



RULE EVALUATION

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LIMITATION OF CONFIDENCE MEASURE

100 transactions:

75 bought movies

60 bought games

40 bought both

Both seem to
be strong
rules.

{movies}->{games}

support $40/100 = 0.4$

confidence $40/75 = 0.53$

{games}->{movies}

support $40/100 = 0.4$

confidence $40/60 = 0.67$

HOWEVER ...

100 transactions:
75 bought movies
60 bought games
40 bought both

$$P(\text{movies}) = 75/100 = 0.75$$

$$P(\text{games}) = 60/100 = 0.6$$

$$P(\text{movies and games}) = 40/100 = 0.4$$

So people tend not to buy
movies and games together!

$$\text{Correlation}(\text{movies, games}) = P(\text{movies and games}) / [P(\text{movies}) \times P(\text{games})] = 0.4 / (0.75 \times 0.6) = 0.89$$

The confidence measure is
sometimes misleading.

Correlation < 1 means negative correlation.

METRIC: LIFT (CORRELATION)

Measure of dependent or correlated events: Lift

Lift ($A \Rightarrow B$) = $\text{support}(\{A,B\}) / (\text{support}(A) \times \text{support}(B))$

$$\text{lift}(A \Rightarrow B) = \frac{P(A \wedge B)}{P(A)P(B)}$$

Association rules should have >1 lift to be meaningful.

THE LIFT (CORRELATION) MEASURE

	Game	Not Game	Total
Movie	40	35	75
Not movie	20	5	25
Total	60	40	100

$$P(\text{buy game}) = 0.6$$

$$P(\text{not buy movie}) = 0.25$$

$$P(\text{buy game and not buy movie}) = 0.20$$

$$\text{Lift}(\text{buy game} \rightarrow \text{not buy movie}) = 0.20 / (0.6 \times 0.25) = 1.33 > 1$$

Strong rule

ALTERNATIVE MEASURES

Association rule algorithms tend to produce too many rules.

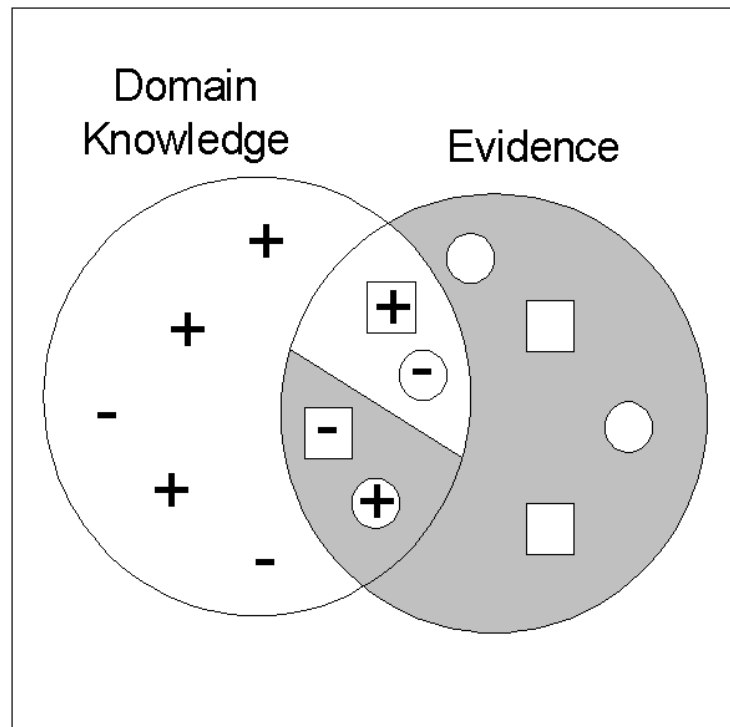
Many of them are uninteresting or redundant.

Uninteresting if it is known knowledge

Redundant if $\{A,B,C\} \rightarrow \{D\}$ and $\{A,B\} \rightarrow \{D\}$
have same support and confidence

INTERESTINGNESS VIA UNEXPECTEDNESS

Need to model expectation of users (domain knowledge)



- + Pattern expected to be frequent
- Pattern expected to be infrequent

- Pattern found to be frequent
- Pattern found to be infrequent

- ⊕ Expected Patterns
- ⊖ Expected Patterns

- ⊞ Unexpected Patterns
- ⊕ Unexpected Patterns

Need to combine expectation of users with evidence from data (i.e., extracted patterns)

WEKA ASSOCIATION RULES

Implemented a variation of Apriori Algorithm that iteratively reduces the minimum support until it finds the required number of rules with the given minimum confidence

Allows mining of “class association rules”: If the data have a class label attribute, the right hand side of a rule can be restricted to that label.

ASSOCIATION RULE MEASURES

In practice, what levels of support, confidence, and lift should we aim for?

Support:

Depends on data set and business problem

Common setting is 20–40% of the transactions

Confidence:

Strong confidence rules $\geq .9$, but .6 to .8 range might be OK

Lift:

Should be above 1.0, the higher the better

Levels of 2 and above can occasionally be seen but more likely to see around 1.3 to 1.5