






Learning Objectives

- After this segment, students will be able to
 - Contrast collocations and associations
 - Describe collocation interest measures

Background: Association Rules

- Association rule e.g. (Diaper in T => Beer in T)

Transaction	Items Bought
1	{socks,  , milk,  , beef, egg, ...}
2	{pillow,  , toothbrush, ice-cream, muffin, ...}
3	{  ,  , pacifier, formula, blanket, ...}
...	...
n	{battery, juice, beef, egg, chicken, ...}

- Support: probability (Diaper and Beer in T) = 2/5
 - Confidence: probability (Beer in T | Diaper in T) = 2/2
- Apriori Algorithm
 - Support based pruning using monotonicity

Apriori Algorithm

How to eliminate infrequent item-sets as soon as possible?

Transaction Id	Time	Item-types bought
1101	18:35	Milk, bread, cookies, juice
792	19:38	Milk, juice
2130	20:05	Milk, eggs
1735	20:40	Bread, cookies, coffee

Support threshold ≥ 0.5

Apriori Algorithm

Eliminate infrequent singleton sets

Transaction Id	Time	Item-types bought	Item-type	Count
1101	18:35	Milk, bread, cookies, juice	Milk	3
792	19:38	Milk, juice	Bread	2
2130	20:05	Milk, eggs	Cookies	2
1735	20:40	Bread, cookies, coffee	Juice	2
			Coffee	1
			Eggs	1

Support threshold ≥ 0.5

Milk

Bread

Cookies

Juice

Coffee

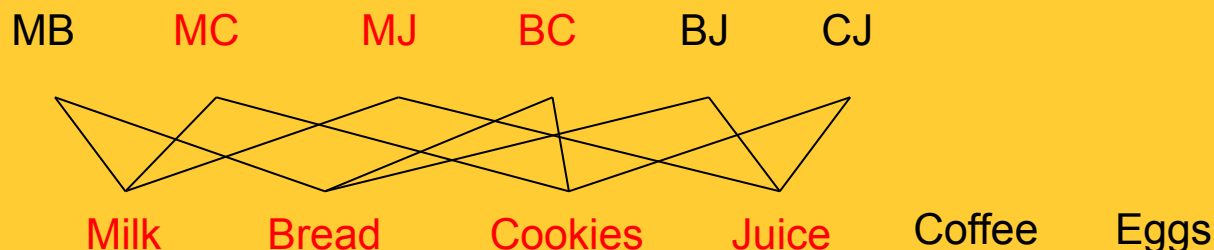
Eggs

Apriori Algorithm

Make pairs from frequent items & prune infrequent pairs!

Transaction Id	Time	Item-types bought	Item-type	Count	Item Pair	Count
1101	18:35	Milk, bread, cookies, juice	Milk	3	Milk, Cookies	2
792	19:38	Milk, juice	Bread	2	Milk, Juice	2
2130	20:05	Milk, eggs	Cookies	2	Bread, Cookies	2
1735	20:40	Bread, cookies, coffee	Juice	2	Milk, Bread	1
			Coffee	1	Bread, Juice	1
			Eggs	1	Cookies, Juice	1

Support threshold ≥ 0.5

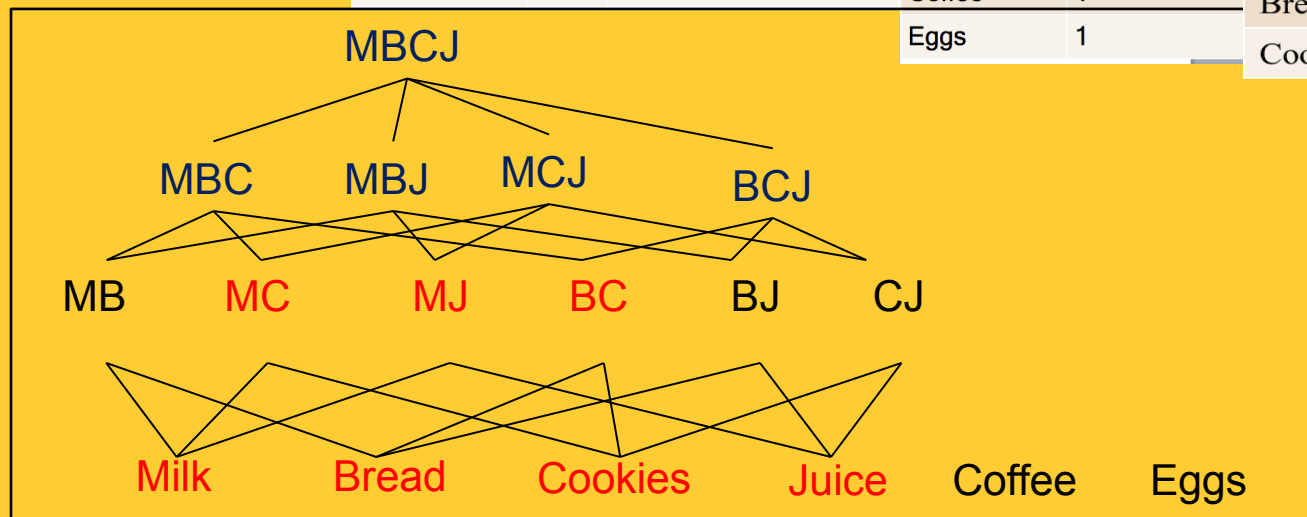


Apriori Algorithm

Make triples from frequent pairs & Prune infrequent triples!

Support threshold
 ≥ 0.5

Transaction Id	Time	Item-types bought	Item-type	Count	Item Pair	Count
1101	18:35	Milk, bread, cookies, juice	Milk	3	Milk, Cookies	2
792	19:38	Milk, juice	Bread	2	Milk, Juice	2
2130	20:05	Milk, eggs	Cookies	2	Bread, Cookies	2
1735	20:40	Bread, cookies, coffee	Juice	2	Milk, Bread	1
			Coffee	1	Bread, Juice	1
			Eggs	1	Cookies, Juice	1








No triples generated
 Due to Monotonicity!

Apriori algorithm
 examined only
 12 subsets
 instead of 64!

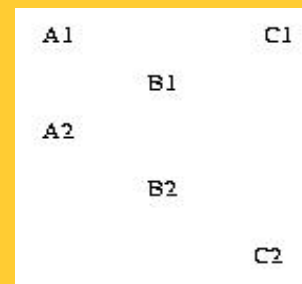


Association Rules Limitations

- Transaction is a core concept!
 - Support is defined using transactions
 - Apriori algorithm uses transaction based Support for pruning

Transaction	Items Bought
1	{socks,  , milk,  , beef, egg, ...}
2	{pillow,  , toothbrush, ice-cream, muffin, ...}
3	{  ,  , pacifier, formula, blanket, ...}
...	...

- However, spatial data is embedded in continuous space
 - Transactionizing continuous space is non-trivial !

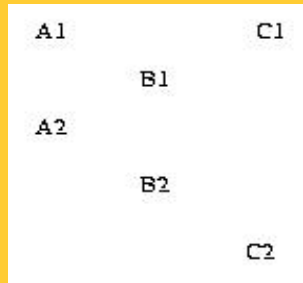


Spatial Association vs. Cross-K Function

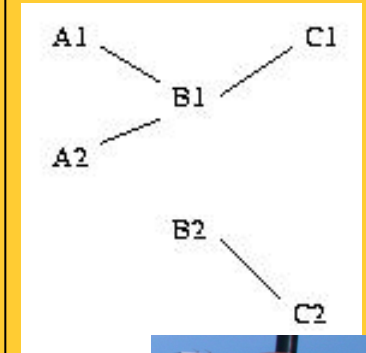
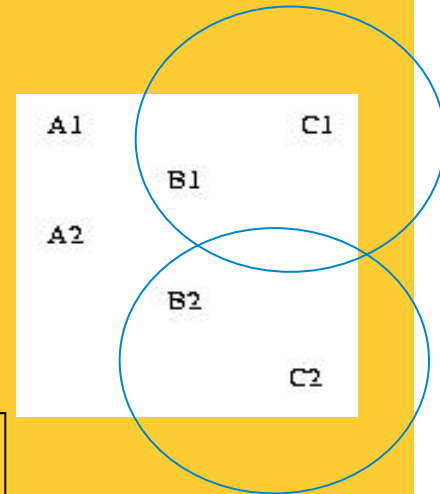
- Spatial Association Rule (Han 95)
- **Output = (B,C)** with threshold 0.5
- **Transactions by Reference feature, e.g. C**
Transactions: (C1, B1), (C2, B2)
Support (A,B) = \emptyset
Support(B,C) = $2 / 2 = 1$

- **Cross-K Function**
Cross-K (A, B) = $2/4 = 0.5$
Cross-K(B, C) = $2/4 = 0.5$
Cross-K(A, C) = 0

Output = (A,B), (B, C) with threshold 0.5



Input = Feature
A,B, and, C, &
instances A1, A2,
B1, B2, C1, C2



Spatial Colocation

Features: A, B, C

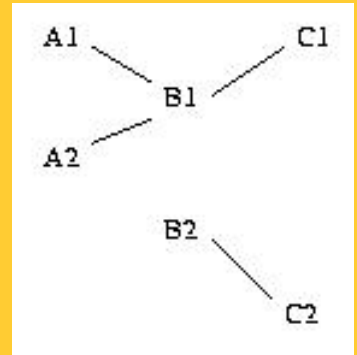
Feature Instances: A1, A2, B1, B2, C1, C2

Feature Subsets: (A,B), (A,C), (B,C), (A,B,C)

Participation ratio (pr):

$\text{pr}(A, (A,B)) = \text{fraction of A instances neighboring feature } \{B\} = 2/2 = 1$

$\text{pr}(B, (A,B)) = 1/2 = 0.5$



Participation index (A,B) = $\text{pi}(A,B) = \min\{ \text{pr}(A, (A,B)), \text{pr}(B, (A,B)) \} = \min(1, 1/2) = 0.5$

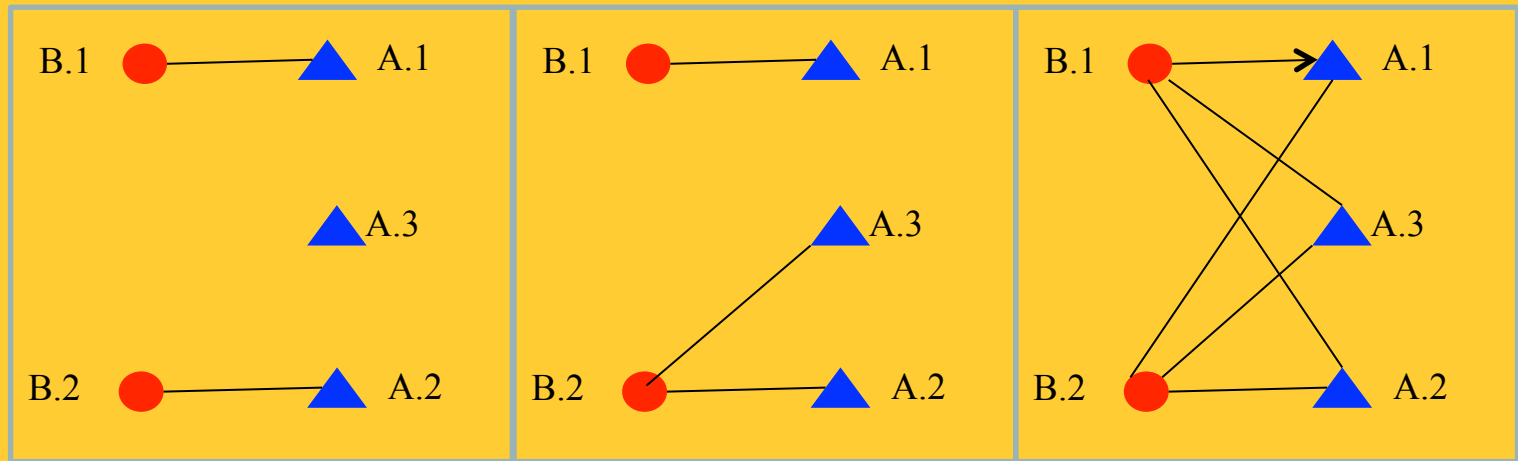
$\text{pi}(B, C) = \min\{ \text{pr}(B, (B,C)), \text{pr}(C, (B,C)) \} = \min(1,1) = 1$

Participation Index Properties:

(1) Computational: Non-monotonically decreasing like support measure

(2) Statistical: Upper bound on Ripley's Cross-K function

Participation Index \geq Cross-K Function

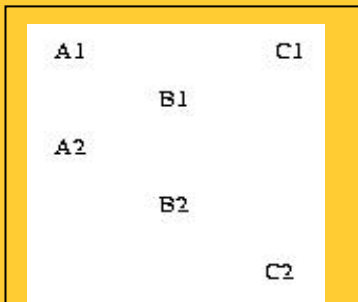


Cross-K (A,B)	$2/6 = 0.33$	$3/6 = 0.5$	$6/6 = 1$
PI (A,B)	$2/3 = 0.66$	1	1

Association Vs. Colocation

	Associations	Colocations
underlying space	Discrete market baskets	Continuous geography
event-types	item-types, e.g., Beer	Boolean spatial event-types
collections	Transaction (T)	Neighborhood N(L) of location L
prevalence measure	Support, e.g., Pr.[Beer in T]	Participation index, a lower bound on Pr.[A in N(L) B at L]
conditional probability measure	Pr.[Beer in T Diaper in T]	Participation Ratio(A, (A,B)) = Pr.[A in N(L) B at L]

Spatial Association Rule vs. Colocation

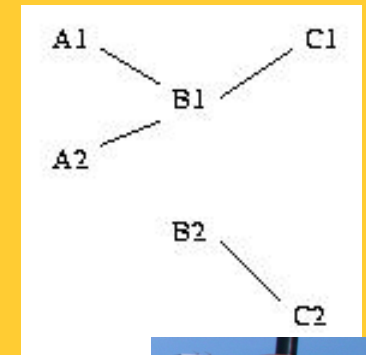
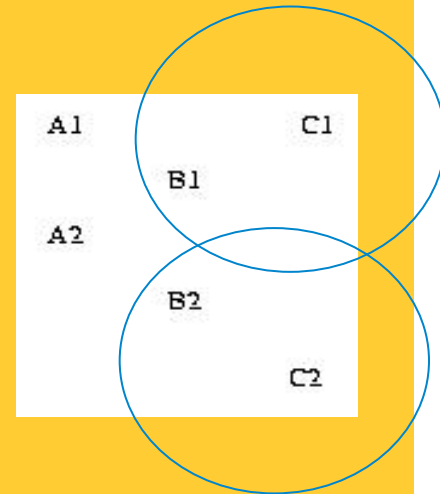


Input = Spatial
feature A,B, C,
& their
instances

- Spatial Association Rule (Han 95)
- **Output = (B,C)**
- **Transactions by Reference feature C**
Transactions: (C1, B1), (C2, B2)
Support (A,B) = \emptyset , Support(B,C) = $2 / 2 = 1$

- **Cross-K Function**
Cross-K (A, B) = $2/4 = 0.5$
Cross-K(B, C) = $2/4 = 0.5$
Output = (A,B), (B, C)

- **Colocation - Neighborhood graph**
Output = (A,B), (B, C)
 $PI(A,B) = \min(2/2, 1/2) = 0.5$
 $PI(B,C) = \min(2/2, 2/2) = 1$



Spatial Colocation: Trends

- Algorithms
 - Join-based algorithms
 - One spatial join per candidate colocation
 - Join-less algorithms
- Spatio-temporal
 - Which events co-occur in space and time?
 - (bar-closing, minor offenses, drunk-driving citations)
 - Which types of objects move together?

