# Analysis of data on canadian women Labour Force participation

G. Marchetti

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## Data on Women's Labour-Force Participation

The data are transformed as follows.

```
library(carData)
data(Womenlf)

wlfdata <- Womenlf
wlfdata$hincome <- factor(0 + (Womenlf$hincome > 14))
wlfdata$partic <- factor(0 + (Womenlf$partic == "fulltime"))
wlfdata$children <-factor(0 + (Womenlf[,3]=="present"))
wlfdata$region <- Womenlf[,4]
colnames(wlfdata) <- c("L", "H", "C", "R")
head(wlfdata)</pre>
```

```
L H C R
1 0 1 1 Ontario
2 0 0 1 Ontario
3 0 1 1 Ontario
4 0 1 1 Ontario
5 0 1 1 Ontario
6 0 0 1 Ontario
```

Contingency table

```
ftable(R ~ H + C + L, data = wlfdata)
```

R Atlantic BC Ontario Prairie Quebec

Η	С	L					
0	0	0	1	4	6	0	5
		1	1	4	11	3	10
	1	0	11	6	25	15	18
		1	2	0	4	4	4
1	0	0	1	4	6	1	5
		1	1	2	10	1	3
	1	0	11	8	44	7	19
		1	2	1	2	0	1

## Test conditional independency

You need package bnlearn for this.

```
ci.test( "H", "C", "R", data = wlfdata, test = "mi")
```

Mutual Information (disc.)

```
data: H ~ C | R
```

mi = 3.7358, df = 5, p-value = 0.588

alternative hypothesis: true value is greater than 0

```
m_full0 <- glm(L ~ C *H * R, family = binomial, data = wlfdata)
round(summary(m_full0)$coef, 3)</pre>
```

	Estimate	Std. Error	${\tt z}$ value	Pr(> z )
(Intercept)	0.000	1.414	0.000	1.000
C1	-1.705	1.610	-1.059	0.290
H1	0.000	2.000	0.000	1.000
RBC	0.000	1.581	0.000	1.000
ROntario	0.606	1.503	0.403	0.687
RPrairie	17.566	2284.102	0.008	0.994
RQuebec	0.693	1.517	0.457	0.648
C1:H1	0.000	2.276	0.000	1.000
C1:RBC	-15.861	1615.105	-0.010	0.992
C1:ROntario	-0.734	1.772	-0.414	0.679
C1:RPrairie	-17.183	2284.102	-0.008	0.994
C1:RQuebec	-0.492	1.788	-0.275	0.783
H1:RBC	-0.693	2.291	-0.303	0.762

```
H1:ROntario
               -0.095
                          2.127 -0.045
                                         0.964
H1:RPrairie
              -17.566
                       2284.103 -0.008
                                         0.994
H1:RQuebec
               -1.204
                          2.198 -0.548
                                         0.584
C1:H1:RBC
               16.180
                       1615.106 0.010
                                        0.992
               -1.163
                          2.553 -0.456 0.649
C1:H1:ROntario
C1:H1:RPrairie
               1.322
                       2730.025 0.000
                                         1.000
C1:H1:RQuebec
               -0.236
                          2.715 -0.087
                                         0.931
```

```
m_red0 <- glm(L ~ C * H, family = binomial, data = wlfdata)
anova(m_red0, m_full0, test = "Chisq")</pre>
```

Analysis of Deviance Table

```
Model 1: L ~ C * H

Model 2: L ~ C * H * R

Resid. Df Resid. Dev Df Deviance Pr(>Chi)

1 259 227.93

2 243 216.66 16 11.269 0.7926
```

#### Fit equation 1

```
m_full1 <- glm(L ~ C + H + R, family = binomial, data = wlfdata)
round(summary(m_full1)$coef, 3)</pre>
```

```
Estimate Std. Error z value Pr(>|z|)
(Intercept)
              1.041
                        0.607 1.715
                                         0.086
C1
             -2.609
                        0.361 -7.234
                                         0.000
H1
             -0.768
                        0.348 - 2.210
                                         0.027
RBC
                        0.745 -1.266
             -0.944
                                         0.206
ROntario
             -0.254
                        0.590 - 0.430
                                         0.667
RPrairie
              0.168
                        0.695 0.241
                                         0.809
RQuebec
             -0.342
                        0.627 -0.545
                                         0.586
```

```
m_red1 <- glm(L ~ C + H, family = binomial, data = wlfdata)
anova(m_red1, m_full1, test = "Chisq")</pre>
```

Analysis of Deviance Table

Model 1: L ~ C + H

```
Model 2: L \sim C + H + R
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1
        260
               228.31
2
        256
                225.50 4 2.8117 0.5898
`LRtest` <- function(m_red, m_full) {
10 <- logLik(m_red)[1]</pre>
lsat <- logLik(m_full)[1]</pre>
w = 2 * (1sat - 10)
df = m_red$df.residual - m_full$df.residual
p = 1 - pchisq(w, df)
c(w = w, df = df, p = p)
LRtest(m_red1, m_full1)
                df
2.8117211 4.0000000 0.5898111
Fit equation 2
m_full2 <- glm(C~ H+R, family = binomial, data = wlfdata)</pre>
round(summary(m_full2)$coef, 3)
            Estimate Std. Error z value Pr(>|z|)
(Intercept)
                       0.551 3.030 0.002
            1.671
H1
              0.437
                        0.282 1.546 0.122
RBC
             -1.827
                        0.656 -2.785 0.005
ROntario
             -1.092
                         0.579 -1.886 0.059
                         0.730 -0.186
RPrairie
             -0.136
                                        0.852
RQuebec
             -1.250
                         0.598 - 2.089
                                          0.037
m_red2 <- glm(C~ R, family = binomial, data = wlfdata)</pre>
LRtest(m_red2, m_full2)
```

### Fit equation 3

```
m_full3 <- glm(H ~ R, family = binomial, data = wlfdata)
round(summary(m_full3)$coef, 3)</pre>
```

```
Estimate Std. Error z value Pr(>|z|)
(Intercept)
                        0.365
                               0.000
             0.000
                                       1.000
RBC
             0.069
                        0.521 0.132
                                       0.895
ROntario
             0.298
                        0.414 0.721
                                       0.471
RPrairie
            -0.894
                        0.538 -1.660
                                       0.097
RQuebec
            -0.279
                        0.443 -0.629
                                       0.529
```

```
m_red3 <- glm(H ~ 1, family = binomial, data = wlfdata)
LRtest(m_red3, m_full3)</pre>
```

```
w df p
9.19254035 4.00000000 0.05646299
```