

MA Assignment 1

Netish Joseph DBA16012

```
require(conjoint)
```

```
## Loading required package: conjoint
```

```
##  
## This is package 'modeest' written by P. PONCET.  
## For a complete list of functions, use 'library(help = "modeest")' or 'help.start()'.
```

```
slevn = read.csv("D:/Business Analytics/LIBA Class Room Materials/Prof Rajesh/assignment/slevn.csv", header = TRUE)  
sprefm = read.csv("D:/Business Analytics/LIBA Class Room Materials/Prof Rajesh/assignment/sprefm.csv", header = TRUE)  
sprof = read.csv("D:/Business Analytics/LIBA Class Room Materials/Prof Rajesh/assignment/sprof.csv", header = TRUE)
```

```
slevn
```

```
##           levels  
## 1           >$250  
## 2    $100 - $250  
## 3           <$100  
## 4           large  
## 5           medium  
## 6           small  
## 7    polyester  
## 8 polycarbonate  
## 9           nylon  
## 10          yes  
## 11          no
```

```
sprof
```

##	price	size	material	sets
## 1	3	2	1	1
## 2	3	2	3	2
## 3	3	3	1	1
## 4	1	3	2	1
## 5	3	2	3	1
## 6	3	1	1	1
## 7	1	2	3	2
## 8	2	1	2	1
## 9	3	2	3	1
## 10	3	1	3	1

sprefm

##	profil1	profil2	profil3	profil4	profil5	profil6	profil7	profil8
## 1	10	6	5	3	8	10	7	1
## 2	1	5	6	4	3	6	10	8
## 3	8	10	8	6	7	4	3	2
## 4	9	2	3	1	2	10	2	2
## 5	5	5	10	10	2	3	4	4
## 6	6	7	4	8	8	6	10	4
## 7	1	4	9	7	1	5	4	7
## 8	2	6	10	1	4	1	2	8
## 9	1	4	9	1	4	5	1	2
## 10	6	2	2	10	1	5	7	3
## 11	7	10	9	5	2	6	3	2
## 12	10	2	2	4	7	7	6	1
## 13	4	7	7	4	9	10	4	4
## 14	7	5	3	5	8	3	6	4
## 15	8	8	6	2	9	2	9	1
## 16	8	5	7	4	8	4	5	5
## 17	3	8	1	1	7	2	6	5
## 18	10	2	3	4	2	9	2	4
## 19	6	2	7	3	6	1	1	5
## 20	10	6	2	4	9	1	1	2
## 21	1	8	6	2	1	3	2	3
## 22	3	4	1	9	4	2	9	4
## 23	4	1	5	7	1	8	5	4
## 24	4	7	2	3	10	7	7	2
## 25	8	1	1	9	5	10	6	2
## 26	6	5	6	10	10	2	9	10
## 27	9	10	4	3	6	8	1	9
## 28	3	6	4	6	6	9	3	5
## 29	10	5	3	1	7	9	4	2
## 30	7	8	7	10	8	10	1	8
## 31	8	1	4	10	4	9	7	9
## 32	6	1	10	3	8	6	4	9
## 33	3	3	2	4	9	3	2	5
## 34	10	4	3	10	1	2	1	3
## 35	7	7	3	9	4	10	6	10
## 36	9	3	3	1	10	7	6	4
## 37	8	2	7	6	1	1	7	8
## 38	10	2	1	2	6	1	4	10
## 39	10	2	8	9	8	7	4	2
## 40	8	6	9	2	8	2	9	2
## 41	8	4	5	2	8	9	4	1
## 42	3	10	4	9	10	9	9	9
## 43	8	8	1	1	5	2	10	10
## 44	9	1	5	5	7	3	2	3
## 45	9	9	3	4	8	8	4	3
## 46	4	9	1	10	2	3	6	9
## 47	6	4	3	4	1	4	4	2
## 48	1	3	8	2	8	1	7	3
## 49	6	4	2	5	10	6	2	10
## 50	4	10	7	2	10	3	5	3
## 51	6	9	2	1	6	5	9	4
## 52	3	3	6	10	7	8	8	9

## 53	6	8	1	1	8	4	8	4
## 54	1	9	4	10	10	8	2	5
## 55	3	2	4	2	7	5	2	1
## 56	6	1	4	10	5	2	6	1
## 57	8	3	6	8	5	4	5	2
## 58	3	9	5	7	6	2	6	8
## 59	4	5	4	3	4	2	4	2
## 60	2	3	9	1	3	4	5	6
## 61	8	6	4	7	7	10	10	9
## 62	1	4	1	6	9	9	8	3
## 63	6	5	3	3	9	2	1	10
## 64	4	10	9	2	9	5	8	9
## 65	4	4	6	1	6	4	6	10
## 66	6	4	8	10	10	5	2	3
## 67	6	2	8	9	6	1	3	7
## 68	6	8	1	8	2	10	6	4
## 69	8	6	10	3	4	1	5	7
## 70	6	10	7	5	9	4	7	1
## 71	7	3	8	8	3	10	7	10
## 72	4	3	2	8	1	4	4	10
## 73	6	10	8	5	4	8	6	8
## 74	2	8	5	6	1	5	1	2
## 75	2	7	1	4	9	5	7	3
## 76	5	8	4	5	7	6	4	4
## 77	2	10	10	8	1	4	4	2
## 78	3	3	7	4	9	8	1	9
## 79	5	9	7	8	8	9	10	6
## 80	3	1	1	9	10	6	10	4
## 81	1	9	3	1	10	2	4	7
## 82	4	6	9	5	1	7	5	1
## 83	9	5	6	9	5	9	7	10
## 84	6	2	2	6	9	3	3	1
## 85	2	9	3	3	5	1	5	5
## 86	3	1	5	3	1	1	2	4
## 87	1	9	4	3	9	2	5	10
## 88	6	10	2	1	3	6	10	9
## 89	8	9	8	1	9	6	9	1
## 90	8	4	6	7	7	6	3	1
## 91	5	2	2	1	5	10	7	10
## 92	1	1	8	3	5	9	7	1
## 93	9	5	2	4	6	7	6	8
## 94	4	6	3	7	6	4	7	1
## 95	10	7	5	10	7	2	5	2
## 96	6	7	8	6	1	1	10	8
## 97	10	7	7	4	4	3	1	9
## 98	7	9	8	10	10	8	6	3
## 99	4	7	5	8	9	5	2	8
## 100	2	8	9	2	9	10	4	4
## 101	5	10	8	2	6	2	10	7
## 102	9	4	7	7	3	2	2	1
## 103	9	2	2	4	2	9	2	10
## 104	6	9	7	9	1	10	7	9
## 105	4	6	3	4	6	6	6	5
## 106	10	9	4	2	5	7	5	5

## 107	7	9	2	9	2	6	6	3
## 108	9	2	10	1	8	5	9	1
## 109	8	8	2	9	7	1	7	2
## 110	7	6	2	10	9	5	4	2
## 111	10	4	4	7	8	2	1	2
## 112	6	4	1	1	3	6	1	9
## 113	8	4	6	7	10	5	5	5
## 114	6	6	7	4	1	5	6	9
## 115	4	9	9	5	5	2	2	10
## 116	7	9	7	6	3	4	6	1
## 117	6	7	3	10	4	4	5	6
## 118	5	7	5	2	2	8	5	1
## 119	9	4	9	5	2	2	10	6
## 120	2	4	2	9	1	10	7	5
## 121	6	4	2	1	10	3	2	3
## 122	5	3	7	1	7	8	7	2
## 123	8	7	4	10	5	1	9	8
## 124	1	6	2	3	2	7	3	1
## 125	10	4	9	3	5	10	9	6
## 126	10	2	1	7	2	3	8	4
## 127	4	4	3	7	5	10	1	8
## 128	2	3	1	7	9	5	10	1
## 129	9	1	6	1	9	7	10	9
## 130	4	5	7	3	3	7	2	2
## 131	3	8	5	5	5	5	3	7
## 132	4	2	6	2	1	5	1	9
## 133	8	8	7	10	5	7	2	7
## 134	5	6	5	8	8	6	1	1
## 135	10	10	7	7	1	6	9	2
## 136	2	10	7	2	4	3	6	6
## 137	9	1	7	6	2	3	2	4
## 138	4	9	1	6	1	8	7	8
## 139	3	5	10	6	7	8	9	7
## 140	9	6	9	6	5	4	6	6
## 141	3	4	5	10	1	10	6	4
## 142	10	7	2	6	2	3	10	3
## 143	9	5	2	2	10	4	6	8
## 144	4	7	1	2	9	1	4	8
## 145	10	10	8	8	6	10	6	5
## 146	9	8	5	5	10	4	2	3
## 147	2	8	1	6	8	1	3	10
## 148	1	7	2	1	3	8	9	5
## 149	4	7	4	8	6	5	5	2
## 150	8	2	3	2	10	10	1	8
## 151	6	10	7	3	4	1	5	10
## 152	3	2	6	4	1	6	6	3
## 153	6	6	2	2	5	1	6	8
## 154	8	6	2	5	3	4	10	8
## 155	1	4	9	3	2	2	1	5
## 156	6	5	6	10	10	3	10	5
## 157	10	8	6	1	5	9	3	4
## 158	2	2	4	7	10	5	2	5
## 159	7	5	1	10	2	2	3	3
## 160	6	4	2	6	6	9	8	5

## 161	10	1	6	9	8	9	6	7
## 162	6	7	4	9	1	2	6	8
## 163	5	7	6	3	4	2	3	1
## 164	10	4	2	2	8	8	8	7
## 165	8	3	5	1	8	5	8	6
## 166	6	10	8	10	5	5	5	8
## 167	3	9	10	7	6	4	8	7
## 168	5	2	4	8	2	9	7	9
## 169	4	4	5	10	4	2	10	2
## 170	10	4	4	7	7	8	2	6
## 171	9	5	6	3	4	3	8	4
## 172	1	4	6	2	10	1	7	7
## 173	6	8	5	1	2	10	2	7
## 174	5	10	8	3	8	10	6	7
## 175	10	8	9	8	4	8	9	3
## 176	5	3	2	1	6	6	10	2
## 177	7	9	5	2	6	2	8	7
## 178	10	8	8	3	7	6	3	6
## 179	5	3	5	10	1	4	6	2
## 180	8	7	7	10	8	5	1	7
## 181	3	2	5	6	10	9	10	6
## 182	10	8	4	9	10	6	7	2
## 183	5	4	9	3	7	6	1	3
## 184	4	8	4	3	4	2	5	10
## 185	1	10	7	1	3	6	4	9
## 186	5	3	10	4	7	7	5	5
## 187	9	4	10	9	10	4	6	4
## 188	4	9	5	2	3	8	10	4
## 189	9	2	2	9	2	4	8	6
## 190	10	7	4	10	3	3	1	8
## 191	10	10	9	1	9	3	3	10
## 192	9	7	3	2	10	6	3	10
## 193	8	9	6	8	8	6	9	5
## 194	8	7	5	1	5	5	4	5
## 195	8	9	7	6	8	7	7	5
## 196	3	8	1	4	1	7	2	9
## 197	1	9	10	4	3	4	8	10
## 198	2	6	10	6	6	1	7	1
## 199	3	9	10	5	10	6	3	5
## 200	5	7	5	8	6	5	10	8

##	profil9	profil10
----	---------	----------

## 1	1	10
## 2	4	4
## 3	9	9
## 4	9	5
## 5	5	4
## 6	2	7
## 7	10	3
## 8	4	3
## 9	10	3
## 10	1	7
## 11	5	4
## 12	8	10
## 13	7	6

## 14	9	7
## 15	6	1
## 16	10	3
## 17	1	3
## 18	5	7
## 19	2	8
## 20	7	4
## 21	7	5
## 22	3	3
## 23	9	5
## 24	8	5
## 25	8	2
## 26	9	8
## 27	7	10
## 28	1	8
## 29	4	8
## 30	9	3
## 31	4	6
## 32	1	5
## 33	3	8
## 34	4	1
## 35	8	8
## 36	1	3
## 37	7	2
## 38	10	10
## 39	6	9
## 40	9	8
## 41	2	2
## 42	10	9
## 43	4	5
## 44	1	4
## 45	9	8
## 46	5	5
## 47	9	4
## 48	9	7
## 49	5	6
## 50	8	10
## 51	9	4
## 52	1	1
## 53	1	10
## 54	7	7
## 55	4	9
## 56	5	3
## 57	5	4
## 58	3	6
## 59	6	1
## 60	2	1
## 61	5	6
## 62	9	3
## 63	6	6
## 64	1	7
## 65	10	5
## 66	1	4
## 67	5	1

## 68	10	7
## 69	5	3
## 70	10	6
## 71	3	1
## 72	5	6
## 73	1	2
## 74	7	2
## 75	1	6
## 76	2	5
## 77	5	3
## 78	6	4
## 79	9	8
## 80	7	2
## 81	5	4
## 82	3	3
## 83	9	10
## 84	7	4
## 85	2	3
## 86	5	5
## 87	4	6
## 88	10	10
## 89	10	3
## 90	9	8
## 91	6	3
## 92	6	2
## 93	6	6
## 94	10	7
## 95	10	6
## 96	10	2
## 97	3	9
## 98	1	8
## 99	6	10
## 100	2	1
## 101	10	8
## 102	2	1
## 103	2	2
## 104	4	10
## 105	5	5
## 106	7	5
## 107	7	5
## 108	9	3
## 109	9	10
## 110	2	2
## 111	6	10
## 112	9	7
## 113	3	4
## 114	10	1
## 115	10	5
## 116	6	9
## 117	2	8
## 118	10	8
## 119	10	7
## 120	4	8
## 121	2	4

## 122	7	2
## 123	4	3
## 124	8	7
## 125	7	5
## 126	1	9
## 127	8	6
## 128	3	4
## 129	1	3
## 130	8	3
## 131	6	10
## 132	7	6
## 133	8	8
## 134	10	6
## 135	10	1
## 136	10	7
## 137	2	7
## 138	10	9
## 139	2	2
## 140	4	7
## 141	4	3
## 142	4	5
## 143	6	2
## 144	2	6
## 145	1	10
## 146	4	8
## 147	10	2
## 148	1	4
## 149	8	10
## 150	4	7
## 151	5	8
## 152	2	6
## 153	6	4
## 154	2	2
## 155	4	3
## 156	2	2
## 157	5	5
## 158	9	5
## 159	2	3
## 160	5	5
## 161	4	1
## 162	7	1
## 163	9	4
## 164	4	7
## 165	7	9
## 166	1	2
## 167	10	3
## 168	1	8
## 169	2	1
## 170	5	3
## 171	7	8
## 172	10	8
## 173	6	10
## 174	10	1
## 175	8	9

```
## 176      4      3
## 177      8      7
## 178     10      3
## 179      6      5
## 180      7      2
## 181      5      5
## 182     10      5
## 183      6      7
## 184      7      7
## 185      1      1
## 186      5      4
## 187      5      4
## 188     10      3
## 189      1      1
## 190      1      3
## 191      3      6
## 192      1      7
## 193      1      4
## 194     10      5
## 195      7      9
## 196      7      2
## 197      9      2
## 198      3     10
## 199      2      7
## 200      3      2
```

```
caModel(y=spre_fm[1,], x=sprof) # all attribute levels with intercept
```

```
##
## Call:
## lm(formula = frml)
##
## Residuals:
##      1      2      3      4      5      6
## 1.571e+00 -2.921e-16 1.262e-15 -5.697e-16 2.714e+00 -1.571e+00
##      7      8      9     10
## 5.406e-16 1.520e-16 -4.286e+00 1.571e+00
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      4.45238    2.48624   1.791   0.215
## factor(x$price)1    3.19048    3.24369   0.984   0.429
## factor(x$price)2   -5.38095    5.57123  -0.966   0.436
## factor(x$size)1     3.23810    2.51255   1.289   0.326
## factor(x$size)2     0.09524    2.51255   0.038   0.973
## factor(x$material)1 2.04762    2.99054   0.685   0.564
## factor(x$material)2 -0.95238    5.57123  -0.171   0.880
## factor(x$sets)1    -0.35714    2.43277  -0.147   0.897
##
## Residual standard error: 4.071 on 2 degrees of freedom
## Multiple R-squared:  0.7064, Adjusted R-squared:  -0.321
## F-statistic: 0.6876 on 7 and 2 DF,  p-value: 0.7037
```

```
Conjoint(y=sprefm, x=sprof, z=slevn) # parameters
```

```
##
## Call:
## lm(formula = frml)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4,8186 -2,3550 -0,1686  2,3393  4,8700
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      5,26488    0,12450  42,289  < 2e-16 ***
## factor(x$price)1  -0,08952    0,16243  -0,551  0,58158
## factor(x$price)2  -0,20595    0,27898  -0,738  0,46046
## factor(x$size)1   -0,09857    0,12582  -0,783  0,43345
## factor(x$size)2    0,39357    0,12582   3,128  0,00178 **
## factor(x$material)1 -0,09571    0,14975  -0,639  0,52279
## factor(x$material)2  0,34929    0,27898   1,252  0,21071
## factor(x$sets)1    -0,03964    0,12182  -0,325  0,74490
## ---
## Signif. codes:  0 '***' 0,001 '**' 0,01 '*' 0,05 '.' 0,1 ' ' 1
##
## Residual standard error: 2,883 on 1992 degrees of freedom
## Multiple R-squared:  0,007332, Adjusted R-squared:  0,003844
## F-statistic: 2,102 on 7 and 1992 DF, p-value: 0,04035
```

```
## [1] "Part worths (utilities) of levels (model parameters for whole sample):"
##           levnms      utls
## 1      intercept  5,2649
## 2          >$250 -0,0895
## 3    $100 - $250 -0,206
## 4          <$100  0,2955
## 5          large -0,0986
## 6          medium  0,3936
## 7           small -0,295
## 8    polyester -0,0957
## 9 polycarbonate  0,3493
## 10          nylon -0,2536
## 11           yes -0,0396
## 12           no  0,0396
## [1] "Average importance of factors (attributes):"
## [1] 35,50 23,14 26,86 14,50
## [1] Sum of average importance: 100
## [1] "Chart of average factors importance"
```

```
caTotalUtilities(y=sprefm, x=sprof) # compare respondents
```

##		[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]
##	[1,]	8.429	6	5	3	5.286	11.571	7	1	5.286	8.429
##	[2,]	2.286	5	6	4	2.857	4.714	10	8	2.857	5.286
##	[3,]	6.571	10	8	6	8.714	5.429	3	2	8.714	7.571
##	[4,]	9.429	2	3	1	5.286	9.571	2	2	5.286	5.429
##	[5,]	4.286	5	10	10	3.857	3.714	4	4	3.857	3.286
##	[6,]	5.429	7	4	8	5.286	6.571	10	4	5.286	6.429
##	[7,]	2.857	4	9	7	4.571	3.143	4	7	4.571	4.857
##	[8,]	2.000	6	10	1	4.000	1.000	2	8	4.000	3.000
##	[9,]	3.286	4	9	1	5.857	2.714	1	2	5.857	5.286
##	[10,]	4.000	2	2	10	2.000	7.000	7	3	2.000	5.000
##	[11,]	6.571	10	9	5	3.714	6.429	3	2	3.714	3.571
##	[12,]	8.429	2	2	4	8.286	8.571	6	1	8.286	8.429
##	[13,]	6.286	7	7	4	6.857	7.714	4	4	6.857	8.286
##	[14,]	6.286	5	3	5	8.857	3.714	6	4	8.857	6.286
##	[15,]	8.143	8	6	2	7.429	1.857	9	1	7.429	1.143
##	[16,]	8.571	5	7	4	8.714	3.429	5	5	8.714	3.571
##	[17,]	3.000	8	1	1	4.000	2.000	6	5	4.000	3.000
##	[18,]	8.714	2	3	4	4.143	10.286	2	4	4.143	5.714
##	[19,]	3.429	2	7	3	5.286	3.571	1	5	5.286	5.429
##	[20,]	8.571	6	2	4	8.714	2.429	1	2	8.714	2.571
##	[21,]	1.286	8	6	2	3.857	2.714	2	3	3.857	5.286
##	[22,]	2.857	4	1	9	3.571	2.143	9	4	3.571	2.857
##	[23,]	5.143	1	5	7	4.429	6.857	5	4	4.429	6.143
##	[24,]	6.000	7	2	3	8.000	5.000	7	2	8.000	7.000
##	[25,]	9.857	1	1	9	5.571	8.143	6	2	5.571	3.857
##	[26,]	5.286	5	6	10	9.857	2.714	9	10	9.857	7.286
##	[27,]	7.714	10	4	3	7.143	9.286	1	9	7.143	8.714
##	[28,]	3.429	6	4	6	3.286	8.571	3	5	3.286	8.429
##	[29,]	9.000	5	3	1	6.000	10.000	4	2	6.000	7.000
##	[30,]	9.429	8	7	10	7.286	7.571	1	8	7.286	5.429
##	[31,]	7.714	1	4	10	4.143	9.286	7	9	4.143	5.714
##	[32,]	5.857	1	10	3	4.571	6.143	4	9	4.571	4.857
##	[33,]	2.429	3	2	4	6.286	3.571	2	5	6.286	7.429
##	[34,]	8.143	4	3	10	3.429	3.857	1	3	3.429	-0.857
##	[35,]	7.286	7	3	9	5.857	9.714	6	10	5.857	8.286
##	[36,]	9.143	3	3	1	5.429	6.857	6	4	5.429	3.143
##	[37,]	6.571	2	7	6	4.714	2.429	7	8	4.714	0.571
##	[38,]	6.857	2	1	2	9.571	4.143	4	10	9.571	6.857
##	[39,]	8.571	2	8	9	7.714	8.429	4	2	7.714	7.571
##	[40,]	6.429	6	9	2	9.286	3.571	9	2	9.286	6.429
##	[41,]	9.143	4	5	2	4.429	7.857	4	1	4.429	3.143
##	[42,]	5.000	10	4	9	9.000	7.000	9	9	9.000	11.000
##	[43,]	6.143	8	1	1	5.429	3.857	10	10	5.429	3.143
##	[44,]	7.286	1	5	5	4.857	4.714	2	3	4.857	2.286
##	[45,]	8.857	9	3	4	8.571	8.143	4	3	8.571	7.857
##	[46,]	3.286	9	1	10	3.857	3.714	6	9	3.857	4.286
##	[47,]	5.714	4	3	4	5.143	4.286	4	2	5.143	3.714
##	[48,]	1.429	3	8	2	8.286	0.571	7	3	8.286	7.429
##	[49,]	6.429	4	2	5	7.286	5.571	2	10	7.286	6.429
##	[50,]	3.429	10	7	2	9.286	3.571	5	3	9.286	9.429
##	[51,]	6.714	9	2	1	7.143	4.286	9	4	7.143	4.714
##	[52,]	5.286	3	6	10	2.857	5.714	8	9	2.857	3.286

##	[53,]	3.857	8	1	1	5.571	6.143	8	4	5.571	7.857
##	[54,]	3.429	9	4	10	7.286	5.571	2	5	7.286	9.429
##	[55,]	2.571	2	4	2	5.714	5.429	2	1	5.714	8.571
##	[56,]	5.429	1	4	10	5.286	2.571	6	1	5.286	2.429
##	[57,]	7.143	3	6	8	5.429	4.857	5	2	5.429	3.143
##	[58,]	2.286	9	5	7	4.857	2.714	6	8	4.857	5.286
##	[59,]	4.571	5	4	3	4.714	1.429	4	2	4.714	1.571
##	[60,]	3.000	3	9	1	2.000	3.000	5	6	2.000	2.000
##	[61,]	8.571	6	4	7	5.714	9.429	10	9	5.714	6.571
##	[62,]	5.000	4	1	6	7.000	5.000	8	3	7.000	7.000
##	[63,]	5.286	5	3	3	7.857	2.714	1	10	7.857	5.286
##	[64,]	3.714	10	9	2	5.143	5.286	8	9	5.143	6.714
##	[65,]	4.857	4	6	1	7.571	3.143	6	10	7.571	5.857
##	[66,]	6.143	4	8	10	5.429	4.857	2	3	5.429	4.143
##	[67,]	5.857	2	8	9	5.571	1.143	3	7	5.571	0.857
##	[68,]	6.857	8	1	8	5.571	9.143	6	4	5.571	7.857
##	[69,]	6.429	6	10	3	5.286	2.571	5	7	5.286	1.429
##	[70,]	6.429	10	7	5	9.286	3.571	7	1	9.286	6.429
##	[71,]	8.429	3	8	8	2.286	8.571	7	10	2.286	2.429
##	[72,]	3.143	3	2	8	3.429	4.857	4	10	3.429	5.143
##	[73,]	6.714	10	8	5	2.143	7.286	6	8	2.143	2.714
##	[74,]	3.429	8	5	6	3.286	3.571	1	2	3.286	3.429
##	[75,]	2.571	7	1	4	4.714	4.429	7	3	4.714	6.571
##	[76,]	5.143	8	4	5	4.429	5.857	4	4	4.429	5.143
##	[77,]	2.571	10	10	8	2.714	3.429	4	2	2.714	3.571
##	[78,]	5.429	3	7	4	6.286	5.571	1	9	6.286	6.429
##	[79,]	6.286	9	7	8	7.857	7.714	10	6	7.857	9.286
##	[80,]	5.714	1	1	9	7.143	3.286	10	4	7.143	4.714
##	[81,]	2.286	9	3	1	6.857	0.714	4	7	6.857	5.286
##	[82,]	4.571	6	9	5	1.714	6.429	5	1	1.714	3.571
##	[83,]	8.143	5	6	9	7.429	9.857	7	10	7.429	9.143
##	[84,]	6.286	2	2	6	7.857	2.714	3	1	7.857	4.286
##	[85,]	1.857	9	3	3	3.571	1.143	5	5	3.571	2.857
##	[86,]	1.857	1	5	3	3.571	2.143	2	4	3.571	3.857
##	[87,]	1.429	9	4	3	6.286	1.571	5	10	6.286	6.429
##	[88,]	5.000	10	2	1	7.000	7.000	10	9	7.000	9.000
##	[89,]	9.286	9	8	1	8.857	4.714	9	1	8.857	4.286
##	[90,]	7.429	4	6	7	8.286	6.571	3	1	8.286	7.429
##	[91,]	7.143	2	2	1	4.429	7.857	7	10	4.429	5.143
##	[92,]	4.286	1	8	3	3.857	5.714	7	1	3.857	5.286
##	[93,]	8.429	5	2	4	6.286	7.571	6	8	6.286	5.429
##	[94,]	4.286	6	3	7	7.857	3.714	7	1	7.857	7.286
##	[95,]	8.429	7	5	10	9.286	3.571	5	2	9.286	4.429
##	[96,]	5.571	7	8	6	5.714	1.429	10	8	5.714	1.571
##	[97,]	6.429	7	7	4	5.286	6.571	1	9	5.286	5.429
##	[98,]	6.571	9	8	10	5.714	8.429	6	3	5.714	7.571
##	[99,]	3.571	7	5	8	7.714	5.429	2	8	7.714	9.571
##	[100,]	5.571	8	9	2	3.714	6.429	4	4	3.714	4.571
##	[101,]	4.143	10	8	2	8.429	2.857	10	7	8.429	7.143
##	[102,]	7.429	4	7	7	3.286	3.571	2	1	3.286	-0.571
##	[103,]	9.000	2	2	4	2.000	9.000	2	10	2.000	2.000
##	[104,]	5.000	9	7	9	3.000	11.000	7	9	3.000	9.000
##	[105,]	4.714	6	3	4	5.143	5.286	6	5	5.143	5.714
##	[106,]	9.429	9	4	2	6.286	7.571	5	5	6.286	4.429

## [107,]	6.571	9	2	9	4.714	6.429	6	3	4.714	4.571
## [108,]	9.429	2	10	1	8.286	4.571	9	1	8.286	3.429
## [109,]	5.429	8	2	9	9.286	3.571	7	2	9.286	7.429
## [110,]	7.429	6	2	10	5.286	4.571	4	2	5.286	2.429
## [111,]	6.857	4	4	7	8.571	5.143	1	2	8.571	6.857
## [112,]	5.714	4	1	1	6.143	6.286	1	9	6.143	6.714
## [113,]	7.857	4	6	7	6.571	5.143	5	5	6.571	3.857
## [114,]	7.000	6	7	4	5.000	4.000	6	9	5.000	2.000
## [115,]	4.143	9	9	5	7.429	1.857	2	10	7.429	5.143
## [116,]	4.857	9	7	6	5.571	6.143	6	1	5.571	6.857
## [117,]	4.000	7	3	10	4.000	6.000	5	6	4.000	6.000
## [118,]	5.286	7	5	2	5.857	7.714	5	1	5.857	8.286
## [119,]	6.714	4	9	5	7.143	4.286	10	6	7.143	4.714
## [120,]	2.714	4	2	9	2.143	9.286	7	5	2.143	8.714
## [121,]	5.714	4	2	1	6.143	3.286	2	3	6.143	3.714
## [122,]	7.286	3	7	1	5.857	5.714	7	2	5.857	4.286
## [123,]	6.429	7	4	10	5.286	2.571	9	8	5.286	1.429
## [124,]	2.143	6	2	3	4.429	5.857	3	1	4.429	8.143
## [125,]	10.286	4	9	3	5.857	9.714	9	6	5.857	5.286
## [126,]	5.857	2	1	7	3.571	7.143	8	4	3.571	4.857
## [127,]	5.857	4	3	7	5.571	8.143	1	8	5.571	7.857
## [128,]	3.429	3	1	7	5.286	3.571	10	1	5.286	5.429
## [129,]	9.000	1	6	1	5.000	7.000	10	9	5.000	3.000
## [130,]	5.571	5	7	3	4.714	5.429	2	2	4.714	4.571
## [131,]	2.286	8	5	5	5.857	5.714	3	7	5.857	9.286
## [132,]	3.714	2	6	2	4.143	5.286	1	9	4.143	5.714
## [133,]	7.286	8	7	10	6.857	7.714	2	7	6.857	7.286
## [134,]	6.143	6	5	8	8.429	4.857	1	1	8.429	7.143
## [135,]	10.143	10	7	7	5.429	5.857	9	2	5.429	1.143
## [136,]	2.286	10	7	2	6.857	2.714	6	6	6.857	7.286
## [137,]	5.857	1	7	6	3.571	6.143	2	4	3.571	3.857
## [138,]	4.143	9	1	6	5.429	7.857	7	8	5.429	9.143
## [139,]	5.143	5	10	6	3.429	5.857	9	7	3.429	4.143
## [140,]	6.857	6	9	6	5.571	6.143	6	6	5.571	4.857
## [141,]	4.857	4	5	10	1.571	8.143	6	4	1.571	4.857
## [142,]	7.429	7	2	6	4.286	5.571	10	3	4.286	2.429
## [143,]	9.286	5	2	2	7.857	3.714	6	8	7.857	2.286
## [144,]	3.000	7	1	2	6.000	2.000	4	8	6.000	5.000
## [145,]	8.143	10	8	8	4.429	11.857	6	5	4.429	8.143
## [146,]	7.286	8	5	5	7.857	5.714	2	3	7.857	6.286
## [147,]	3.714	8	1	6	8.143	-0.714	3	10	8.143	3.714
## [148,]	2.429	7	2	1	1.286	6.571	9	5	1.286	5.429
## [149,]	3.429	7	4	8	7.286	5.571	5	2	7.286	9.429
## [150,]	8.571	2	3	2	6.714	9.429	1	8	6.714	7.571
## [151,]	3.571	10	7	3	5.714	3.429	5	10	5.714	5.571
## [152,]	2.571	2	6	4	1.714	6.429	6	3	1.714	5.571
## [153,]	5.000	6	2	2	6.000	2.000	6	8	6.000	3.000
## [154,]	7.000	6	2	5	3.000	5.000	10	8	3.000	1.000
## [155,]	1.286	4	9	3	2.857	1.714	1	5	2.857	3.286
## [156,]	6.286	5	6	10	5.857	2.714	10	5	5.857	2.286
## [157,]	9.714	8	6	1	5.143	9.286	3	4	5.143	4.714
## [158,]	4.143	2	4	7	8.429	2.857	2	5	8.429	7.143
## [159,]	5.286	5	1	10	2.857	3.714	3	3	2.857	1.286
## [160,]	7.000	4	2	6	5.000	8.000	8	5	5.000	6.000

```

## [161,] 11.143    1    6    9 5.429  7.857    6    7 5.429  2.143
## [162,]  5.714    7    4    9 4.143  2.286    6    8 4.143  0.714
## [163,]  4.857    7    6    3 6.571  2.143    3    1 6.571  3.857
## [164,]  9.143    4    2    2 6.429  8.857    8    7 6.429  6.143
## [165,]  6.714    3    5    1 8.143  6.286    8    6 8.143  7.714
## [166,]  6.000   10    8   10 3.000  5.000    5    8 3.000  2.000
## [167,]  4.714    9   10    7 7.143  2.286    8    7 7.143  4.714
## [168,]  4.286    2    4    8 1.857  9.714    7    9 1.857  7.286
## [169,]  4.000    4    5   10 3.000  2.000   10    2 3.000  1.000
## [170,] 10.286    4    4    7 5.857  7.714    2    6 5.857  3.286
## [171,]  6.571    5    6    3 6.714  5.429    8    4 6.714  5.571
## [172,]  1.571    4    6    2 9.714  0.429    7    7 9.714  8.571
## [173,]  5.429    8    5    1 4.286 10.571    2    7 4.286  9.429
## [174,]  8.714   10    8    3 7.143  6.286    6    7 7.143  4.714
## [175,]  8.571    8    9    8 6.714  9.429    9    3 6.714  7.571
## [176,]  5.857    3    2    1 4.571  5.143   10    2 4.571  3.857
## [177,]  5.571    9    5    2 7.714  3.429    8    7 7.714  5.571
## [178,] 10.429    8    8    3 8.286  5.571    3    6 8.286  3.429
## [179,]  4.286    3    5   10 3.857  4.714    6    2 3.857  4.286
## [180,]  8.714    7    7   10 7.143  4.286    1    7 7.143  2.714
## [181,]  5.429    2    5    6 6.286  6.571   10    6 6.286  7.429
## [182,] 10.286    8    4    9 9.857  5.714    7    2 9.857  5.286
## [183,]  5.143    4    9    3 6.429  5.857    1    3 6.429  7.143
## [184,]  3.000    8    4    3 6.000  3.000    5   10 6.000  6.000
## [185,]  2.714   10    7    1 1.143  4.286    4    9 1.143  2.714
## [186,]  6.143    3   10    4 5.429  5.857    5    5 5.429  5.143
## [187,]  8.571    4   10    9 7.714  4.429    6    4 7.714  3.571
## [188,]  6.143    9    5    2 5.429  5.857   10    4 5.429  5.143
## [189,]  7.714    2    2    9 2.143  5.286    8    6 2.143 -0.286
## [190,]  7.714    7    4   10 3.143  5.286    1    8 3.143  0.714
## [191,]  8.000   10    9    1 7.000  5.000    3   10 7.000  4.000
## [192,]  7.714    7    3    2 6.143  7.286    3   10 6.143  5.714
## [193,]  7.571    9    6    8 4.714  6.429    9    5 4.714  3.571
## [194,]  7.857    7    5    1 7.571  5.143    4    5 7.571  4.857
## [195,]  7.286    9    7    6 7.857  7.714    7    5 7.857  8.286
## [196,]  4.714    8    1    4 3.143  5.286    2    9 3.143  3.714
## [197,]  3.000    9   10    4 5.000  2.000    8   10 5.000  4.000
## [198,]  0.143    6   10    6 5.429  2.857    7    1 5.429  8.143
## [199,]  3.571    9   10    5 5.714  5.429    3    5 5.714  7.571
## [200,]  5.714    7    5    8 4.143  4.286   10    8 4.143  2.714

```

```
caSegmentation(y=sprefm, x=sprof, c=3) # clusters
```

```
## K-means clustering with 3 clusters of sizes 67, 62, 71
##
## Cluster means:
##      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]      [,8]
## 1 4.238806 6.194030 4.865672 3.059701 5.537313 4.283582 4.014925 6.865672
## 2 6.000016 4.338710 4.612903 7.612903 4.032258 5.661274 6.403226 5.306452
## 3 7.150986 6.535211 5.830986 5.084507 7.199239 6.018028 5.704225 3.732394
##      [,9]      [,10]
## 1 5.537313 5.582090
## 2 4.032258 3.693565
## 3 7.199239 6.066479
##
## Clustering vector:
##  [1] 3 2 3 3 2 2 1 1 1 2 3 3 3 3 3 1 2 1 3 1 2 2 3 2 3 1 1 3 3 2 1 1 2 2
## [36] 3 2 1 3 3 3 3 1 2 3 2 2 1 1 3 3 2 1 3 1 2 2 1 1 1 2 3 1 1 1 2 2 3 1 3
## [71] 2 2 2 1 1 1 2 1 3 2 1 2 3 3 1 1 1 1 3 3 1 2 3 3 3 2 1 3 1 1 3 2 2 2 1
## [106] 3 2 3 3 2 3 1 3 1 1 3 2 3 3 2 1 3 2 1 3 2 1 2 2 1 1 1 3 3 3 1 2 1 2 3
## [141] 2 2 1 1 3 3 1 1 3 1 1 2 1 2 1 2 3 1 2 2 2 2 3 3 3 2 3 2 2 2 3 1 1 3 3
## [176] 2 1 3 2 3 2 3 3 1 1 3 3 3 2 2 1 1 2 3 3 1 1 3 1 2
##
## Within cluster sum of squares by cluster:
## [1] 3529.905 3290.644 3670.672
## (between_SS / total_SS =  20.6 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"
## [5] "tot.withinss" "betweenss"    "size"         "iter"
## [9] "ifault"
```

```
library(arules)
```

```
## Loading required package: Matrix
```

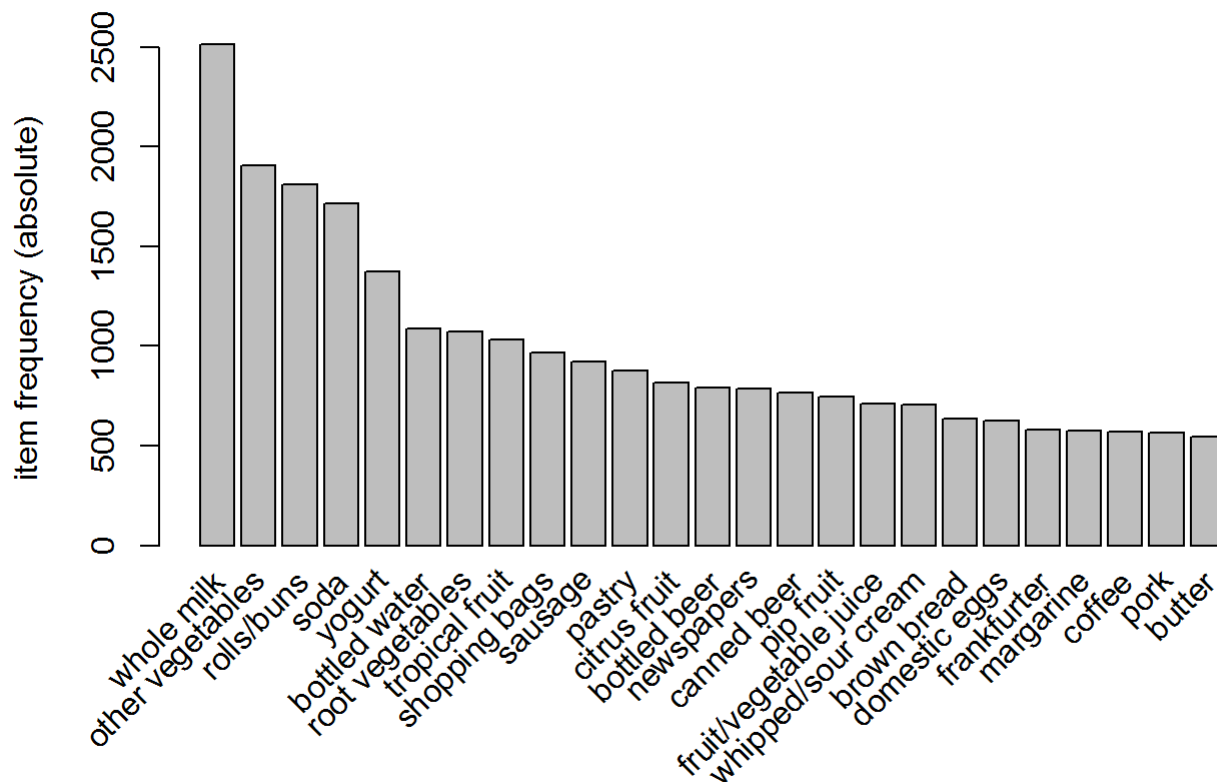
```
##
## Attaching package: 'arules'
```

```
## The following objects are masked from 'package:base':
##
##      abbreviate, write
```

```
library(arulesViz)
```

```
## Loading required package: grid
```

```
library(datasets)
data(Groceries)
itemFrequencyPlot(Groceries,topN=25,type="absolute")
```

```
rules <- apriori(Groceries, parameter = list(supp = 0.001, conf = 0.8))
```

```
## Apriori
##
## Parameter specification:
## confidence minval smax arem aval originalSupport maxtime support minlen
##          0.8    0.1    1 none FALSE                TRUE      5  0.001    1
## 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: 9
##
## set item appearances ...[0 item(s)] done [0.00s].
## set transactions ...[169 item(s), 9835 transaction(s)] done [0.00s].
## sorting and recoding items ... [157 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 ... [410 rule(s)] done [0.00s].
## creating S4 object ... done [0.00s].
```

```
options(digits=2)
inspect(rules[1:5])
```

```
##      lhs                      rhs      support confidence lift
## [1] {liquor,red/blush wine} => {bottled beer} 0.0019  0.90    11.2
## [2] {curd,cereals}          => {whole milk}  0.0010  0.91     3.6
## [3] {yogurt,cereals}        => {whole milk}  0.0017  0.81     3.2
## [4] {butter,jam}            => {whole milk}  0.0010  0.83     3.3
## [5] {soups,bottled beer}    => {whole milk}  0.0011  0.92     3.6
```

```
rules<-sort(rules, by="confidence", decreasing=TRUE)
```

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

```
rules<-apriori(data=Groceries, parameter=list(supp=0.001,conf = 0.08),
               appearance = list(default="lhs",rhs="soda"),
               control = list(verbose=F))
rules<-sort(rules, decreasing=TRUE,by="confidence")
inspect(rules[1:5])
```

```
##      lhs                      rhs      support confidence lift
## [1] {coffee,
##      misc. beverages} => {soda}  0.0010      0.77  4.4
## [2] {yogurt,
##      rolls/buns,
##      bottled water,
##      newspapers}      => {soda}  0.0010      0.77  4.4
## [3] {sausage,
##      bottled water,
##      bottled beer}    => {soda}  0.0011      0.73  4.2
## [4] {sausage,
##      white bread,
##      shopping bags}   => {soda}  0.0010      0.67  3.8
## [5] {rolls/buns,
##      bottled water,
##      chocolate}       => {soda}  0.0013      0.65  3.7
```

```
rules<-apriori(data=Groceries, parameter=list(supp=0.001,conf = 0.15,minlen=2),
               appearance = list(default="rhs",lhs="soda"),
               control = list(verbose=F))
rules<-sort(rules, decreasing=TRUE,by="confidence")
inspect(rules[1:5])
```

##	lhs	rhs	support	confidence	lift
## [1]	{soda}	=> {whole milk}	0.040	0.23	0.90
## [2]	{soda}	=> {rolls/buns}	0.038	0.22	1.20
## [3]	{soda}	=> {other vegetables}	0.033	0.19	0.97
## [4]	{soda}	=> {bottled water}	0.029	0.17	1.50
## [5]	{soda}	=> {yogurt}	0.027	0.16	1.12

```
library(arulesViz)
plot(rules,method="graph",interactive=TRUE,shading=NA)
```

```
require(ggplot2)
```

```
## Loading required package: ggplot2
```

```
bank = read.csv("D:/Business Analytics/LIBA Class Room Materials/Prof Rajesh/assignment/bank.csv", header = TRUE, stringsAsFactors = FALSE)
str(bank)
```

```
## 'data.frame': 4521 obs. of 9 variables:
## $ age : int 30 33 35 30 59 35 36 39 41 43 ...
## $ Jobtype : int 3 2 1 1 2 1 1 2 1 2 ...
## $ marital : chr "married" "married" "single" "married" ...
## $ education: chr "primary" "secondary" "tertiary" "tertiary" ...
## $ default : chr "no" "no" "no" "no" ...
## $ balance : int 1787 4789 1350 1476 0 747 307 147 221 -88 ...
## $ housing : chr "no" "yes" "yes" "yes" ...
## $ loan : chr "no" "yes" "no" "yes" ...
## $ response : chr "no" "no" "no" "no" ...
```

```
bank$jobtype=as.factor(bank$Jobtype)
bank$jobtype = factor(bank$jobtype, levels = c(1, 2, 3),
                      labels = c("1", "2", "3"))
bank$marital = factor(bank$marital,
                      labels = c("Divorced", "Married", "Single"))
bank$education = factor(bank$education,
                        labels = c("Primary", "Secondary", "Tertiary", "Unknown"))
bank$default = factor(bank$default, labels = c("No", "Yes"))
bank$housing = factor(bank$housing, labels = c("No", "Yes"))
bank$loan = factor(bank$loan, labels = c("No", "Yes"))
bank$response = factor(bank$response, labels = c("No", "Yes"))
bank2 = subset(bank, select = c("response", "age", "jobtype", "marital", "education","default",
"balance", "housing", "loan"))
str(bank2 )
```

```
## 'data.frame': 4521 obs. of 9 variables:
## $ response : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 ...
## $ age : int 30 33 35 30 59 35 36 39 41 43 ...
## $ jobtype : Factor w/ 3 levels "1","2","3": 3 2 1 1 2 1 1 2 1 2 ...
## $ marital : Factor w/ 3 levels "Divorced","Married",...: 2 2 3 2 2 3 2 2 2 2 ...
## $ education: Factor w/ 4 levels "Primary","Secondary",...: 1 2 3 3 2 3 3 2 3 1 ...
## $ default : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 1 1 1 1 1 ...
## $ balance : int 1787 4789 1350 1476 0 747 307 147 221 -88 ...
## $ housing : Factor w/ 2 levels "No","Yes": 1 2 2 2 2 1 2 2 2 2 ...
## $ loan : Factor w/ 2 levels "No","Yes": 1 2 1 2 1 1 1 1 1 2 ...
```

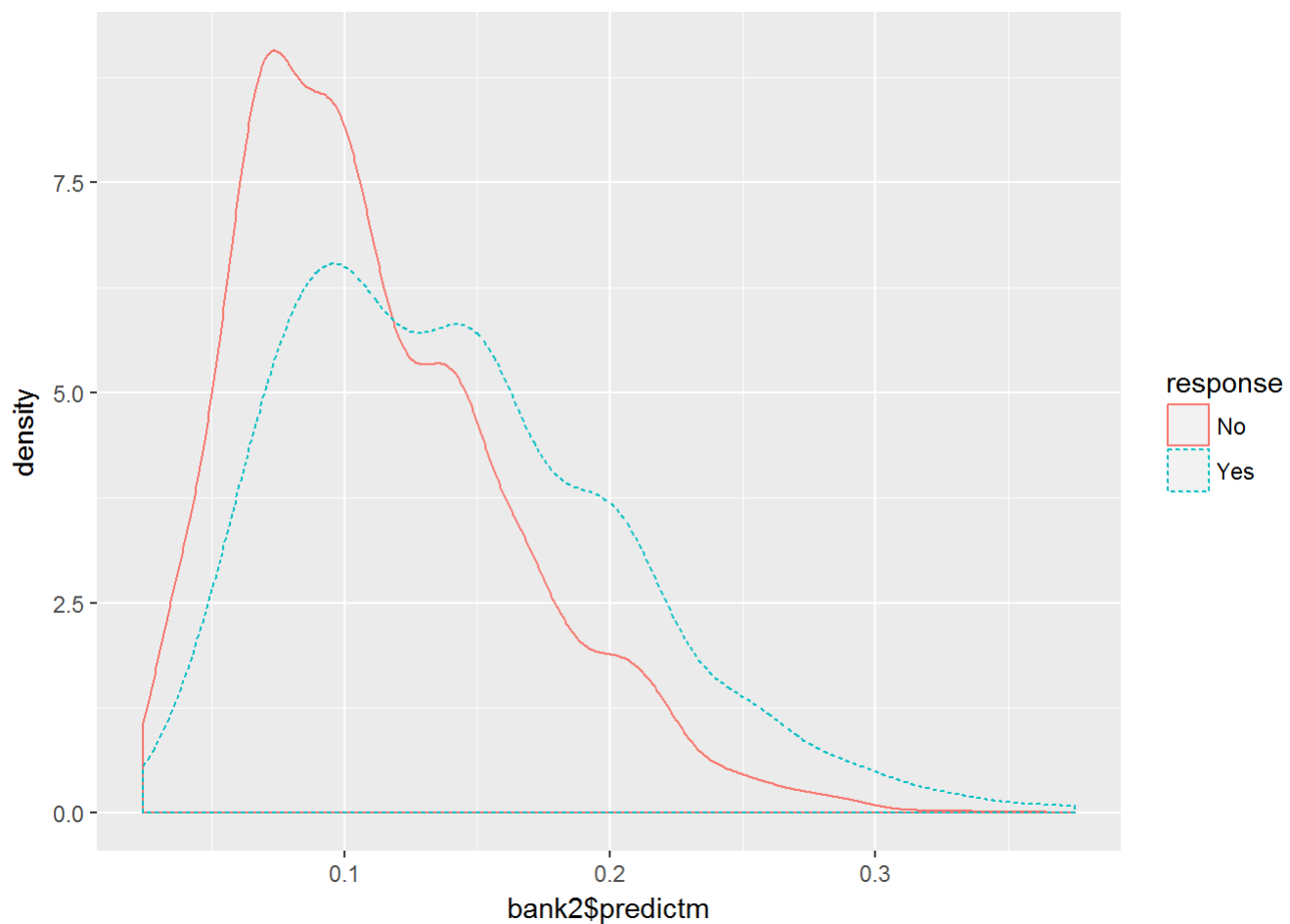
```
summary(bank2 )
```

```
## response      age      jobtype      marital      education
## No :4000      Min.   :19      1:1798      Divorced: 528      Primary   : 678
## Yes: 521      1st Qu.:33      2:2131      Married :2797      Secondary:2306
##              Median :39      3: 592      Single  :1196      Tertiary :1350
##              Mean   :41              Unknown  : 187
##              3rd Qu.:49
##              Max.   :87
## default      balance      housing      loan
## No :4445      Min.   : -3313      No :1962      No :3830
## Yes: 76      1st Qu.: 69      Yes:2559      Yes: 691
##              Median : 444
##              Mean   : 1423
##              3rd Qu.: 1480
##              Max.   :71188
```

```
bankf = {response ~ age + jobtype + education + marital + default + balance + housing + loan}
bankmodel = glm(bankf, family=binomial, data=bank2 )
summary(bankmodel)
```

```
##
## Call:
## glm(formula = bankf, family = binomial, data = bank2)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.939  -0.543  -0.444  -0.361   2.622
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -2.71e+00  2.98e-01  -9.07  < 2e-16 ***
## age          1.25e-02  5.04e-03   2.47  0.0133 *
## jobtype1     -5.37e-02  7.41e-02  -0.73  0.4680
## jobtype2     -2.36e-01  7.32e-02  -3.22  0.0013 **
## education1   -1.59e-01  1.25e-01  -1.27  0.2039
## education2    1.24e-01  9.14e-02   1.36  0.1743
## education3    2.92e-01  1.03e-01   2.84  0.0045 **
## marital1      1.22e-01  9.39e-02   1.30  0.1936
## marital2     -2.77e-01  6.82e-02  -4.06  4.9e-05 ***
## default1     -7.90e-02  1.82e-01  -0.43  0.6649
## balance       9.46e-07  1.47e-05   0.06  0.9487
## housing1      2.47e-01  4.99e-02   4.95  7.6e-07 ***
## loan1         3.60e-01  8.37e-02   4.30  1.7e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 3231.0  on 4520  degrees of freedom
## Residual deviance: 3108.1  on 4508  degrees of freedom
## AIC: 3134
##
## Number of Fisher Scoring iterations: 5
```

```
bank2$predictm = predict.glm(bankmodel, type = "response")
ggplot(bank2,aes(x=bank2$predictm, color=response, linetype=response))+ geom_density()
```



```
bank2$Predictm =ifelse((bank2 $predictm > 0.1), 1, 0) #@10% cutoff
bank2$Predictm = factor(bank2$Predictm,levels = c(0, 1), labels = c("NO", "YES"))
confusion_matrix = table(bank2$Predictm, bank2$response)
print(confusion_matrix)
```

```
##
##           No  Yes
##  NO  1978  160
##  YES  2022  361
```