Examen 2. Métodos Estadísticos Avanzados

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Objetivo

Reproducir los resultados del Ejemplo 1 (Modelo I y II) de la sección 3 del artículo **Fitzmaurice**, **G. M.** and **N. M. Laird** (1993). A likelihood - based for analysingg longitudinal binary responses. Biometrika 80 (1), 141 - 151, utilizando:

- 1) El método propuesto en el artículo.
- 2) El enfoque bayesiano con el método computacional de su preferencia.

Resultados

Preparación

```
data <- read.csv("Ohio.csv",header=T)
data$resp <- as.factor(data$resp)
data$smoke <- as.factor(data$smoke)
#data$id <- as.factor(data$id)</pre>
```

Modelo I: Independencia entre las observaciones en el tiempo

```
m1 <- glm(resp ~ age + smoke + age*smoke, family = binomial(link="logit"),
          data = data)
summary(m1)
##
## glm(formula = resp ~ age + smoke + age * smoke, family = binomial(link = "logit"),
##
       data = data)
##
## Deviance Residuals:
##
                1Q
                     Median
                                           Max
## -0.6503 -0.6014 -0.5636 -0.4940
                                        2.0804
##
## Coefficients:
              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.90084
                          0.08874 -21.420
                                             <2e-16 ***
              -0.14125
                          0.06951 -2.032
                                             0.0422 *
## age
                                    2.252
                          0.13944
## smoke1
               0.31395
                                             0.0244 *
## age:smoke1 0.07084
                          0.11072
                                    0.640
                                            0.5223
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
```

```
##
      Null deviance: 1829.1 on 2147 degrees of freedom
## Residual deviance: 1819.5 on 2144 degrees of freedom
## AIC: 1827.5
## Number of Fisher Scoring iterations: 4
Modelo II: Correlación intercambiable entre años sucesivos
library(gee)
m2 <- gee(resp ~ age + smoke + age*smoke, id = id,
          data=data, family=binomial, corstr = "exchangeable")
## Beginning Cgee S-function, @(#) geeformula.q 4.13 98/01/27
## running glm to get initial regression estimate
## (Intercept)
                                smoke1 age:smoke1
                       age
## -1.9008426 -0.1412531
                             0.3139540
                                        0.0708441
summary(m2)
##
##
   GEE: GENERALIZED LINEAR MODELS FOR DEPENDENT DATA
   gee S-function, version 4.13 modified 98/01/27 (1998)
##
## Model:
## Link:
                               Logit
## Variance to Mean Relation: Binomial
## Correlation Structure:
                               Exchangeable
##
## Call:
## gee(formula = resp ~ age + smoke + age * smoke, id = id, data = data,
      family = binomial, corstr = "exchangeable")
##
## Summary of Residuals:
                             Median
                                                      Max
                      1Q
## -0.1906393 -0.1654776 -0.1468831 -0.1148906 0.8851094
##
##
## Coefficients:
##
                  Estimate Naive S.E.
                                          Naive z Robust S.E.
                                                                 Robust z
## (Intercept) -1.90049539 0.11871090 -16.0094430 0.11908696 -15.9588874
              -0.14123592 0.05608034 -2.5184570 0.05820089 -2.4266968
## age
## smoke1
                0.31382583 0.18719721
                                        1.6764450 0.18784180
                                                                1.6706922
                0.07083185 0.08917757
                                        0.7942788 0.08827886
                                                                0.8023647
## age:smoke1
## Estimated Scale Parameter: 1.001273
## Number of Iterations:
##
## Working Correlation
             [,1]
                       [,2]
                                 [,3]
## [1,] 1.0000000 0.3543843 0.3543843 0.3543843
## [2,] 0.3543843 1.0000000 0.3543843 0.3543843
```

[3,] 0.3543843 0.3543843 1.0000000 0.3543843 ## [4,] 0.3543843 0.3543843 0.3543843 1.0000000