

```
In [1]: library(arules)

Warning message:
"package 'arules' was built under R version 3.3.3"Loading required package: Matrix

Attaching package: 'arules'

The following objects are masked from 'package:base':

    abbreviate, write
```

Preprocessing steps required for association rule mining

```
In [2]: bk <-read.csv("bankdata.csv",header = TRUE, sep = ',')
```

```
In [3]: #remove id field
bk[,1] <-NULL
```

```
In [4]: head(bk)
```

age	sex	region	income	married	children	car	save_act	current_act	mortgage	pep
54	MALE	INNER_CITY	26707.9	YES	1	NO	YES	YES	YES	YES
27	FEMALE	INNER_CITY	11604.4	YES	2	YES	YES	YES	NO	NO
42	MALE	INNER_CITY	15499.9	YES	0	YES	NO	YES	YES	YES
43	MALE	TOWN	33088.5	NO	0	NO	YES	YES	YES	NO
64	FEMALE	INNER_CITY	34513.6	YES	1	NO	YES	YES	NO	YES
43	MALE	TOWN	32395.5	YES	3	YES	YES	YES	NO	NO

```
In [5]: #discretize continuous attribute to nominal
bk[["age"]] <- ordered(cut(bk[["age"]], c(10, 40, 100)), labels = c("Young", "Senior"))
```

```
In [6]: bk[["income"]] <- discretize(bk[["income"]], categories = 2)
```

```
In [7]: #bk[["children"]] <- discretize(bk[["children"]], categories = 2)
bk[["children"]] <- ordered(cut(bk[["children"]], c(-1,0,10)), labels = c("NoChildren", "HaveChildren"))
```

```
In [8]: head(bk)
```

age	sex	region	income	married	children	car	save_act	current_act	mortgage	pep
Senior	MALE	INNER_CITY	[6294,34712)	YES	HaveChildren	NO	YES	YES	YES	YES
Young	FEMALE	INNER_CITY	[6294,34712)	YES	HaveChildren	YES	YES	YES	NO	NO
Senior	MALE	INNER_CITY	[6294,34712)	YES	NoChildren	YES	NO	YES	YES	YES
Senior	MALE	TOWN	[6294,34712)	NO	NoChildren	NO	YES	YES	YES	NO
Senior	FEMALE	INNER_CITY	[6294,34712)	YES	HaveChildren	NO	YES	YES	NO	YES
Senior	MALE	TOWN	[6294,34712)	YES	HaveChildren	YES	YES	YES	NO	NO

```
In [9]: bk[sapply(bk, is.character)] <- lapply(bk[sapply(bk, is.character)],as.factor)
```

```
In [10]: summary(bk)
```

```
      age      sex      region      income      married
Young :234  FEMALE:248  INNER_CITY:221  [ 6294,34712):361  NO :169
Senior:266   MALE  :252   RURAL      : 83  [34712,63130]:139  YES:331
              SUBURBAN : 55
              TOWN      :141

      children      car      save_act      current_act      mortgage      pep
NoChildren  :222   NO :248   NO :156   NO :124   NO :324   NO :267
HaveChildren:278  YES:252  YES:344  YES:376  YES:176  YES:233
```

```
In [11]: bk_tran <- as(bk, "transactions")
```

```
In [12]: rules = apriori(bk_tran)
```

Apriori

Parameter specification:

confidence	minval	smax	arem	aval	originalSupport	maxtime	support	minlen
0.8	0.1	1	none	FALSE	TRUE	5	0.1	1

maxlen target ext

10	rules	FALSE
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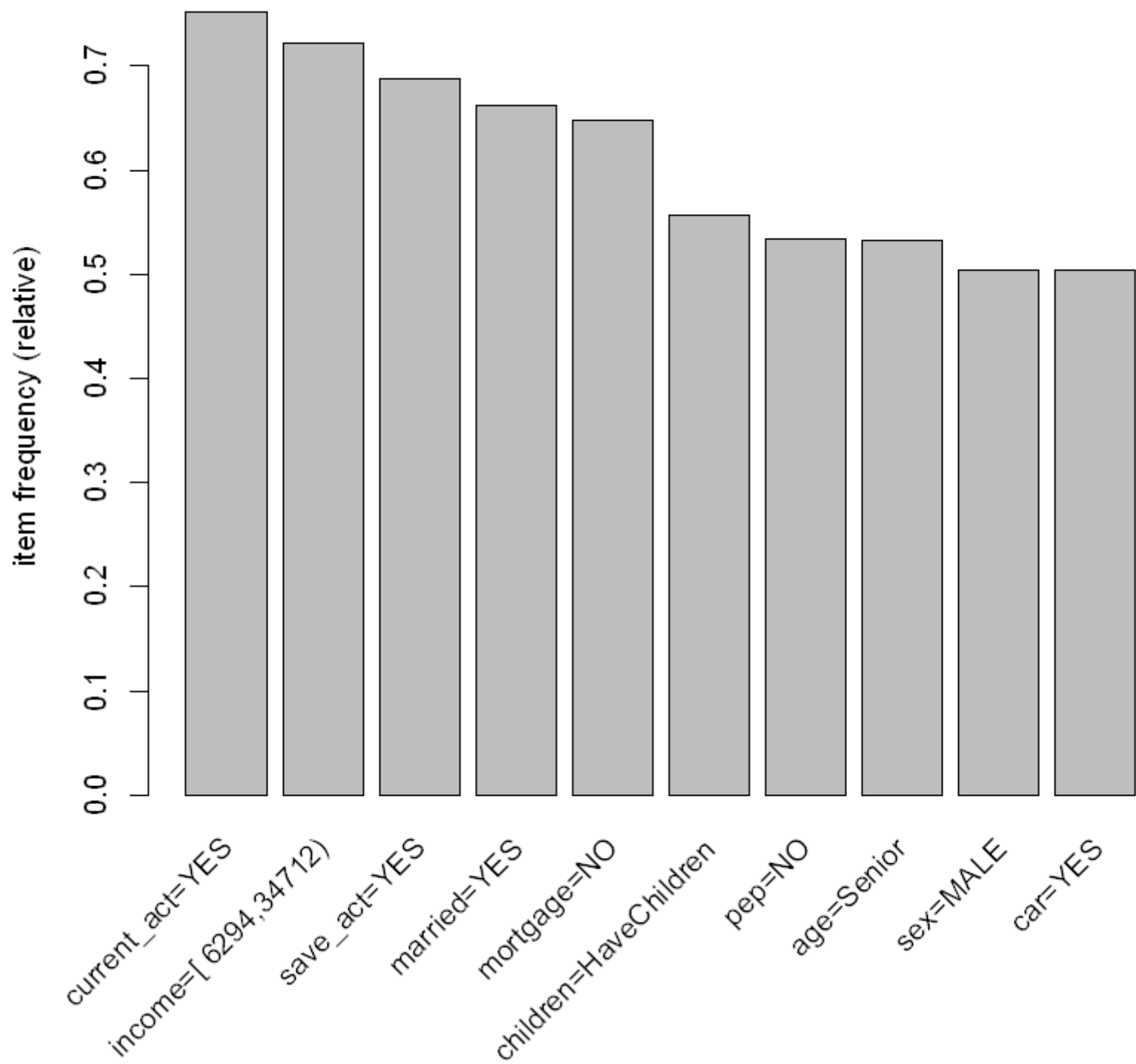
Algorithmic control:

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

Absolute minimum support count: 50

set item appearances ... [0 item(s)] done [0.00s].
set transactions ... [24 item(s), 500 transaction(s)] done [0.00s].
sorting and recoding items ... [24 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 6 done [0.01s].
writing ... [508 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].

```
In [13]: itemFrequencyPlot(bk_tran, topN = 10)
```



Experiment 0

min_suport = 0.2, confidence = 0.4, minlen = 2, pep=TRUE as RHS

```
In [14]: rules = apriori(bk_tran, parameter = list(support = 0.2, confidence = 0.4,minlen = 2))
pep_rules = subset(rules,rhs %in% "pep=YES")
#summary(pep_rules)
inspect(sort(pep_rules, by = "lift"))
```

Apriori

Parameter specification:

confidence	minval	smax	arem	aval	originalSupport	maxtime	support	minlen
0.4	0.1	1	none	FALSE	TRUE	5	0.2	2

maxlen target ext

10	rules	FALSE
----	-------	-------

Algorithmic control:

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

Absolute minimum support count: 100

set item appearances ... [0 item(s)] done [0.00s].
set transactions ... [24 item(s), 500 transaction(s)] done [0.00s].
sorting and recoding items ... [22 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 done [0.00s].
writing ... [758 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].

	lhs	rhs	support	confidence
[1]	{age=Senior, children=HaveChildren}	=> {pep=YES}	0.200	0.6622517
[2]	{married=NO}	=> {pep=YES}	0.204	0.6035503
[3]	{children=HaveChildren, save_act=YES}	=> {pep=YES}	0.228	0.5757576
[4]	{children=HaveChildren, mortgage=NO}	=> {pep=YES}	0.208	0.5621622
[5]	{children=HaveChildren, current_act=YES}	=> {pep=YES}	0.228	0.5402844
[6]	{age=Senior, current_act=YES}	=> {pep=YES}	0.214	0.5297030
[7]	{children=HaveChildren}	=> {pep=YES}	0.294	0.5287770
[8]	{age=Senior}	=> {pep=YES}	0.278	0.5225564
[9]	{age=Senior, save_act=YES}	=> {pep=YES}	0.208	0.5098039
[10]	{save_act=YES, mortgage=NO}	=> {pep=YES}	0.224	0.4977778
[11]	{sex=MALE}	=> {pep=YES}	0.244	0.4841270
[12]	{car=YES}	=> {pep=YES}	0.244	0.4841270
[13]	{current_act=YES, mortgage=NO}	=> {pep=YES}	0.242	0.4840000
[14]	{mortgage=NO}	=> {pep=YES}	0.306	0.4722222
[15]	{current_act=YES}	=> {pep=YES}	0.354	0.4707447
[16]	{region=INNER_CITY}	=> {pep=YES}	0.206	0.4660633
[17]	{save_act=YES, current_act=YES}	=> {pep=YES}	0.240	0.4545455
[18]	{save_act=YES}	=> {pep=YES}	0.308	0.4476744
[19]	{car=NO}	=> {pep=YES}	0.222	0.4475806
[20]	{sex=FEMALE}	=> {pep=YES}	0.222	0.4475806
[21]	{income=[6294, 34712], current_act=YES}	=> {pep=YES}	0.224	0.4258555
[22]	{income=[6294, 34712]}	=> {pep=YES}	0.306	0.4238227

lift

[1]	1.4211409
[2]	1.2951723
[3]	1.2355313
[4]	1.2063566
[5]	1.1594085
[6]	1.1367017
[7]	1.1347145
[8]	1.1213656
[9]	1.0939998
[10]	1.0681927
[11]	1.0388991
[12]	1.0388991
[13]	1.0386266
[14]	1.0133524
[15]	1.0101817
[16]	1.0001359
[17]	0.9754194
[18]	0.9606747
[19]	0.9604735
[20]	0.9604735
[21]	0.9138530
[22]	0.9094908

Experiment 1

min_suport = 0.2, confidence = 0.4, minlen = 2, sex=MALE & married=NO & pep=TRUE as RHS

In [15]: rules = apriori(bk_tran, parameter = list(support = 0.1, confidence = 0.4,minlen = 2))
pep_rules = subset(rules,lhs %ain% c("sex=MALE","married=NO")& rhs %in% "pep=YES")
#summary(pep_rules)
inspect(sort(pep_rules, by = "lift"))

Apriori

Parameter specification:

confidence	minval	smax	arem	aval	originalSupport	maxtime	support	minlen
0.4	0.1	1	none	FALSE	TRUE	5	0.1	2
maxlen	target	ext						
10	rules	FALSE						

Algorithmic control:

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

Absolute minimum support count: 50

set item appearances ... [0 item(s)] done [0.00s].
set transactions ... [24 item(s), 500 transaction(s)] done [0.00s].
sorting and recoding items ... [24 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 6 done [0.00s].
writing ... [5667 rule(s)] done [0.59s].
creating S4 object ... done [0.01s].

lhs	rhs	support	confidence	lift
[1] {sex=MALE,married=NO}	=> {pep=YES}	0.1	0.5952381	1.277335

Experiment 2

min_suport = 0.3, confidence = 0.4, minlen = 2, pep=TRUE as RHS

6 rules were generated

In [16]: rules = apriori(bk_tran, parameter = list(support = 0.1, confidence = 0.04,minlen = 2))
pep_rules = subset(rules, lhs %in% "age=Young" & rhs %in% "pep=YES")
#summary(pep_rules)
inspect(sort(pep_rules, by = "support"))

Apriori

Parameter specification:

confidence	minval	smax	arem	aval	originalSupport	maxtime	support	minlen
0.04	0.1	1	none	FALSE	TRUE	5	0.1	2
maxlen	target	ext						
10	rules	FALSE						

Algorithmic control:

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

Absolute minimum support count: 50

set item appearances ... [0 item(s)] done [0.00s].
set transactions ... [24 item(s), 500 transaction(s)] done [0.00s].
sorting and recoding items ... [24 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 6 done [0.01s].
writing ... [6092 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].

	lhs	rhs	support	confidence	lift
[1]	{age=Young}	=> {pep=YES}	0.188	0.4017094	0.8620373
[2]	{age=Young, income=[6294, 34712]}	=> {pep=YES}	0.176	0.3911111	0.8392942
[3]	{age=Young, current_act=YES}	=> {pep=YES}	0.140	0.4022989	0.8633023
[4]	{age=Young, income=[6294, 34712], current_act=YES}	=> {pep=YES}	0.130	0.3892216	0.8352394
[5]	{age=Young, mortgage=NO}	=> {pep=YES}	0.114	0.3851351	0.8264702
[6]	{age=Young, sex=MALE}	=> {pep=YES}	0.110	0.4365079	0.9367123
[7]	{age=Young, income=[6294, 34712], mortgage=NO}	=> {pep=YES}	0.108	0.3776224	0.8103484
[8]	{age=Young, car=NO}	=> {pep=YES}	0.100	0.4000000	0.8583691
[9]	{age=Young, save_act=YES}	=> {pep=YES}	0.100	0.3571429	0.7664010
[10]	{age=Young, sex=MALE, income=[6294, 34712]}	=> {pep=YES}	0.100	0.4237288	0.9092893

Experiment 3

min_suport = 0.1, confidence = 0.8, minlen = 2, income=[34712,63130] & pep=TRUE as RHS & lift > 1.0

10 rules were generated

```
In [17]: rules = apriori(bk_tran, parameter = list(support = 0.1, confidence = 0.8,minlen = 3))
pep_rules = subset(rules, lhs %in% "income=[34712,63130]" & rhs %in% "pep=YES" & lift > 1.0)
#summary(pep_rules)
inspect(sort(pep_rules, by = "lift"))
```

Apriori

Parameter specification:

confidence	minval	smax	arem	aval	originalSupport	maxtime	support	minlen
0.8	0.1	1	none	FALSE	TRUE	5	0.1	3
maxlen	target	ext						
10	rules	FALSE						

Algorithmic control:

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

Absolute minimum support count: 50

set item appearances ... [0 item(s)] done [0.00s].
set transactions ... [24 item(s), 500 transaction(s)] done [0.00s].
sorting and recoding items ... [24 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 6 done [0.00s].
writing ... [503 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].

	lhs	rhs	support	confidence	lift
[1]	{income=[34712,63130], children=HaveChildren, mortgage=NO}	=> {pep=YES}	0.102	0.8644068	1.854950
[2]	{income=[34712,63130], children=HaveChildren, save_act=YES, current_act=YES}	=> {pep=YES}	0.102	0.8360656	1.794132
[3]	{income=[34712,63130], children=HaveChildren, save_act=YES}	=> {pep=YES}	0.124	0.8266667	1.773963
[4]	{age=Senior, income=[34712,63130], children=HaveChildren, current_act=YES}	=> {pep=YES}	0.104	0.8253968	1.771238
[5]	{income=[34712,63130], children=HaveChildren, current_act=YES}	=> {pep=YES}	0.112	0.8235294	1.767230
[6]	{income=[34712,63130], children=HaveChildren}	=> {pep=YES}	0.134	0.8170732	1.753376
[7]	{age=Senior, income=[34712,63130], children=HaveChildren, save_act=YES}	=> {pep=YES}	0.116	0.8169014	1.753007
[8]	{age=Senior, income=[34712,63130], children=HaveChildren}	=> {pep=YES}	0.124	0.8157895	1.750621

Three interesting rules

lhs	rhs	support	confidence	lift
{sex=MALE, married=NO}	=> {pep=YES}	0.1	0.5952381	1.277335
{age=Senior, children=HaveChildren, save_act=YES}	=> {pep=YES}	0.212	0.6235294	1.3380459
{income=[34712, 63130], children=HaveChildren, mortgage=NO}	=> {pep=YES}	0.102	0.8644068	1.854950

Observations

1. From **experiment 0**, we can see that compares with female, male has more interest in the PEP product, also, senior people have higher change to buy PEP.
2. From **expriment 1**, we can see a sinlge male has significant relationship with the PEP purchase behavior.
3. From **expriment 2** and **experiment 3**, we can see that the younger people(age < 35) doesn't have much interest buying our product.
4. **Experiment 4** investigated the customers who have a higher income, since those people have much interest in purchasing the PEP product.

Recommendations

The target customer should have the following feathers: **1.single male; 2.having no mortgage; 3.having a saving account; 4.senior people who has children; 5.having higher income(income > 34712\$).**

Calculate support, confidence and lift

```
In [18]: bk2<-as.data.frame(bk)
```

```
In [19]: summary(bk2)
```

age	sex	region	income	married	
Young :234	FEMALE:248	INNER_CITY:221	[6294,34712):361	NO :169	
Senior:266	MALE :252	RURAL : 83	[34712,63130]:139	YES:331	
		SUBURBAN : 55			
		TOWN :141			
children	car	save_act	current_act	mortgage	pep
NoChildren :222	NO :248	NO :156	NO :124	NO :324	NO :267
HaveChildren:278	YES:252	YES:344	YES:376	YES:176	YES:233

```
In [20]: ## this doesnot work in jupyter notebook, but works in RStudio!  
sigma_male_marryNo <- nrow(bk2[ bk2$sex=="MALE" & bk2$married == "NO",]) # = 83  
sigma_male_marryNo_pepYes <-nrow (bk[( bk$sex=="MALE" & bk$married == "NO" & bk$pep=="YES"),]) # = 50
```

now caculate the support, confidence and lift for the following rule;

{sex=MALE,married=NO} => {pep=YES}

support = 50/500 = 0.1

confidence = 50/84 = 0.5952381

lift = (50x500)/(233x84) = 1.277335

```
In [ ]:
```