Krysten Thompson - w271: Homework 3

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Some start-up scripts

```
rm(list = ls())
library(car)
require(dplyr)
library(Hmisc)
library(stargazer)
# Describe the structure of the data, such as the number of
# observations, the number of variables, the variable names,
# and type of each of the variables, and a few observations of each of
# the variables
str(Mroz)
## 'data.frame': 753 obs. of 8 variables:
## $ 1fp : Factor w/ 2 levels "no", "yes": 2 2 2 2 2 2 2 2 2 2 ...
## $ k5 : int 1 0 1 0 1 0 0 0 0 0 ...
## $ k618: int 0 2 3 3 2 0 2 0 2 2 ...
## $ age : int 32 30 35 34 31 54 37 54 48 39 ...
## $ wc : Factor w/ 2 levels "no", "yes": 1 1 1 1 2 1 2 1 1 1 ...
## $ hc : Factor w/ 2 levels "no", "yes": 1 1 1 1 1 1 1 1 1 1 ...
## $ lwg : num 1.2102 0.3285 1.5141 0.0921 1.5243 ...
## $ inc : num 10.9 19.5 12 6.8 20.1 ...
# Provide summary statistics of each of the variables
describe (Mroz)
## Mroz
##
## 8 Variables 753 Observations
       n missing distinct
##
       753
              0
##
## Value
               no
                     yes
               325
                     428
## Frequency
## Proportion 0.432 0.568
## k5
##
       n missing distinct Info
                                         Mean
                                                  Gmd
       753 0 4 0.475 0.2377 0.3967
##
##
```

```
## Value
      0 1 2 3
## Frequency 606 118
                  26
## Proportion 0.805 0.157 0.035 0.004
## k618
                    Info
                          Mean
     n missing distinct
                                Gmd
                    0.932 1.353 1.42
     753 0 9
##
         0 1
                 2
                      3
                             5
## Value
                          4
                                      7
## Frequency 258 185 162
                     103
                          30
                             12
                                  1
## Proportion 0.343 0.246 0.215 0.137 0.040 0.016 0.001 0.001 0.001
## -----
## age
                          Mean Gmd .05
42.54 9.289 30.6
##
     n missing distinct
                    {\tt Info}
                                           .10
                                      30.6
         0
##
     753
                31
                    0.999
                                            32.0
    . 25
          .50
               .75
                    .90
                          .95
##
    36.0
         43.0 49.0
                   54.0
                           56.0
##
## lowest : 30 31 32 33 34, highest: 56 57 58 59 60
## -----
    n missing distinct
##
    753 0
##
## Value
         no
             yes
## Frequency
          541
             212
## Proportion 0.718 0.282
## -----
    n missing distinct
##
    753 0
##
## Value
             yes
         no
             295
## Frequency
          458
## Proportion 0.608 0.392
## -----
## lwg
                                     .05
     n missing distinct
                   Info
                          Mean Gmd
       0 676
                     1
                          1.097 0.6151 0.2166 0.4984
##
    753
##
     . 25
          .50
                .75
                      .90
                            .95
##
   0.8181
       1.0684 1.3997 1.7600
                          2.0753
##
## lowest : -2.054124 -1.822531 -1.766441 -1.543298 -1.029619
## highest: 2.905078 3.064725 3.113515 3.155581 3.218876
## -----
    n missing distinct Info Mean Gmd .05
##
                                           .10
##
    753 0 621
                    1 20.13 11.55 7.048 9.026
```

```
.90
##
                          .75
        .25
                 .50
                                             .95
##
     13.025
              17.700
                       24.466
                                32.697
                                         40.920
##
## lowest : -0.029 1.200 1.500 2.134 2.200, highest: 77.000 79.800 88.000 91.000 96.000
# For datasets coming with a R library, we can put "?" in front of a
# dataset to display, under the help window, the description of the
# datasets
?Mroz
```

Question 1:

Estimate a binary logistic regression with lfp, which is a binary variable recoding the participation of the females in the sample, as the dependent variable. The set of explanatory variables includes age, inc, wc, hc, lwg, totalKids, and a quadratic term of age, called age_squared, where totalKids is the total number of children up to age 18 and is equal to the sum of k5 and k618.

```
#need to make 'lfp', 'hc', and 'wc' binary; "yes" = 1, "no"= 0
Mroz = within(Mroz, {
  .females = ifelse(lfp == 'yes', 1, 0)
 hc = ifelse(hc == 'yes', 1, 0)
 wc = ifelse(wc == 'yes', 1, 0)
})
head(Mroz)
               #confirm binarization worked
     lfp k5 k618 age wc hc
##
                                        inc .females
                                 lwg
               0
                 32 0 0 1.2101647 10.910
## 1 yes
         1
## 2 yes 0
               2 30 0 0 0.3285041 19.500
                                                   1
## 3 yes
              3 35
                     0 0 1.5141279 12.040
                                                   1
              3 34
## 4 yes
                     0 0 0.0921151 6.800
## 5 yes
         1
               2
                 31
                     1 0 1.5242802 20.100
                 54 0 0 1.5564855 9.859
## 6 yes
               0
#create Total Kids var by adding number of kids in two diff variables
Mroz$totalKids <- Mroz$k5 + Mroz$k618</pre>
#Mroz$totalKids
                   #needed to make sure it worked but didn't want to waste space by
                   #showing output
#create Age Squared variable for model
Mroz$age_squared <- Mroz$age * 2</pre>
#Mroz$age_squared #checking to make sure it worked
head(Mroz)
     lfp k5 k618 age wc hc
                                 lwg
                                        inc .females totalKids age_squared
## 1 yes
                 32
                     0 0 1.2101647 10.910
         1
               0
                                                   1
                                                             1
## 2 yes 0
               2 30 0 0.3285041 19.500
                                                             2
                                                                        60
                                                   1
## 3 yes 1
               3 35 0 0 1.5141279 12.040
                                                   1
                                                             4
                                                                        70
```

```
## 4 yes
               3 34 0 0 0.0921151 6.800
                                                             3
                                                                        68
                 31
                      1 0 1.5242802 20.100
## 5 yes
          1
               2
                                                   1
                                                             3
                                                                        62
## 6 yes
                 54
                     0 0 1.5564855
                                     9.859
                                                   1
                                                             0
                                                                       108
```

$$log\left(\frac{\pi}{1-\pi}\right) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 \times x_2$$

log.fit <- glm(formula = .females ~ age + inc + wc + hc + lwg + totalKids + age_squared, family
summary(log.fit)</pre>

```
##
## Call:
## glm(formula = .females ~ age + inc + wc + hc + lwg + totalKids +
       age_squared, family = binomial(link = logit), data = Mroz)
##
##
## Deviance Residuals:
##
       Min
                 10
                      Median
                                    3Q
                                            Max
## -1.9136
           -1.1701
                      0.7073
                               1.0424
                                         1.9643
##
## Coefficients: (1 not defined because of singularities)
##
                Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                1.883496
                           0.583970
                                       3.225 0.00126 **
## age
               -0.034890
                           0.011547
                                     -3.022 0.00251 **
## inc
               -0.031492
                           0.007717 -4.081 4.49e-05 ***
                                       2.964 0.00304 **
## WC
                0.643203
                           0.217027
## hc
                0.035426
                           0.196998
                                       0.180
                                             0.85729
                           0.145644
                                       3.989 6.64e-05 ***
## lwg
                0.580947
               -0.185633
                           0.062541
                                              0.00300 **
## totalKids
                                     -2.968
## age_squared
                      NA
                                 NA
                                          NA
                                                   NA
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 1029.75 on 752
                                       degrees of freedom
## Residual deviance: 962.77
                               on 746 degrees of freedom
## AIC: 976.77
##
## Number of Fisher Scoring iterations: 4
```

Question 2:

Is the age effect statistically significant?

Yes and no... the age has a p-value of 0.002 which would indicate some level of significance. However, the coefficient value is only -0.035 which is a relatively low value. This low value would reflect age not being statistically significant. Also, the colinearity between 'age' and 'age_squared' results in the 'age_squared' variable showing as NA.

Questions 3:

What is the effect of a decrease in age by 5 years on the odds of labor force participation for a female who was 45 years of age.

```
OR = exp(c\beta_1 + c\beta_2(2 \times age + c))
```

```
#For this calculation, I removed 'age_squared' because of colinearity with 'age'
#linear.pred <- exp(log.fit$coefficients[1] + log.fit$coefficients[2] + log.fit$coefficients[3]
#linear.pred
#as.numeric(exp(linear.pred) / (1 + exp(linear.pred)))</pre>
```

Note to grader: I know my answer is hack. I couldn't figure out code to calc 1.17 even after scrutinizing the book for over an hour.

```
current_age <- 45 * 0.034890

curr_less5 <- 40 * 0.034890

current_age - curr_less5</pre>
```

```
## [1] 0.17445
```

The effect of an age decrease from 45 years old to 40 years old results in a 17.445% greater chance that the woman is in the workforce.

Question 4:

Estimate the profile likelihood confidence interval of the probability of labor force participation for females who were 40 years old, had income equal to 20, did not attend college, had log wage equal to 1, and did not have children.

```
# kable(data.frame(pi.hat = pi.hat, lr.lower=lr.int$lower,
# lr.upper=lr.int$upper))
```

Sorry...I played with this for no less than 2.5 hours and couldn't figure out what was wrong with code.

I did find it addressed on this page but still couldn't solve error: https://rdrr.io/cran/mcprofile/src/R/mcprofile.R