Association Rule Mining with R*

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Introduction

Association Rule Mining

Removing Redundancy

Interpreting Rules

Visualizing Association Rules

Association Rule Mining with R †

- basic concepts of association rules
- association rules mining with R
- pruning redundant rules
- interpreting and visualizing association rules
- recommended readings

[†]Chapter 9: Association Rules, *R* and *Data Mining: Examples and Case Studies*. http://www.rdatamining.com/docs/RDataMining.pdf

Association Rules

Association rules are rules presenting association or correlation between itemsets.

$$support(A \Rightarrow B) = P(A \cup B)$$

$$confidence(A \Rightarrow B) = P(B|A)$$

$$= \frac{P(A \cup B)}{P(A)}$$

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

$$= \frac{P(A \cup B)}{P(A)P(B)}$$

where P(A) is the percentage (or probability) of cases containing A.

Association Rule Mining Algorithms in R

APRIORI

- a level-wise, breadth-first algorithm which counts transactions to find frequent itemsets and then derive association rules from them
- apriori() in package arules

ECLAT

- finds frequent itemsets with equivalence classes, depth-first search and set intersection instead of counting
- eclat() in the same package

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The Titanic Dataset

- ▶ The Titanic dataset in the *datasets* package is a 4-dimensional table with summarized information on the fate of passengers on the Titanic according to social class, sex, age and survival.
- ➤ To make it suitable for association rule mining, we reconstruct the raw data as titanic.raw, where each row represents a person.
- ► The reconstructed raw data can also be downloaded at http://www.rdatamining.com/data/titanic.raw.rdata.

```
load("./data/titanic.raw.rdata")
## draw a sample of 5 records
idx <- sample(1:nrow(titanic.raw), 5)</pre>
titanic.raw[idx, ]
##
      Class
              Sex Age Survived
## 2096 2nd Female Adult
                      Yes
## 1210 Crew Male Adult No
## 329 3rd Male Adult No
## 1727 Crew Male Adult Yes
## 1305 Crew Male Adult No
summary(titanic.raw)
                           Age Survived
##
   Class
                Sex
   1st :325 Female: 470 Adult:2092 No :1490
##
##
   2nd :285 Male :1731 Child: 109 Yes: 711
## 3rd :706
## Crew:885
```

Function apriori()

Mine frequent itemsets, association rules or association hyperedges using the Apriori algorithm. The Apriori algorithm employs level-wise search for frequent itemsets.

Default settings:

- minimum support: supp=0.1
- minimum confidence: conf=0.8
- ▶ maximum length of rules: maxlen=10

```
library(arules)
rules.all <- apriori(titanic.raw)</pre>
##
## parameter specification:
  confidence minval smax arem aval originalSupport support
##
##
          0.8 0.1 1 none FALSE
                                               TRUE
                                                        0.1
## minlen maxlen target ext
##
       1 10 rules FALSE
##
## algorithmic control:
## filter tree heap memopt load sort verbose
      0.1 TRUE TRUE FALSE TRUE 2 TRUE
##
##
## apriori - find association rules with the apriori algorithm
## version 4.21 (2004.05.09) (c) 1996-2004 Christian ...
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[10 item(s), 2201 transaction(s)] done ...
## sorting and recoding items ... [9 item(s)] done [0.00s].
## creating transaction tree ... done [0.00s].
## checking subsets of size 1 2 3 4 done [0.00s].
## writing ... [27 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].
```

inspect(rules.all)

```
##
     lhs
                     rhs
                                  support confidence
                                                     lift
     {}
                   => {Age=Adult} 0.9505
                                            0.9505 1.0000
## 1
## 2 {Class=2nd}
                   => {Age=Adult} 0.1186
                                            0.9158 0.9635
                   => {Age=Adult} 0.1449
## 3 {Class=1st}
                                            0.9815 1.0327
                   => {Age=Adult} 0.1931
## 4 {Sex=Female}
                                            0.9043 0.9514
## 5 {Class=3rd}
                  => {Age=Adult} 0.2849
                                             0.8881 0.9344
## 6 {Survived=Yes} => {Age=Adult} 0.2971
                                             0.9198 0.9678
## 7 {Class=Crew}
                   => {Sex=Male} 0.3916
                                             0.9740 1.2385
## 8 {Class=Crew}
                   => {Age=Adult} 0.4021
                                             1.0000 1.0521
                   => {Sex=Male} 0.6197
## 9 {Survived=No}
                                             0.9154 1.1640
## 10 {Survived=No}
                   => {Age=Adult} 0.6533
                                             0.9651 1.0154
                   => {Age=Adult} 0.7574
## 11 {Sex=Male}
                                             0.9630 1.0132
## 12 {Sex=Female,
      Survived=Yes} => {Age=Adult}
                                   0.1436
##
                                             0.9186 0.9665
## 13 {Class=3rd,
      Sex=Male}
                   => {Survived=No} 0.1917
                                             0.8275 1.2223
##
## 14 {Class=3rd,
##
      Survived=No}
                   => {Age=Adult}
                                   0.2163
                                             0.9015 0.9485
## 15 {Class=3rd,
##
      Sex=Male}
                   => {Age=Adult} 0.2099
                                             0.9059 0.9531
## 16 {Sex=Male,
                                                           11/30
##
      Survived=Yes} => {Age=Adult}
                                   0.1536
                                             0.9210 0.9690
```

inspect(rules.sorted)

```
##
     lhs
                   rhs
                                support confidence lift
## 1 {Class=2nd,
## Age=Child} => {Survived=Yes} 0.011 1.000 3.096
## 2 {Class=2nd,
##
     Sex=Female,
## Age=Child => {Survived=Yes} 0.006
                                           1.000 3.096
## 3 {Class=1st,
##
     Sex=Female > {Survived=Yes} 0.064
                                           0.972 3.010
## 4 {Class=1st,
##
     Sex=Female,
     Age=Adult => {Survived=Yes} 0.064
                                           0.972 3.010
##
## 5 {Class=2nd,
     Sex=Female > {Survived=Yes} 0.042
                                           0.877 2.716
##
## 6 {Class=Crew.
##
     Sex=Female > {Survived=Yes} 0.009
                                           0.870 2.692
## 7 {Class=Crew.
##
     Sex=Female,
##
     Age=Adult => {Survived=Yes} 0.009
                                           0.870 2.692
## 8 {Class=2nd,
##
      Sex=Female,
##
     Age=Adult => {Survived=Yes} 0.036
                                           0.860 2.663
## 9
     {Class=2nd.
```

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Redundant Rules

- ▶ Rule #2 provides no extra knowledge in addition to rule #1, since rules #1 tells us that all 2nd-class children survived.
- When a rule (such as #2) is a super rule of another rule (#1) and the former has the same or a lower lift, the former rule (#2) is considered to be redundant.
- ▶ Other redundant rules in the above result are rules #4, #7 and #8, compared respectively with #3, #6 and #5.

Remove Redundant Rules

```
## find redundant rules
subset.matrix <- is.subset(rules.sorted, rules.sorted)
subset.matrix[lower.tri(subset.matrix, diag = T)] <- NA
redundant <- colSums(subset.matrix, na.rm = T) >= 1
```

```
## which rules are redundant
which(redundant)

## [1] 2 4 7 8

## remove redundant rules
rules.pruned <- rules.sorted[!redundant]</pre>
```

Remaining Rules

```
inspect(rules.pruned)
   lhs
##
                 rhs
                             support confidence lift
## 1 {Class=2nd,
    Age=Child => {Survived=Yes}
                              0.011
                                       1.000 3.096
##
## 2 {Class=1st,
    Sex=Female > {Survived=Yes}
                              0.064
                                       0.972 3.010
##
## 3 {Class=2nd,
    Sex=Female > {Survived=Yes}
                              0.042
                                       0.877 2.716
##
## 4 {Class=Crew,
##
    Sex=Female > {Survived=Yes}
                              0.009
                                       0.870 2.692
## 5 {Class=2nd,
##
    Sex=Male,
   0.070
##
                                       0.917 1.354
## 6 {Class=2nd,
    0.070
                                       0.860 1.271
##
## 7 {Class=3rd,
    Sex=Male,
##
## Age=Adult} => {Survived=No}
                              0.176
                                       0.838 1.237
## 8 {Class=3rd,
    0.192
                                       0.827 1.222
##
```

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Did children of the 2nd class have a higher survival rate than other children?

Did children of the 2nd class have a higher survival rate than other children?

The rule states only that all children of class 2 survived, but provides no information at all to compare the survival rates of different classes.

Rules about Children

```
rules <- apriori(titanic.raw, control = list(verbose=F),
    parameter = list(minlen=3, supp=0.002, conf=0.2),
     appearance = list(default="none", rhs=c("Survived=Yes"),
                      lhs=c("Class=1st", "Class=2nd", "Class=3rd",
                            "Age=Child", "Age=Adult")))
rules.sorted <- sort(rules, by="confidence")</pre>
inspect(rules.sorted)
## lhs
                   rhs
                                   support confidence lift
## 1 {Class=2nd,
##
    Age=Child} => {Survived=Yes} 0.010904
                                              1.0000 3.0956
## 2 {Class=1st.
##
     Age=Child => {Survived=Yes} 0.002726
                                               1.0000 3.0956
## 3 {Class=1st,
     Age=Adult } => {Survived=Yes} 0.089505
                                               0.6176 1.9117
##
## 4 {Class=2nd,
     Age=Adult => {Survived=Yes} 0.042708
                                               0.3602 1.1149
##
## 5 {Class=3rd,
     Age=Child} => {Survived=Yes} 0.012267
                                               0.3418 1.0580
##
## 6 {Class=3rd,
##
     Age=Adult >> {Survived=Yes} 0.068605
                                               0.2408 0.7455
```

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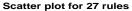
Association Rule Mining

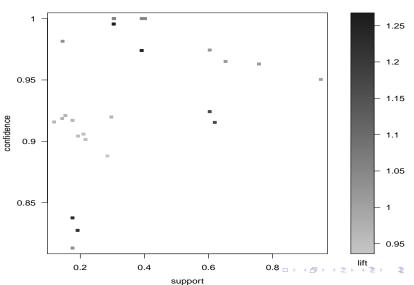
Removing Redundancy

Interpreting Rules

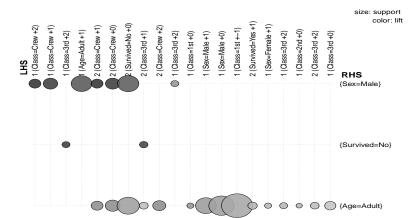
Visualizing Association Rules

library(arulesViz) plot(rules.all)





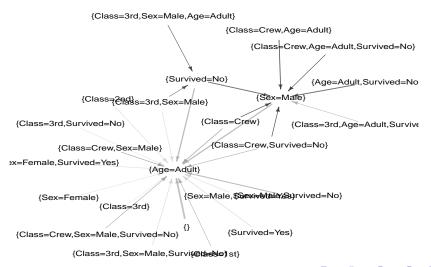
Grouped matrix for 27 rules



```
plot(rules.all, method = "graph")
```

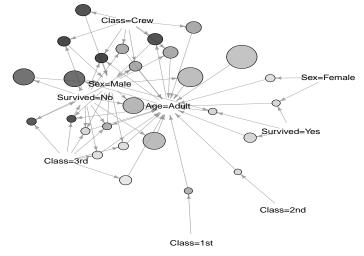
Graph for 27 rules

width: support (0.119 - 0.95) color: lift (0.934 - 1.266)



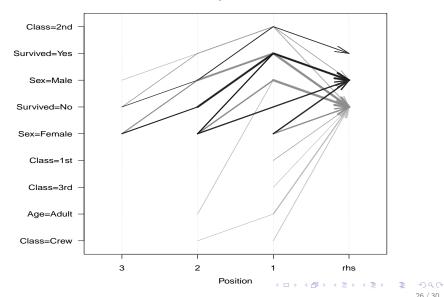
Graph for 27 rules

size: support (0.119 - 0.95) color: lift (0.934 - 1.266)



plot(rules.all, method = "paracoord", control = list(reorder = TRUE))

Parallel coordinates plot for 27 rules



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Further Readings

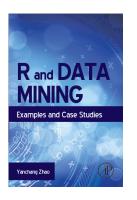
- More than 20 interestingness measures, such as chi-square, conviction, gini and leverage Tan, P.-N., Kumar, V., and Srivastava, J. (2002). Selecting the right interestingness measure for association patterns. In Proc. of KDD '02, pages 32-41, New York, NY, USA. ACM Press.
- Post mining of association rules, such as selecting interesting association rules, visualization of association rules and using association rules for classification
 Yanchang Zhao, Chengqi Zhang and Longbing Cao (Eds.). "Post-Mining of Association Rules: Techniques for Effective Knowledge Extraction", ISBN 978-1-60566-404-0, May 2009. Information Science Reference.
- ► Package arulesSequences: mining sequential patterns http://cran.r-project.org/web/packages/arulesSequences/

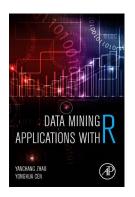
Online Resources

- Chapter 9: Association Rules, in book R and Data Mining: Examples and Case Studies http://www.rdatamining.com/docs/RDataMining.pdf
- R Reference Card for Data Mining http://www.rdatamining.com/docs/R-refcard-data-mining.pdf
- ► Free online courses and documents http://www.rdatamining.com/resources/
- RDataMining Group on LinkedIn (7,000+ members) http://group.rdatamining.com
- ► RDataMining on Twitter (1,800+ followers)

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The End





Thanks!

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