WU 10 – Math 150, Spring 2023

Thursday, 2/16/23

Your Name:

Names of people you worked with: _____

- Cat or dog?
- How / where can you read the feedback you receive on graded HW assignments?

Task: HERS is a clinical trial to test the efficacy and safety of estrogen + progestin therapy. The logistic model aims to predict whether the individuals had a pre-existing medical condition, medcond, using age, drinkany, and diabetes.

glm(medcond ~ age + diabetes + drinkany, data = HERS, family="binomial") %>% tidy()

```
## # A tibble: 4 x 5
##
     term
                 estimate std.error statistic
                                                    p.value
##
     <chr>
                    <dbl>
                               <dbl>
                                         <dbl>
                                                      <dbl>
                 -1.72
                             0.413
                                         -4.17 0.0000300
## 1 (Intercept)
                   0.0176
                             0.00605
                                          2.90 0.00369
## 2 age
## 3 diabetes
                   0.442
                             0.0895
                                          4.94 0.00000786
                  -0.252
                             0.0834
                                         -3.01 0.00257
## 4 drinkany
glm(medcond ~ (age + diabetes + drinkany)^2, data = HERS, family="binomial") %>% tidy()
## # A tibble: 7 x 5
##
     term
                       estimate std.error statistic p.value
##
     <chr>
                           <dbl>
                                     <dbl>
                                                <dbl>
                                                         <dbl>
                                               -3.47 0.000515
## 1 (Intercept)
                        -2.14
                                   0.616
## 2 age
                         0.0235
                                   0.00905
                                               2.60 0.00941
## 3 diabetes
                         2.65
                                   0.940
                                               2.82 0.00481
## 4 drinkany
                        -0.714
                                   0.864
                                              -0.827 0.408
                        -0.0326
                                              -2.33 0.0201
## 5 age:diabetes
                                   0.0140
## 6 age:drinkany
                         0.00757
                                   0.0128
                                               0.593 0.553
## 7 diabetes:drinkany -0.153
                                   0.198
                                              -0.770 0.441
glm(medcond ~ age + diabetes + drinkany, data = HERS, family="binomial") %>% glance()
## # A tibble: 1 x 8
##
     null.deviance df.null logLik
                                     AIC
                                           BIC deviance df.residual nobs
                     <int> <dbl> <dbl> <dbl>
##
             \langle dh \rangle
                                                   <dbl>
                                                               <int> <int>
## 1
             3644.
                      2760 -1799. 3605. 3629.
                                                   3597.
                                                                2757 2761
glm(medcond ~ (age + diabetes + drinkany)^2, data = HERS, family="binomial") %>% glance()
## # A tibble: 1 x 8
##
    null.deviance df.null logLik
                                     AIC
                                           BIC deviance df.residual nobs
##
             <dbl>
                     <int> <dbl> <dbl> <dbl>
                                                   <dbl>
                                                               <int> <int>
## 1
             3644.
                      2760 -1795. 3604. 3646.
                                                   3590.
                                                                2754 2761
```

Conduct a drop in deviance test (also called a likelihood ratio test) to evaluate whether any interaction at all is warranted in the model.

- 1. Write down the null hypothesis in terms of the relevant β values.
- 2. Find the deviance for the model under the null hypothesis and the deviance for the model under the alternative hypothesis.
- 3. Using the test statistic (difference in deviances) and the Chi-squared distribution with the appropriate degrees of freedom, calculate the p-value (you'll likely need pchisq()).
- 4. From the p-value, what is the conclusion with respect to the model? That is, which model do you put forward to your client?

Solution:

- 1. Using the second model (the one with interaction terms), $H_0: \beta_4 = \beta_5 = \beta_6 = 0$
- 2. null deviance = 3597.325; full model deviance = 3590.151
- 3. The p-value is computed using the chi-square distribution.
- 1 pchisq(3597 3590, 3)

[1] 0.07189777

4. Although borderline, the p-value is greater than 0.05 indicating that we can't reject H_0 . Here that means we don't have evidence to believe that the coefficients are non-zero. In addition to the p-value, we can also use Occam's Razor as a strategy: simple models are often better. So we will not move forward with any interaction terms in the model.