ASSOCIATION

BY: CANDACE MCQUEEN

ASSOCIATION means they have:

> a common purpose and having a formal structure

➢ A connection or a combination

➢ Friendship

➢ A Correlation

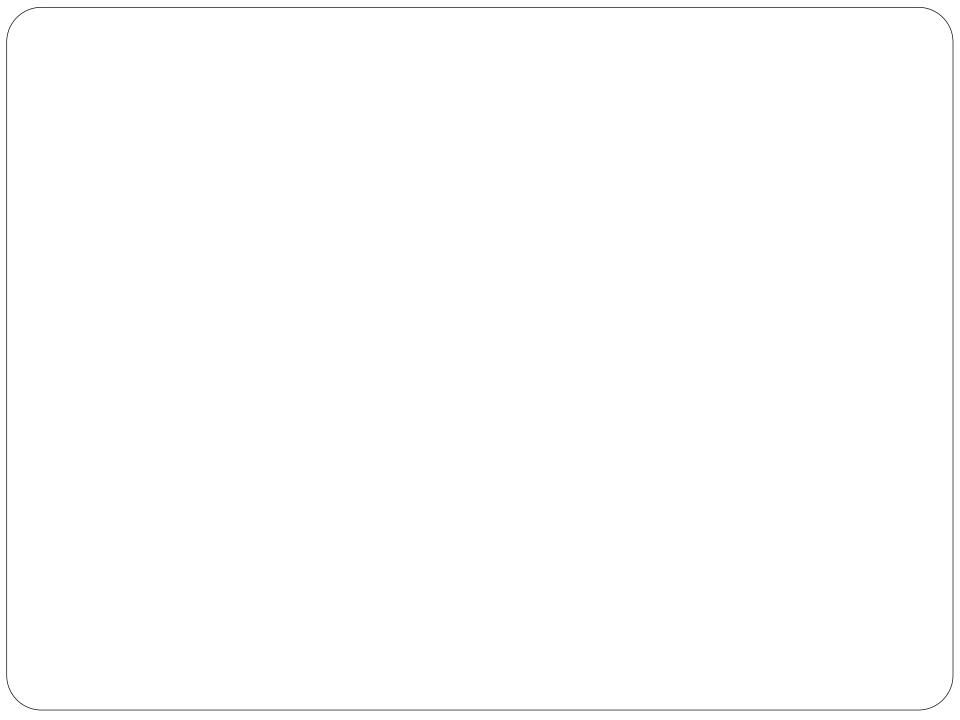


Their close association did not last long.

.....Some examples of Associations.....

A FIST IN THE FACE>>>>>>BRUISE





An association mining problem can be decomposed using **APRIORI**.....

What APRIORI does...

- Calculate rules that express the probable co-occurrence of items within frequent item sets.
- Apriori calculates the probability of an item being present in a frequent item set, given that another item or items is present

Association Rules

 The Apriori algorithm calculates rules that express probabilistic relationships between items in frequent item sets. For example, a rule derived from frequent item sets containing A, B, and C might state that if A and B are included in a transaction, then C is likely to also be included.

>An <u>Association Rule</u> states that an item or group of items implies the presence of another item with some probability.

>Unlike decision tree rules, which predict a target, association rules simply express correlation.

Antecedent and Consequent

IF.....

The IF component of an association rule is known as the antecedent.

THEN.....

 The THEN component is known as the consequent.

✓ The antecedent and the consequent are disjoint;they have no items in common.



Items on the Sonic menu.....

















INFORMATION ABOUT MY DATA.....

- DATA TAKEN ON NOV. 30 2011
 - 21 ATTRIBUTES
 - Breakfast 1, Breakfast 2, burger 1, burger 2, burger 3, WP 1, WP 2, CHK 1, CHK 2, SWAMP 1, SWAMP 2, SWAMP 3, SWAMP 4, SIDE 1 SIDE2, SIDE 3, FOUNTAIN 1, FOUNTAIN 2, FOUNTIAN 3, FOUNTAIN 4, FROZEN 1
 - •99 TICKETS/RECIEPTS

.....PLEASE LOOK AT THE RECIEPT THAT I HAVE GIVEN YOU.....

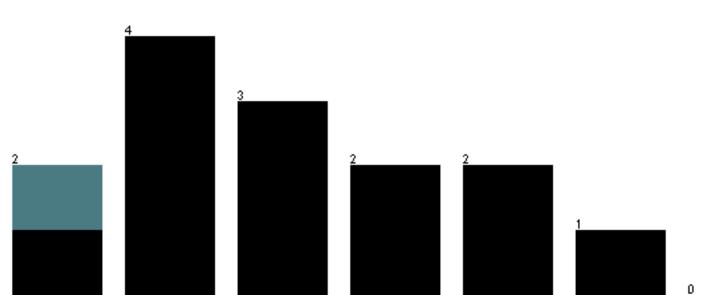
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	A1 🔹 🏂 Breakfast 1										¥						
	А	В	С	D	E	F	G	Н		J	K	L	М	N	0	Р	
1	Breakfast	Breakfast	burger 1	burger 2	burger 3	WP 1	WP 2	CHK 1	SWAMP 1	SWAMP 2	SWAMP 3	SWAMP 4	SIDE 1	SIDE2	SIDE 3	FOUNTAIN FO	סו
2	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	XL DR PEP ?	=
3	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	XL SWEET TE	A
4	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	LG COKE ?	
5	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	LG COFFEE LO	G '
6	SEC CROS	?	?	?	?	?	?	?	?	?	?	?	тот	?	?	MD DR PEI?	
7	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	MD COFFF M	D
8	SEC BURR	?	?	?	?	?	?	?	?	?	?	?	?	?	?	???	
9	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	MD WATE ?	
10	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	MD WATE ?	
11	?	?	?	?	?	?	?	?	?	?	?	?	SM FF	?	?	LG GRAPE SL	U.
12	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	LG DIET CC?	
13	JR SEC BUI	?	?	?	?	?	?	?	?	?	?	?	?	?	?	????	
14	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	???????????????????????????????????????	
15	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	???????????????????????????????????????	
16	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	LG COKE ?	
17	JR SEC BUI	?	?	?	?	?	?	?	?	?	?	?	?	?	?	???????????????????????????????????????	
18	?	?	?	?	?	?	?	?	5 PC STRIP	5 PC STRIP	?	?	CCTT	CCTT	?	LG ROOT E LO	GI
19	REC BURR	2)	2	2	2	2	2	2	?	2	2	2	2	2	IG SWEET ?	
	WEKA DATA EXCEL A																

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```
@relation Sonic_wide
@attribute Breakfast1 {SECCROS,SECBURR,JRSECBUR,BECBURR,SSBUR,BECTST,SECTST,'?'}
@attribute Breakfast2 {SECCROS, SECBURR, JRSECBUR, BECBURR, SSBUR, BECTST, SECTST, '?'}
@attribute burger1 {BLT,BCNMELT,GRILLCHZ,JRBCNMLT,JRCBDLX,JRDBLCB,JRMSHMLT,MAYOCB,MAYOHB,M/
@attribute burger2 {BLT,BCNMELT,GRILLCHZ,JRBCNMLT,JRCBDLX,JRDBLCB,JRMSHMLT,MAYOCB,MAYOHB,M
@attribute burger3 {BLT,BCNMELT,GRILLCHZ,JRBCNMLT,JRCBDLX,JRDBLCB,JRMSHMLT,MAYOCB,MAYOHB,M
@attribute WP1 {WPCB,WPCD,WPCKSTP,WPHB.'?'}
@attribute WP2 {WPCB,WPCD,WPCKSTP,WPHB,'?'}
@attribute CHK1 {CKCLBTST,GLWRAP,5PCSTRIP,'?'}
@attribute CHK2 {CKCLBTST,GLWRAP,5PCSTRIP,'?'}
@attribute SWAMP1 {AMERDOG,CHICAGO,CORNDOG,KICKNCNY,QTRCCCONEY,CCDOG,'?'}
@attribute SWAMP1 {AMERDOG,CHICAGO,CORNDOG,KICKNCNY,QTRCCCONEY,CCDOG,'?'}
@attribute SWAMP2 {AMERDOG,CHICAGO,CORNDOG,KICKNCNY,OTRCCCONEY,CCDOG,'?'}
@attribute SWAMP3 {AMERDOG,CHICAGO,CORNDOG,KICKNCNY,QTRCCCONEY,CCDOG,
@attribute SWAMP4 {AMERDOG,CHICAGO,CORNDOG,KICKNCNY,QTRCCCONEY,CCDOG,'?'}
@attribute SIDE2 {CCTT,CFF,CHEDBITE,CTT,FF,LGCTT,LGFF,LGRINGS,LGTT,MOZZ,RINGS,SMCFF,SMFF,SM
@attribute FOUNTAIN1 {BOTTLEWTR,KDCHERRYSLUSH,KDHIC,KDPOWSLUSH,LGCHOCSHAKE,LGCHRYLA,LGCOFF
@attribute FOUNTAIN2 {BOTTLEWTR,KDCHERRYSLUSH,KDHIC,KDPOWSLUSH,LGCHOCSHAKE,LGCHRYLA,LGCOFFI
@attribute FOUNTIAN3 {BOTTLEWTR,KDCHERRYSLUSH,KDHIC,KDPOWSLUSH,LGCHOCSHAKE,LGCHRYLA,LGCOFFI
@attribute FOUNTAIN4 {BOTTLEWTR,KDCHERRYSLUSH,KDHIC,KDPOWSLUSH,LGCHOCSHAKE,LGCHRYLA,LGCOFF
@attribute FROZEN1 {DISH,LGVANSHAKE,MDRTBRFLOAT,REGBANASHAKE,REGRTBRFLOAT,REGSTRWSHAKE,REG
?,?,?,?,?,?,?,?,?,?,?,?,?,?,?,XLSWEETTEA,?,?,?,?
?,?,?,?,?,?,?,?,?,?,?,?,?,?,?,?,LGCOKE,?,?,?,?
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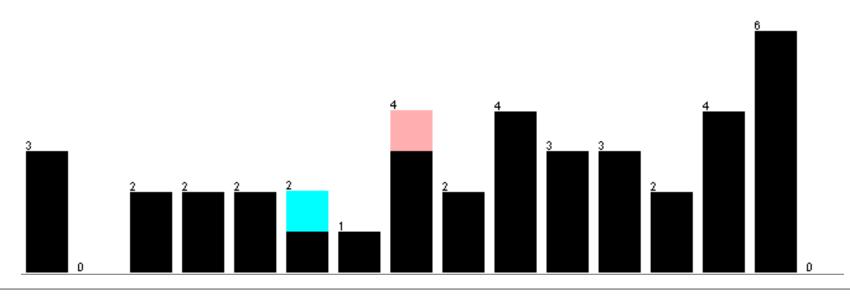
Breakfast 1.....

	ed attribute				
	ne: Breakfast1	Distinct: 7	Type: Nominal		
MISSIN	ig: 84 (85%)	Distinct: 7	Unique: 2 (2%)		
No.	Label	Count	Weight		
	1 SECCROS	1	1.0		
	2 SECBURR	2	2.0		
	3 JRSECBUR	4	4.0		
	4 BECBURR	3	3.0		
	5 SSBUR	2	2.0		
	6 BECTST	2	2.0		
	7 SECTST	1	1.0		
	8 ?	0	0.0		



Burger 1.....

Selected	attribute				
	: burger1 : 59 (60%)	Distinct: 14	Type: Nominal Unique: 1 (1%)		
No.	Label	Count	Weight		
	L BLT	3	3.0		
	2 BCNMELT	0	0.0		
	3 GRILLCHZ	2	2.0		
	I JRBCNMLT	2	2.0		
	JRCBDLX	2	2.0		
(5 JRDBLCB	2	2.0		
	JRMSHMLT	1	1.0		
1	MAYOCB	4	4.0		
9	MAYOHB	2	2.0		
1	MAYOSSCB	4	4.0		



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Rules.....

Best rules found:

1. SIDE1=SMTT 2 ==> Breakfast1=JRSECBUR 2 <conf:(1)> lift:(24.75) lev:(0.02) [1] conv:(1.92) 2. Breakfast1=SSBUR 2 ==> FOUNTAIN1=LGDIETDRPEP 2 <conf:(1)>lift:(33) lev:(0.02) [1] conv:(1.94) 3. burger2=MUSTSSCB 2 ==> burger1=MAYOSSCB 2 <conf:(1)> lift:(24.75) lev:(0.02) [1] conv:(1.92) 4. FOUNTAIN2=MDCHRYLA 2 ==> FOUNTAIN1=MDCHRYLA 2 <conf: (1)> lift: (24.75) lev: (0.02) [1] conv: (1.92) SWAMP1=QTRCCCONEY 4 ==> SIDE1=TT 3 <conf: (0.75)> lift: (7.43) lev: (0.03) [2] conv: (1.8) 5. FOUNTAIN1=LGDIETDRPEP 3 ==> Breakfast1=SSBUR 2 <conf:(0.67)> lift:(33) lev:(0.02) [1] conv:(1.47) 6. <conf: (0.67)> lift: (13.2) lev: (0.02) [1] conv: (1.42) 7. SIDE1=CCTT 3 ==> CHK1=CKCLBTST 2 8. SIDE1=CTT 3 ==> FOUNTAIN1=MDCOKE 2 <conf:(0.67)> lift:(9.43) lev:(0.02) [1] conv:(1.39) 9. FOUNTAIN1=MDDRPEP 3 ==> SIDE1=TT 2 <conf:(0.67)> lift:(6.6) lev:(0.02) [1] conv:(1.35) 10. burger1=SSDBLBCB 6 ==> FOUNTAIN1=MDCOKE 3 <conf:(0.5)> lift:(7.07) lev:(0.03) [2] conv:(1.39)

PRE PROCESSING DATA

- PREPROCESSING steps should be applied to make the data more suitable for results
- Increases/higher Support
- Taking out inferences that will not affect the data that is be sought for
 - HAPPY HOUR
 - •.99 LG. BEVERAGES
- Strip minority combinations out
 - Chocolate milk (2)

CONTINUATION OF PRE PROCESSING DATA.....

Issues:

- Small set/One day
- Short coming- ability to handle large data sets.
 - Errors occurring
 - Much manual labor
 - First time using this program

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