

Titanic Dataset is a dimensional table in the Datasets package, with information about the future of travelers on social class, gender, age, survival etc. according to the future of Titanic. To make it suitable for the association rule mining, we rebuild the raw information as a titanic, where each row represents a person.

Function `apriori()` Mine frequent item sets, association rules or association hyperedges using the Apriori algorithm. The Apriori algorithm employs level-wise search for frequent item sets.

Default settings:

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

```
> rules.all <- apriori(titanic)
```

```
Apriori
```

Parameter specification:

```
confidence minval smax arem aval originalSupport maxtime support minlen maxlen target
      0.8      0.1    1 none FALSE              TRUE      5      0.1      1     10 rules
ext
FALSE
```

Algorithmic control:

```
filter tree heap memopt load sort verbose
  0.1 TRUE TRUE  FALSE TRUE    2    TRUE
```

Absolute minimum support count: 88

```
set item appearances ...[0 item(s)] done [0.00s].
set transactions ...[1254 item(s), 888 transaction(s)] done [0.00s].
sorting and recoding items ... [11 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 done [0.00s].
writing ... [43 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].
```

```
> inspect(rules.sorted)
```

	lhs	rhs	support	confidence	lift	count
[1]	{Class=2nd, Age=Child}	=> {Survived=Yes}	0.011	1.000	3.096	24
[2]	{Class=2nd, Sex=Female, Age=Child}	=> {Survived=Yes}	0.006	1.000	3.096	13
[3]	{Class=1st, Sex=Female}	=> {Survived=Yes}	0.064	0.972	3.010	141
[4]	{Class=1st, Sex=Female, Age=Adult}	=> {Survived=Yes}	0.064	0.972	3.010	140
[5]	{Class=2nd, Sex=Female}	=> {Survived=Yes}	0.042	0.877	2.716	93
[6]	{Class=Crew, Sex=Female}	=> {Survived=Yes}	0.009	0.870	2.692	20
[7]	{Class=Crew, Sex=Female, Age=Adult}	=> {Survived=Yes}	0.009	0.870	2.692	20
[8]	{Class=2nd, Sex=Female, Age=Adult}	=> {Survived=Yes}	0.036	0.860	2.663	80
[9]	{Class=2nd, Sex=Male, Age=Adult}	=> {Survived=No}	0.070	0.917	1.354	154
[10]	{Class=2nd, Sex=Male}	=> {Survived=No}	0.070	0.860	1.271	154
[11]	{Class=3rd, Sex=Male, Age=Adult}	=> {Survived=No}	0.176	0.838	1.237	387
[12]	{Class=3rd, Sex=Male}	=> {Survived=No}	0.192	0.827	1.222	422

```
> inspect(rules.sorted[1:2])
```

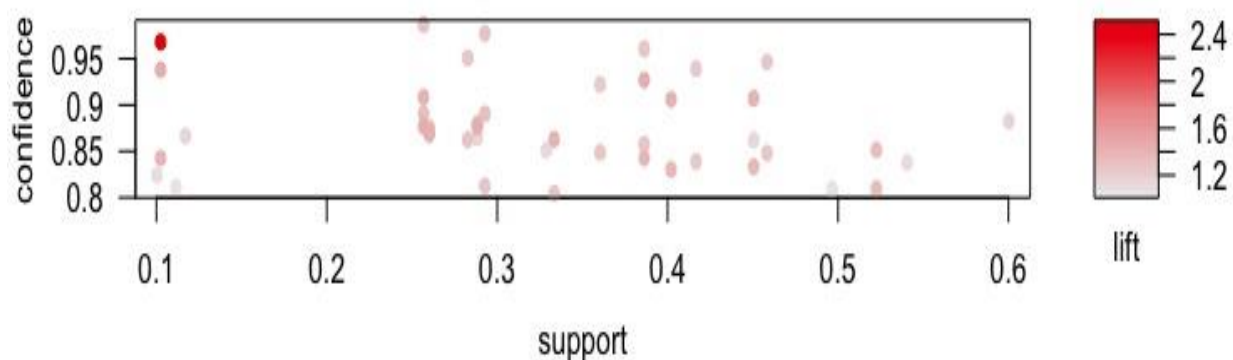
	lhs	rhs	support	confidence	lift	count
[1]	{Class=2nd, Age=Child}	=> {Survived=Yes}	0.011	1	3.096	24
[2]	{Class=2nd, Sex=Female, Age=Child}	=> {Survived=Yes}	0.006	1	3.096	13

Rule: 2 does not give any special knowledge except Rule: 1, because rules: 1 tells us that all the 3rd class children have survived. When a rule (such as: 2) is a super rule of the second rule (: 1) and the former has the same or lower lift, the former rule (: 2) is considered unnecessary. I have the comparison 4, 7 and 8 rules respectively with other unnecessary rules 3, 6 and 5 in the above results.

```
> rules_lift <- sort (rules, by="lift", decreasing=TRUE) # 'high-lift' rules.
> inspect(head(rules_lift))
```

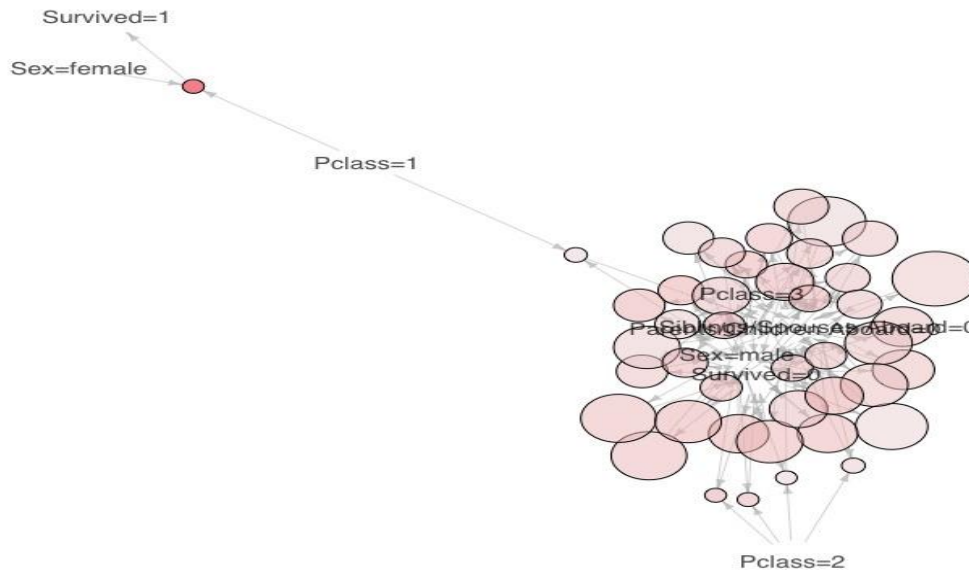
	lhs	rhs	support	confidence	lift	count
[1]	{Fare=73.5}	=> {Pclass=2}	0.006	1	4.826	5
[2]	{Fare=21}	=> {Pclass=2}	0.007	1	4.826	6
[3]	{Fare=26.25}	=> {Pclass=2}	0.007	1	4.826	6
[4]	{Fare=10.5}	=> {Pclass=2}	0.027	1	4.826	24
[5]	{Fare=13}	=> {Pclass=2}	0.047	1	4.826	42
[6]	{Survived=0, Fare=73.5}	=> {Pclass=2}	0.006	1	4.826	5

Scatter plot for 43 rules



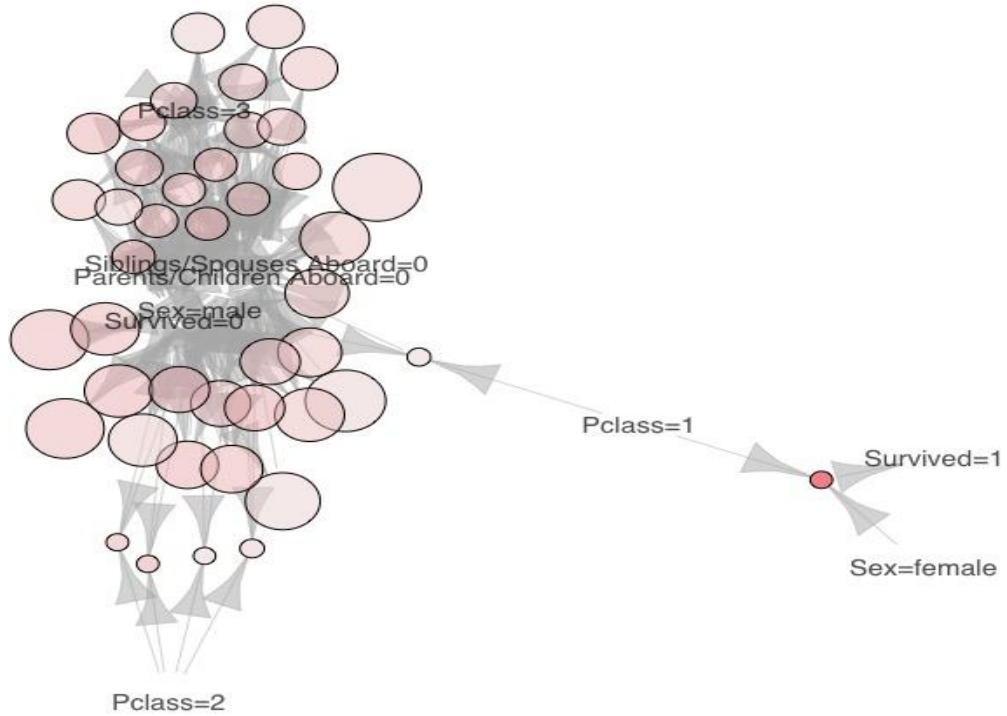
Graph for 43 rules

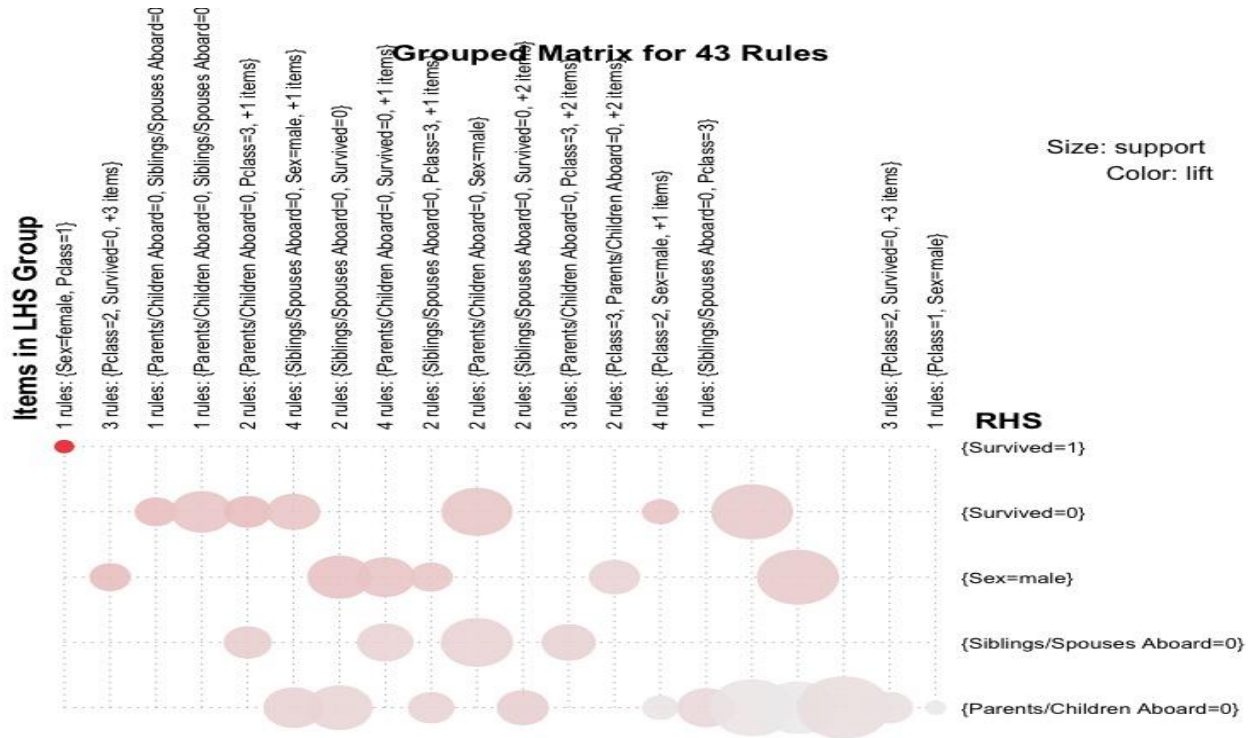
size: support (0.1 - 0.6)  
color: lift (1.066 - 2.514)



Graph for 43 rules

size: support (0.1 - 0.6)  
color: lift (1.066 - 2.514)





**Parallel coordinates plot for 43 rules**

