

Knee Injuries - Marginal Models

February 5, 2020

First the dataset knee is loaded:

```
> library(catdata)
> data(knee)
> attach(knee)
```

To obtain a simple binary model the response variables are dichotomized. The groups are constructed by pain level up to level 2 und pain level higher than level 2.

```
> R2D <- rep(0, length(R2))
> R3D <- rep(0, length(R3))
> R4D <- rep(0, length(R3))
> R2D[R2>2] <- 1
> R3D[R3>2] <- 1
> R4D[R4>2] <- 1
```

Now the covariates have to be transformed so that they can be used for the functions "gee" from the "gee"-library and "geeglm" from the "geepack"-library, which will be employed for fitting the models.

```
> N <- rep(knee$N, each=3)
> Th <- rep(knee$Th, each=3)
> Age <- rep(knee$Age, each=3)
> Sex <- rep(knee$Sex, each=3)
```

Now the response vector is built and the quadratic age-effect "Age2" is computed.

```
> Response <- c(rbind(R2D,R3D,R4D))
> Age2 <- Age^2
```

The covariates therapy and sex are treated as factors:

```
> Th <- as.factor(Th)
> Sex <- as.factor(Sex)
```

First the GEEs are fitted with the function "gee" from library "gee".

```
> library(gee)
```

The first model is a GEE with independent correlation structure:

```

> gee1a <- gee(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit))

> summary(gee1a)

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: Independent

Call:
gee(formula = Response ~ Th + Sex + Age + Age2, id = N, family = binomial(link = logit))

Summary of Residuals:
      Min       1Q   Median       3Q      Max
-0.8093879 -0.4771034  0.2393885  0.3881349  0.7660757

Coefficients:
            Estimate  Naive S.E.  Naive z Robust S.E.  Robust z
(Intercept) -4.749831307 1.279009755 -3.713679 1.821208734 -2.6080653
Th2          -0.673979165 0.223825316 -3.011184 0.334205312 -2.0166620
Sex1          0.265689239 0.241933592  1.098191 0.366733338  0.7244753
Age           0.381150842 0.087570504  4.352503 0.125833185  3.0290169
Age2         -0.006124345 0.001381627 -4.432705 0.001984839 -3.0855622

Estimated Scale Parameter: 1.019152
Number of Iterations: 1

Working Correlation
 [,1] [,2] [,3]
[1,]    1    0    0
[2,]    0    1    0
[3,]    0    0    1

```

The second model is a GEE with exchangeable correlation structure:

```

> gee2a <- gee(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit), corstr="exchangeable")

> summary(gee2a)

```

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 = Response ~ Th + Sex + Age + Age2, id = N, family = binomial(link = logit),
     corstr = "exchangeable")

Summary of Residuals:
      Min       1Q    Median       3Q      Max
-0.8093879 -0.4771034  0.2393885  0.3881349  0.7660757

Coefficients:
            Estimate   Naive S.E.   Naive z Robust S.E.   Robust z
(Intercept) -4.749831307 1.903793545 -2.4949298 1.821208734 -2.6080653
Th2          -0.673979165 0.333161800 -2.0229785 0.334205312 -2.0166620
Sex1          0.265689239 0.360115792  0.7377884 0.366733338  0.7244753
Age           0.381150842 0.130347841  2.9241055 0.125833185  3.0290169
Age2         -0.006124345 0.002056538 -2.9779873 0.001984839 -3.0855622

Estimated Scale Parameter: 1.019152
Number of Iterations: 1

Working Correlation
      [,1]      [,2]      [,3]
[1,] 1.0000000 0.6078016 0.6078016
[2,] 0.6078016 1.0000000 0.6078016
[3,] 0.6078016 0.6078016 1.0000000

Finally a GEE with exponential correlation structure is fitted:

> gee3a <- gee(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit), corstr="AR-M", Mv=1)

> summary(gee3a)

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: AR-M , M = 1

Call:
gee(formula = Response ~ Th + Sex + Age + Age2, id = N, family = binomial(link = logit),
     corstr = "AR-M", Mv = 1)

Summary of Residuals:
      Min       1Q    Median       3Q      Max
-0.8061636 -0.4668263  0.2354196  0.3833613  0.7933803

```

```

Coefficients:
            Estimate   Naive S.E.   Naive z Robust S.E.   Robust z
(Intercept) -4.72614143 1.912768743 -2.4708379 1.784861526 -2.6479037
Th2          -0.74849866 0.333910055 -2.2416176 0.328084283 -2.2814219
Sex1          0.19277195 0.361954995  0.5325854 0.362465544  0.5318352
Age           0.38489413 0.131285068  2.9317434 0.123289666  3.1218685
Age2          -0.00621548 0.002074949 -2.9954850 0.001945528 -3.1947525

```

```

Estimated Scale Parameter: 1.018095
Number of Iterations: 3

```

```

Working Correlation
      [,1]      [,2]      [,3]
[1,] 1.0000000 0.7058422 0.4982131
[2,] 0.7058422 1.0000000 0.7058422
[3,] 0.4982131 0.7058422 1.0000000

```

In the following the corresponding marginal models are fitted with the function "geeglm" from the library "geepack".

```
> library(geepack)
```

Model with independent correlation structure:

```

> gee1b <- geeglm(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit))
> summary(gee1b)

```

Call:

```
geeglm(formula = Response ~ Th + Sex + Age + Age2, family = binomial(link = logit),
       id = N)
```

Coefficients:

	Estimate	Std.err	Wald Pr(> W)
(Intercept)	-4.749831	1.821209	6.802 0.00911 **
Th2	-0.673979	0.334205	4.067 0.04373 *
Sex1	0.265689	0.366733	0.525 0.46877
Age	0.381151	0.125833	9.175 0.00245 **
Age2	-0.006124	0.001985	9.521 0.00203 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Correlation structure = independence

Estimated Scale Parameters:

	Estimate	Std.err
(Intercept)	1.006	0.103

Number of clusters: 127 Maximum cluster size: 3

Model with exchangeable correlation structure:

```

> gee2b <- geeglm(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit), corstr="exchangeable")

> summary(gee2b)

Call:
geeglm(formula = Response ~ Th + Sex + Age + Age2, family = binomial(link = logit),
       id = N, corstr = "exchangeable")

Coefficients:
            Estimate Std.err Wald Pr(>|W|)
(Intercept) -4.74983  1.82121 6.80   0.0091 **
Th2          -0.67398  0.33421 4.07   0.0437 *
Sex1          0.26569  0.36673 0.52   0.4688
Age           0.38115  0.12583 9.17   0.0025 **
Age2          -0.00612  0.00198 9.52   0.0020 **

---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Correlation structure = exchangeable
Estimated Scale Parameters:

            Estimate Std.err
(Intercept)    1.01    0.103
Link = identity

Estimated Correlation Parameters:
            Estimate Std.err
alpha        0.608   0.0883
Number of clusters: 127 Maximum cluster size: 3

Model with exponential correlation structure:

> gee3b <- geeglm(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit), corstr="ar1")

> summary(gee3b)

Call:
geeglm(formula = Response ~ Th + Sex + Age + Age2, family = binomial(link = logit),
       id = N, corstr = "ar1")

Coefficients:
            Estimate Std.err Wald Pr(>|W|)
(Intercept) -4.72712  1.78605 7.00   0.0081 **
Th2          -0.74443  0.32828 5.14   0.0233 *
Sex1          0.19674  0.36257 0.29   0.5874
Age           0.38467  0.12338 9.72   0.0018 **
Age2          -0.00621  0.00195 10.17  0.0014 **

---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Correlation structure = ar1
Estimated Scale Parameters:

          Estimate Std.err
(Intercept)      1   0.102
Link = identity

Estimated Correlation Parameters:
          Estimate Std.err
alpha      0.676  0.0766
Number of clusters: 127 Maximum cluster size: 3

```

For comparison a simple GLM with logit-link is fitted with the same covariates as in the marginal models above:

```

> glm1 <- glm(Response ~ Th + Sex + Age + Age2,
+ family=binomial(link=logit))
> summary(glm1)

Call:
glm(formula = Response ~ Th + Sex + Age + Age2, family = binomial(link = logit))

Deviance Residuals:
    Min      1Q  Median      3Q     Max
-1.821 -1.139  0.740  0.991  1.705

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept) -4.74983   1.26693  -3.75  0.00018 ***
Th2         -0.67398   0.22171  -3.04  0.00237 **
Sex1        0.26569   0.23965   1.11  0.26758
Age          0.38115   0.08674   4.39  1.1e-05 ***
Age2        -0.00612   0.00137  -4.47  7.6e-06 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 513.32 on 380 degrees of freedom
Residual deviance: 478.61 on 376 degrees of freedom
AIC: 488.6

```

Number of Fisher Scoring iterations: 4

It is often advantageous to center the variables like age around a value in the middle of its range. So now the marginal models from above are replicated with age centered around 30 years.

```

> Age <- Age-30
> Age2 <- Age^2

```

Again we use the function "gee" from the "gee"-library for fitting those models.

Model with independent correlation structure and centered age:

```
> gee1c <- gee(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit))

> summary(gee1c)

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: Independent

Call:
gee(formula = Response ~ Th + Sex + Age + Age2, id = N, family = binomial(link = logit))

Summary of Residuals:
    Min      1Q Median      3Q      Max
-0.809 -0.477  0.239  0.388  0.766

Coefficients:
            Estimate Naive S.E. Naive z Robust S.E. Robust z
(Intercept) 1.17278   0.28476    4.12    0.44806   2.617
Th2        -0.67398   0.22383   -3.01    0.33421  -2.017
Sex1        0.26569   0.24193    1.10    0.36673   0.724
Age         0.01369   0.01204    1.14    0.01736   0.789
Age2       -0.00612   0.00138   -4.43    0.00198  -3.086

Estimated Scale Parameter: 1.02
Number of Iterations: 1

Working Correlation
 [,1] [,2] [,3]
[1,]    1    0    0
[2,]    0    1    0
[3,]    0    0    1
```

Model with exchangeable correlation structure and centered age:

```
> gee2c <- gee(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit), corstr="exchangeable")

> summary(gee2c)

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 = Response ~ Th + Sex + Age + Age2, id = N, family = binomial(link = logit),
     corstr = "exchangeable")

Summary of Residuals:
    Min      1Q Median      3Q      Max
-0.809 -0.477  0.239  0.388  0.766

Coefficients:
            Estimate Naive S.E. Naive z Robust S.E. Robust z
(Intercept) 1.17278   0.42386   2.767   0.44806   2.617
Th2         -0.67398   0.33316  -2.023   0.33421  -2.017
Sex1        0.26569   0.36012   0.738   0.36673   0.724
Age          0.01369   0.01792   0.764   0.01736   0.789
Age2        -0.00612   0.00206  -2.978   0.00198  -3.086

Estimated Scale Parameter: 1.02
Number of Iterations: 1

Working Correlation
[,1] [,2] [,3]
[1,] 1.000 0.608 0.608
[2,] 0.608 1.000 0.608
[3,] 0.608 0.608 1.000

Model with exponential correlation structure and centered age:

> gee3c <- gee(Response ~ Th + Sex + Age + Age2, id=N,
+ family=binomial(link=logit), corstr="AR-M", Mv=1)

> summary(gee3c)

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: AR-M , M = 1

Call:
gee(formula = Response ~ Th + Sex + Age + Age2, id = N, family = binomial(link = logit),
     corstr = "AR-M", Mv = 1)

```

Summary of Residuals:

	Min	1Q	Median	3Q	Max
	-0.806	-0.467	0.235	0.383	0.793

Coefficients:

	Estimate	Naive S.E.	Naive z	Robust S.E.	Robust z
(Intercept)	1.22675	0.42743	2.870	0.44356	2.766
Th2	-0.74850	0.33391	-2.242	0.32808	-2.281
Sex1	0.19277	0.36195	0.533	0.36247	0.532
Age	0.01197	0.01797	0.666	0.01699	0.704
Age2	-0.00622	0.00207	-2.995	0.00195	-3.195

Estimated Scale Parameter: 1.02

Number of Iterations: 3

Working Correlation

	[,1]	[,2]	[,3]
[1,]	1.000	0.706	0.498
[2,]	0.706	1.000	0.706
[3,]	0.498	0.706	1.000