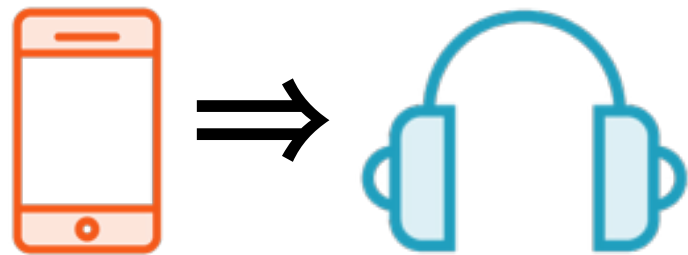




What proportion of all transactions contain both items?

$$P(\text{headphones}, \text{smartphone}) = 1\%$$

Support

Measuring Rule Strength

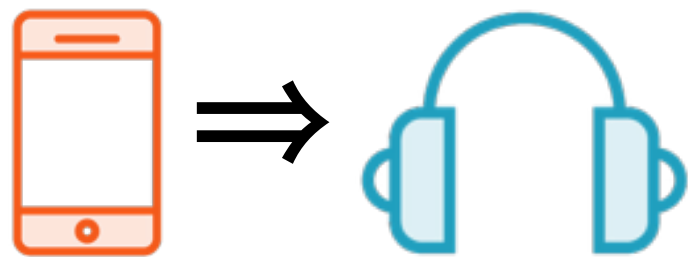


Out of all transactions with  how many include  ?

$$P(\text{headset} / \text{smartphone}) = 5\%$$

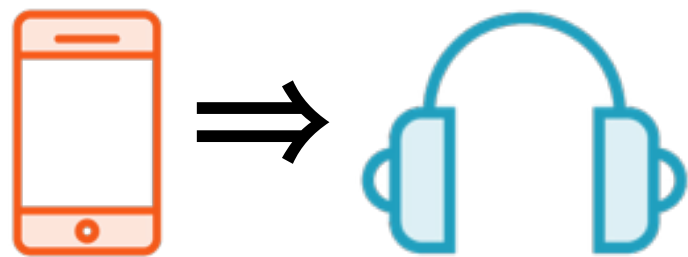
Confidence

Measuring Rule Strength



$$\text{Confidence} = \frac{P(\text{headphones}, \text{smartphone})}{P(\text{smartphone})}$$

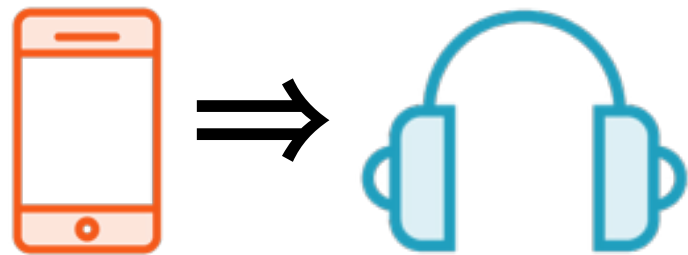
Measuring Rule Strength





$$\text{Confidence} = \frac{P(\text{headphones} / \text{smartphone})}{\text{Supp}(\text{smartphone})}$$

The equation defines the confidence of a rule. The numerator is the probability of the consequent (headphones) given the antecedent (smartphone), represented as $P(\text{headphones} / \text{smartphone})$. The denominator is the support of the antecedent, represented as $\text{Supp}(\text{smartphone})$. The word "Confidence" is written in orange below the numerator.

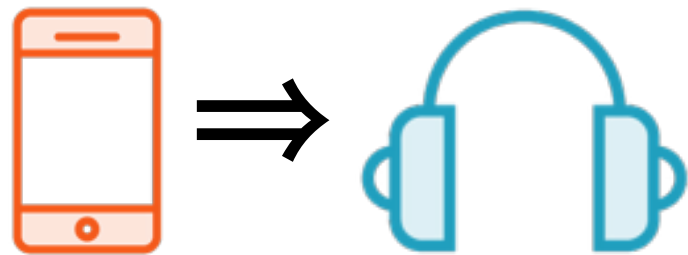
Measuring Rule Strength





How much does the likelihood of buying  increase when  is bought?

Lift

Measuring Rule Strength



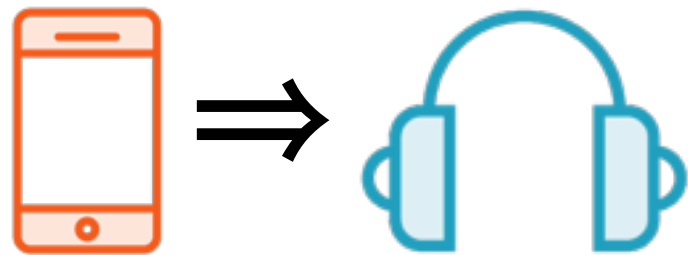
How much does the likelihood of buying  increase when  is bought?



$$P(\text{headphones}) = 3\%$$

$$P(\text{headphones} / \text{smartphone}) = 5\%$$

$$\text{Lift}(\text{smartphone} \Rightarrow \text{headphones}) = \frac{5\%}{3\%} = 1.67$$

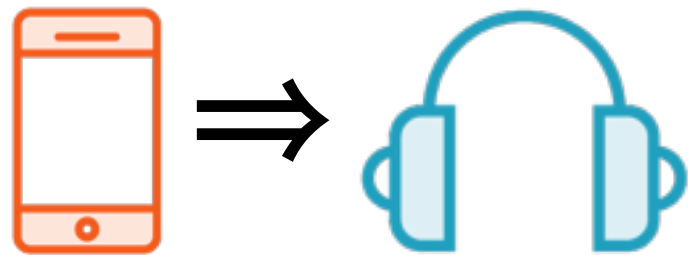
Measuring Rule Strength



How much does the likelihood of buying  increase when  is bought?

$$\text{Lift}(\text{smartphone} \Rightarrow \text{headphones}) = \frac{P(\text{headphones} / \text{smartphone})}{P(\text{smartphone})}$$
$$= \frac{\text{Conf}(\text{headphones} / \text{smartphone})}{\text{Supp}(\text{smartphone})}$$

Measuring Rule Strength



Support

Likelihood of all items in the rule being in a single transaction

Confidence

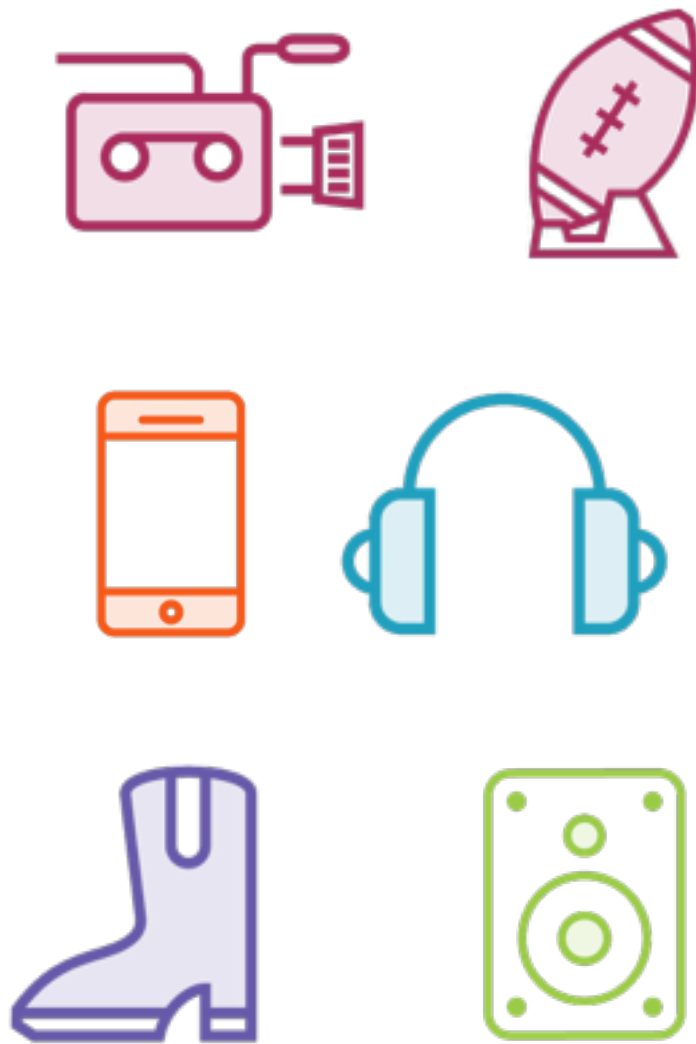
Likelihood of second item being bought once the first is bought

Lift

Change in likelihood of second item being bought once the first is bought

Mining for Rules Using the Apriori Algorithm

Mining for Association Rules



Catalog

Brute Force

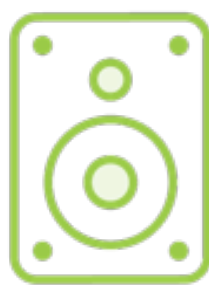
Find all possible 2 item sets **x 2**

Find all possible 3 item sets **x 3**

⋮

Find all possible N item sets **x N**

Mining for Association Rules



Catalog

Apriori Algorithm

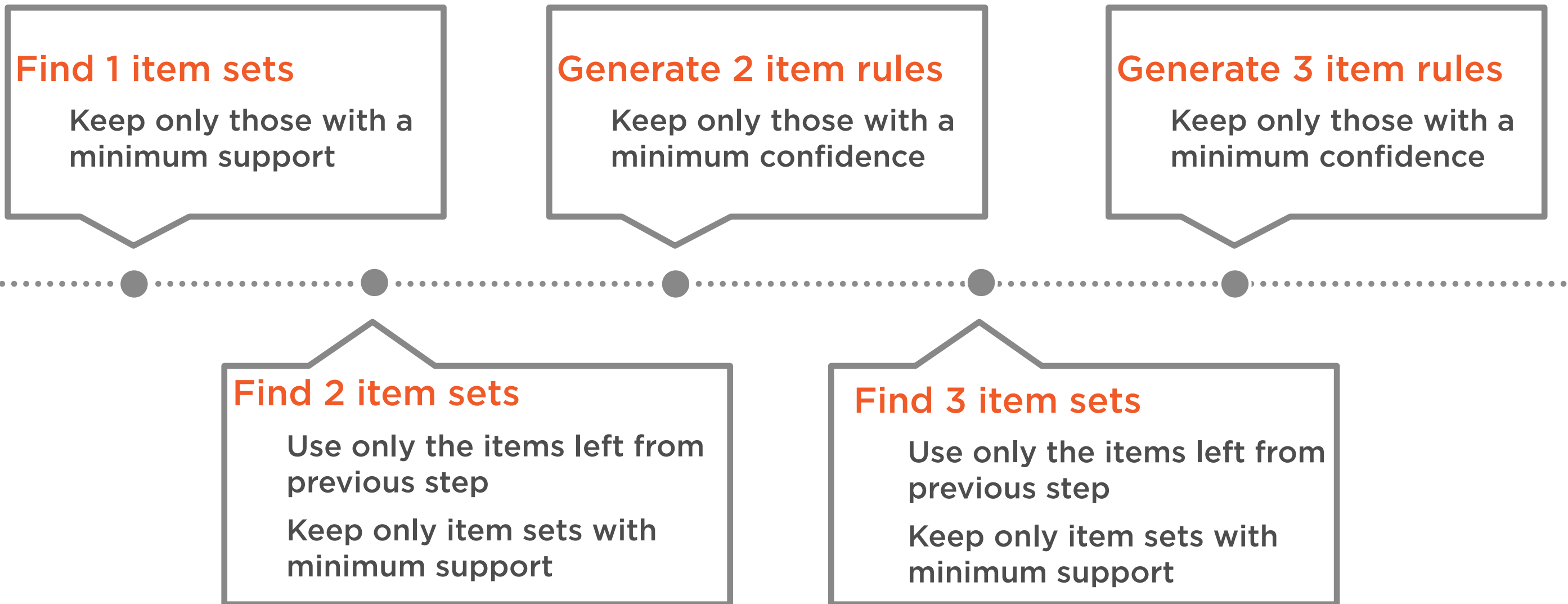
Prune the number of items in each stage

Use metrics to check how important an item set is

Support

Confidence

Apriori Algorithm



Apriori Algorithm

Find 1 item sets

Keep only those with a minimum support

















Apriori Algorithm

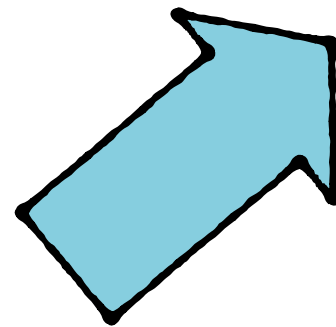
Find all possible 1 item sets



Compute the support of each set

2/5	1/5	1/5	1/5	2/5	3/5	4/5
-----	-----	-----	-----	-----	-----	-----

1	  
2	  
3	   
4	 
5	 



Apriori Algorithm

Find all possible 1 item sets



Compute the support of each set

2/5	1/5	1/5	1/5	2/5	3/5	4/5
-----	-----	-----	-----	-----	-----	-----

Drop item sets with support below a minimum threshold



Apriori Algorithm

Find 1 item sets

Keep only those with a minimum support

Find 2 item sets

Use only the items left from previous step

Keep only item sets with minimum support

















Apriori Algorithm

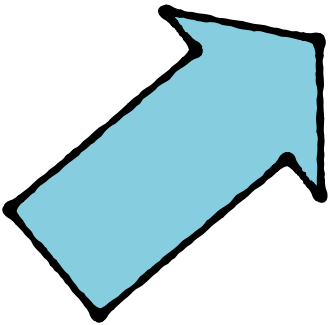
Find all possible 2 item sets

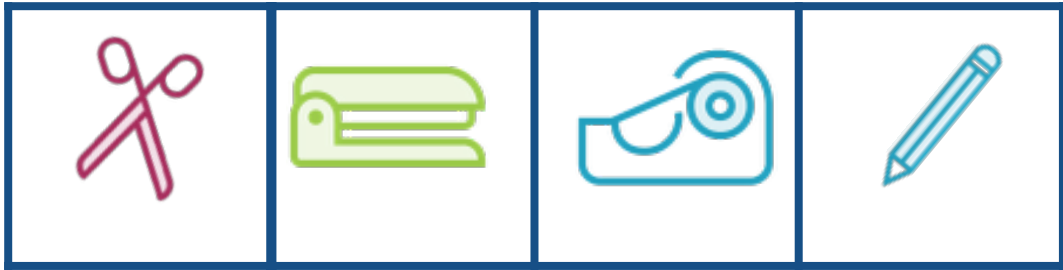


Compute the support of each set

0/5	1/5	2/5	2/5	0/5	2/5
-----	-----	-----	-----	-----	-----

1	  
2	  
3	   
4	 
5	 





Apriori Algorithm

Find all possible 2 item sets



Compute the support of each set

0/5	1/5	2/5	2/5	0/5	2/5
-----	-----	-----	-----	-----	-----

Drop item sets with support below a minimum threshold



Apriori Algorithm





Apriori Algorithm

Find all possible 2 item sets



Compute the support of each set

0/5	1/5	2/5	2/5	0/5	2/5
-----	-----	-----	-----	-----	-----

Drop item sets with support below a minimum threshold

















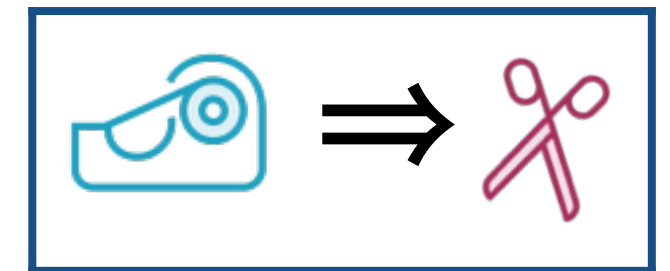
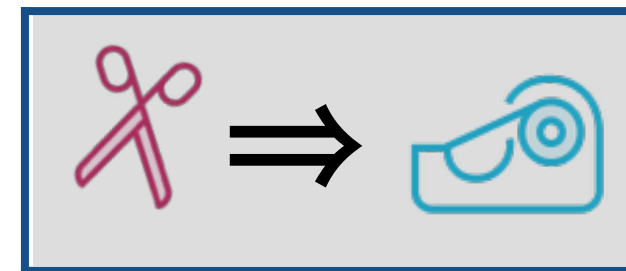
Apriori Algorithm

Drop item sets with support below a minimum threshold

From each item set, generate rules

Keep rules with a minimum confidence

1	  
2	  
3	   
4	 
5	 

















2/2

Apriori Algorithm

Drop item sets with support below a minimum threshold

From each item set, generate rules

Keep rules with a minimum confidence

1	  
2	  
3	   
4	 
5	 



2/2

2/3

Apriori Algorithm

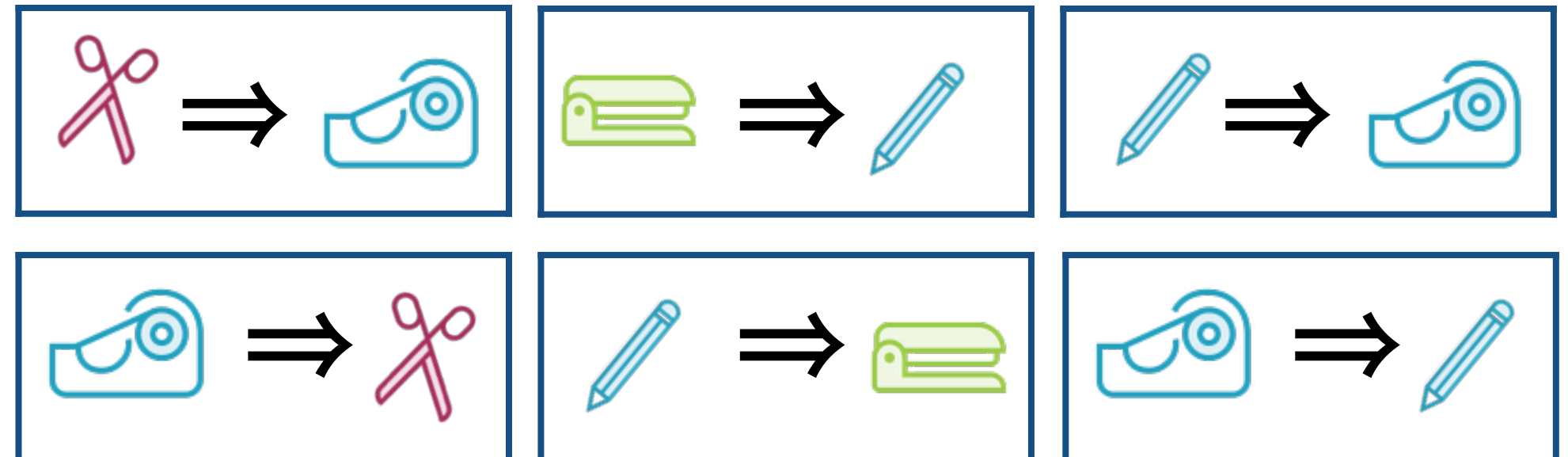
Drop item sets with support below a minimum threshold



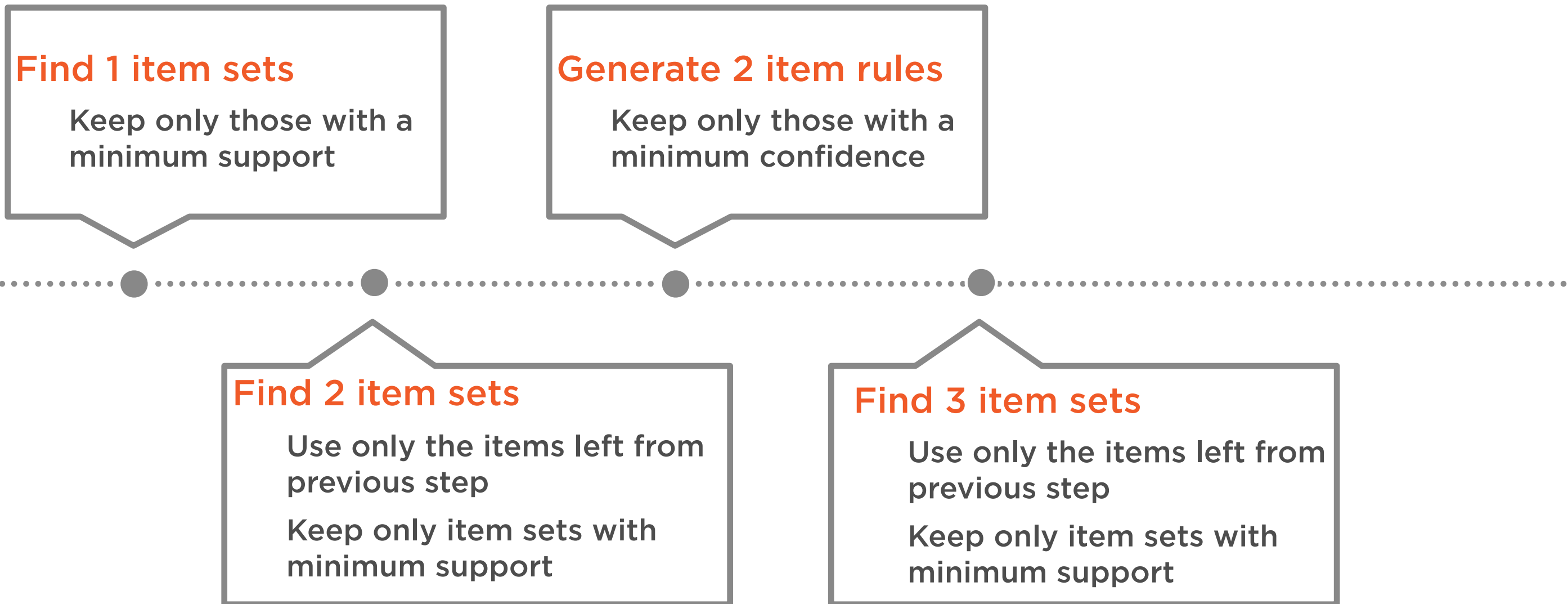
From each item set, generate rules



Keep rules with a minimum confidence



Apriori Algorithm



















Apriori Algorithm

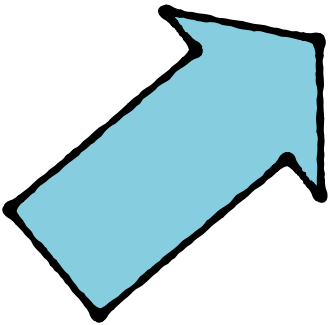
Find all possible 3 item sets

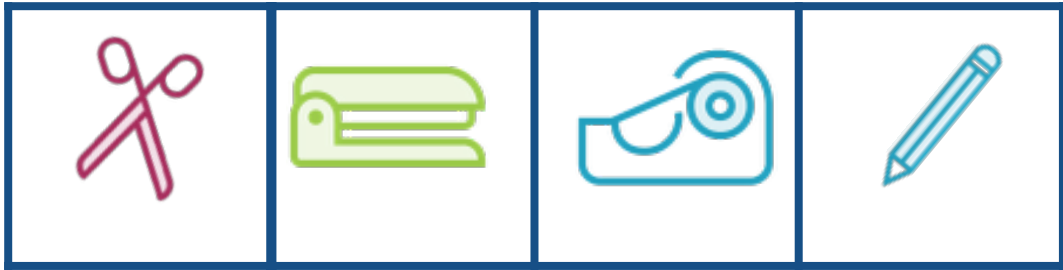


Compute the support of each set

0/5	1/5	0/5
-----	-----	-----

1	  
2	  
3	   
4	 
5	 





Apriori Algorithm

Find all possible 3 item sets



Compute the support of each set

0/5	1/5	0/5
-----	-----	-----

Drop item sets with support below a minimum threshold

Algorithm stops

Apriori Algorithm



Demo

Find association rules in a bakery dataset

Association Rules for Bakery Items

Set up the data

Receipts and item meta data

Implement the Apriori Algorithm

Compute the support

Set up a function to compute support for any items

Summary

Understand what association rules are

Mine transaction data for association rules using the apriori algorithm

Implement the apriori algorithm on a bakery sales dataset