

Association rule learning

Paweł Teisseyre

- Introduction.
- Market Basket Analysis.
- Generalized Association rules.



Project co-financed by European Union within the framework of European Social Fund

- example of unsupervised learning,
- the main goal is to find joint values of variable $x = (x_1, \dots, x_p)$ that appear most frequently in the data base.
- General task: find s_1, \dots, s_p such that:

$$P \left[\bigcap_{j=1}^p (x_j \in s_j) \right] \quad (1)$$

is **large**.

- In other words: find regions of x where probability is high.



Project co-financed by European Union within the framework of European Social Fund

- General approaches to solving (1) are not feasible in commercial applications ($p \approx 10^4$, $n \approx 10^8$).
- Some simplifications lead to MBA (all variables are binary).
- One can apply the technique of dummy variables to turn (1) into a problem involving only binary-valued variables.



Project co-financed by European Union within the framework of European Social Fund

- **Transactions:**

Transaction Id	Milk	Bread	Butter	Beer
1	1	1	0	0
2	0	0	1	0
3	0	0	0	1
4	1	1	1	0
5	0	1	0	0



Project co-financed by European Union within the framework of European Social Fund

Market Basket Analysis

Notation:

- \mathcal{I} - set of all products.
- $X, Y \in \mathcal{I}$ -itemsets.
- $\text{supp}(X)$ -% of transactions containing X . Example:
 $\text{supp}(\{Milk, Bread\}) = \frac{2}{5}$.

Goal of MBA:

- Find all itemsets X such that $\text{supp}(X) \geq t$, where t is user-specified threshold.



Project co-financed by European Union within the framework of European Social Fund

Apriori algorithm:

- 1 Find itemsets X containing only one product such that $\text{supp}(X) \geq t$.
- 2 Among itemsets found in step 1 find itemsets X with two elements such that $\text{supp}(X) \geq t$.
- 3 Among itemsets found in step 2 find itemsets X with three elements such that $\text{supp}(X) \geq t$.
- 4 Continue until all candidate itemsets from the previous pass have support less than the specified threshold.



Project co-financed by European Union within the framework of European Social Fund

Rules:

- $X, Y \in \mathcal{I}$. We write $X \implies Y$ to denote $X \cup Y$.
- We define $\text{supp}(X \implies Y) := \text{supp}(X \cup Y)$.
- Example: $\{Butter, Bread\} \implies \{Milk\}$. We have:

$$\text{supp}(\{Butter, Bread\} \implies \{Milk\}) = \frac{1}{5}$$



Project co-financed by European Union within the framework of European Social Fund

Useful quantities:

- Confidence (predict ability):

$$\text{conf}(X \implies Y) := \frac{\text{supp}(X \cup Y)}{\text{supp}(X)}$$

- Confidence is an estimate of $P(Y|X)$.
- Example:

$$\text{conf}(\{Butter, Bread\} \implies \{Milk\}) = \frac{1}{1} = 1.$$

Useful quantities:

- Lift (association measure):

$$\text{lift}(X \implies Y) := \frac{\text{supp}(X \cup Y)}{\text{supp}(X) \cdot \text{supp}(Y)}$$

- Lift is an estimate of $\frac{P(X,Y)}{P(X)P(Y)}$.
- Example:

$$\text{lift}(\{Butter, Bread\} \implies \{Milk\}) = \frac{1/5}{1/5 \cdot 2/5} = 2.5.$$



Project co-financed by European Union within the framework of European Social Fund

Problem:

- Rules with high confidence or lift, but low support, will not be discovered.
- For example, a high confidence rule such as *vodka* \implies *caviar* will not be uncovered owing to the low sales volume of the consequent caviar.



Project co-financed by European Union within the framework of European Social Fund

Unsupervised as supervised learning

- g - unknown density function. Our goal : estimate g based on observed independent data points $x_1, \dots, x_n \sim g$.



Project co-financed by European Union within the framework of European Social Fund

Procedure:

- 1 g_0 - reference, known distribution, e.g. uniform over a range of x .
- 2 Generate artificial sample from g_0 (N_0 points).
- 3 Assign weights $N_0/(N + N_0)$ to artificial sample points and $N/(N + N_0)$ to original points.
- 4 Assign $Y = 1$ to original data points and $Y = 0$ to artificial ones.
- 5 Estimate $\mu = E(Y|x) = \frac{g(x)}{g(x)+g_0(x)}$ using supervised methods (e.g. logistic regression or decision tree).
- 6 Estimate $\hat{g}(x) := g_0(x) \frac{\hat{\mu}(x)}{1-\hat{\mu}(x)}$.

Unsupervised as supervised learning

- Choice of g_0 is important.
- Sometimes g_0 is chosen to represent departures of g from g_0 . For example:
 - if departures from uniformity are of interest, $g_0(x)$ might be the uniform density over the range of the variables.
 - if departures from joint normality are of interest, a good choice for $g_0(x)$ would be a Gaussian distribution with the same mean vector and covariance matrix as the data.
 - departures from independence could be investigated by using

$$g_0(x) = \prod_{j=1}^p g_j(x_j),$$

where $g_j(x_j)$ is the marginal data density of x_j , the j th coordinate of x .

Generalized Association Rules

- The goal is to find regions of high probability.
- We can use the idea of 'unsupervised as supervised learning'.
- As reference distribution we take

$$g_0(x) = \prod_{j=1}^p g_j(x_j),$$

where $g_j(x_j)$ is the marginal data density of x_j , the j th coordinate of x . A sample from this independent density is easily generated from the data itself by applying a different random permutation to the data values of each of the variables.



Project co-financed by European Union within the framework of European Social Fund

Generalized Association Rules

- Assign $Y = 1$ to original data points and $Y = 0$ to artificial ones.
- Find regions

$$R = \bigcap_j (x_j \in s_j)$$

for which $P(Y = 1|x)$ is large with additional requirement that support is not too small.

- The regions are defined by conjunctive rules. Hence supervised methods that learn such rules would be most appropriate in this context (e.g. ???)



Project co-financed by European Union within the framework of European Social Fund