

# Statistical Analysis of Preference Data – Euler Group

Tuesday 21 January, 2020

## Methods and Software

Statistical analysis was based on the local odds ratios Generalized Estimating Equations method (Touloumis, Agresti, and Kateri 2013) as this is implemented in the R (R Core Team 2019) package `multgee` (Touloumis 2015). In addition, the function `ComparisonStats` was developed to evaluate the statistical significance of the desired comparisons for the accuracy and time data.

```
> ComparisonStats <- function(FittedModel, Lmatrix, alpha = 0.05) {
+   Lmatrix <- matrix(Lmatrix, nrow = 1)
+   ModelBetas <- FittedModel$coefficients
+   ModelVCov <- FittedModel$robust.variance
+   Estimate <- drop(Lmatrix %*% ModelBetas)
+   SdError <- sqrt(drop(Lmatrix %*% ModelVCov %*% t(Lmatrix)))
+   CBs <- Estimate + qnorm(c(alpha/2, 1 - alpha/2)) * SdError
+   pvalue <- 2 * pnorm(-abs(Estimate/SdError))
+   ans <- c(exp(c(Estimate, CBs)), round(pvalue, 4))
+   names(ans) <- c("Estimate", paste0((1 - alpha) * 100, "% LB"), paste0((1 - alpha) *
+     100, "% UB"), "p-value")
+   ans
+ }
```

## Import Data

The preference data for the Euler group were obtained by executing the following commands:

```
> library("readxl")
> preference_data <- read_excel("data/mainAllPreference_GS.xlsx")
> names(preference_data) <- gsub("[.]", "_", make.names(names(preference_data), unique = TRUE))
> euler_preference_data <- preference_data[preference_data$Notation == "Euler", ]
> euler_preference_data <- euler_preference_data[euler_preference_data$Preference_Value ==
+   1, ]
```

## Analysis of Preference Data

First, the following regression model was fitted to the preference data

$$\log \left[ \frac{\Pr(Y_{ij} = k)}{\Pr(Y_{ij} = 4)} \right] = \beta_{0k} + \beta_{1k} x_{ij1}$$

for  $k = 1, 2, 3$  and where

- $\Pr(Y_{ij} = k)$  is the probability for participant  $i$  to find preferable diagram  $k$  (1=NWM, 2=NWMA, 3=WM and 4=WMA) for task  $j$  (1=subset, 2=disjoint),
- $x_{ij1}$  is the indicator for the *subset* task type,

for  $i = 1, \dots, 103$ , corresponding to the individual participants.

```
> library("multgee")
> fitted_model <- nomLORgee(formula = Treatment ~ Question_Type, id = Participant,
+                           data = euler_preference_data, LORstr = "independence")
```

By comparing this model to the null model

```
> reduced_model <- nomLORgee(formula = Treatment ~ 1, id = Participant,
+                             data = euler_preference_data, LORstr = "independence")
> Waldts <- waldts(fitted_model, reduced_model)
> Waldts
Goodness of Fit based on the Wald test

Model under H_0: Treatment ~ 1
Model under H_1: Treatment ~ Question_Type

Wald Statistic=2.5212, df=3, p-value=0.4715
> fitted_model <- reduced_model
```

we can conclude that the preference of a treatment did not depend on the task type ( $p$  - value = 0.4715).

The regression model was further simplified to

$$\log \left[ \frac{\Pr(Y_{ij} = k)}{\Pr(Y_{ij} = 4)} \right] = \beta_{0k}$$

for  $k = 1, 2, 3$ .

## Comparison of treatments

WM > WMA  $\approx$  NWM > NWMA

```
> ## NWMA vs. NWM
> ComparisonStats(fitted_model, c(1, 0, -1))
  Estimate    95% LB    95% UB  p-value
0.27272727 0.09006333 0.82586515 0.02150000
> ## WM vs. NWM
> ComparisonStats(fitted_model, c(0, 0, -1))
  Estimate    95% LB    95% UB  p-value
17.27272727 9.147204 32.616207 0.00000000
> ## WMA vs. NWM
> ComparisonStats(fitted_model, c(1, 0, -1))
  Estimate    95% LB    95% UB  p-value
0.27272727 0.09006333 0.82586515 0.02150000
> ## NWMA vs. WM
> ComparisonStats(fitted_model, c(1, 0, 0))
  Estimate    95% LB    95% UB  p-value
0.015789474 0.005125851 0.048637284 0.000000000
```

```
> ## NWMA vs. WMA
> ComparisonStats(fitted_model, c(1, -1, 0))
  Estimate    95% LB    95% UB  p-value
0.17647059 0.05887137 0.52898159 0.00200000
> ## WM vs. WMA
> ComparisonStats(fitted_model, c(0, -1, 0))
  Estimate    95% LB    95% UB  p-value
11.176471  7.139097 17.497099 0.000000
```

## References

- R Core Team. 2019. *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R Foundation for Statistical Computing. <http://www.R-project.org/>.
- Touloumis, A. 2015. “R Package Multgee: A Generalized Estimating Equations Solver for Multinomial Responses.” *Journal of Statistical Software* 64 (8): 1–14.
- Touloumis, A., A. Agresti, and M. Kateri. 2013. “Generalized Estimating Equations for Multinomial Responses Using a Local Odds Ratios Parameterization.” *Biometrics* 69 (3): 633–40.