Featback analyses: GSES

Data preparation

Primaire uitkomstmaat

• Eetstoornispathologie (EDE-Q_TOT)

Secundaire uitkomstmaten

- Self-efficacy (GSES_TOT)
- Angst en depressie (PHQ_TOT)
- Sociale steun (SSL TOT)

Conditie indicatoren

- 1 = Featback
- 2 = Featback + ondersteuning van een ervaringsdeskundige via chat of email
- 3 = Ondersteuning van een ervaringsdeskundige via chat of email
- 4 = Wachtlijst controle conditie (hen werd conditie 2 aangeboden na 12 maanden + 8 weken wachttijd)

Tijdsindicatoren

- 1 = baseline
- 2 = post-interventie (8 weken)
- 3 = 3 maanden follow-up (i.e., 3 maanden + 8 weken)
- 4 = 6 maanden FU
- 5 = 9 maanden FU
- 6 = 12 maanden FU

Mogelijke moderatoren

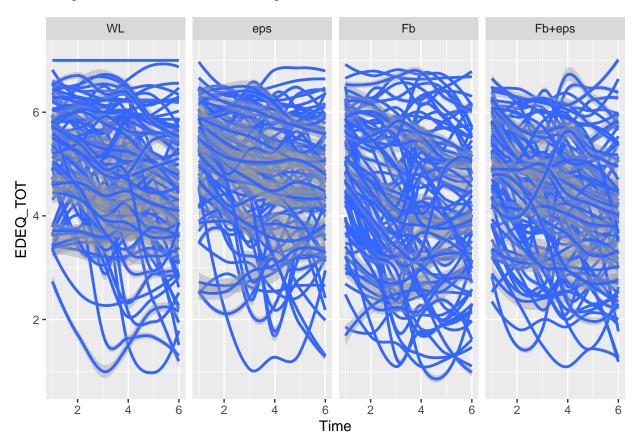
- Leeftijd (Age)
- Educatie (moet nog omgezet worden naar 3 levels; laag, middel, hoog) (T0_edu)
- Behandelgeschiedenis (T0 treatment)
- BMI bij Baseline (BMI)
- Eetstoornispathologie bij baseline (EDEQ TOT)
- (Evt. duur eetstoornis, maar is zeer hoog gecorreleerd met leeftijd; T0_yrsED)
- Motivatie om te veranderen bij baseline (T0_Motiv_TOT)
- Zelfwaardering bij baseline (T0_RSES_TOT)
- Self-efficacy bij baseline (GSES TOT)
- Angst en depressie bij baseline (PHQ TOT)
- Type eetstoornis bij baseline (staat nog niet tussen de variabelen; nog even kijken hoe we dit moeten aanpakken . . .)

Toevoeging 06-08-2021:

- T0_EDEQ_ObjEet_Keer (= aantal (objectieve) eetbuien in de afgelopen 28 dagen)
- TO SSL TOT

```
#names(data)
## Set aprropriate variable classes
data$ID <- factor(data$ID)</pre>
data$T0 edu <- ordered(data$T0 edu)</pre>
## Abbreviate condition levels
levels(data$Condition)[levels(data$Condition)=="Waiting list"] <- "WL"</pre>
levels(data$Condition) [levels(data$Condition) == "Featback"] <- "Fb"</pre>
levels(data$Condition) [levels(data$Condition) == "Featback + expert-patient support"] <- "Fb+eps"</pre>
levels(data$Condition)[levels(data$Condition)=="expert-patient support"] <- "eps"</pre>
data$Condition <- factor(data$Condition, levels = c("WL", "eps", "Fb", "Fb+eps"))
## Check if every subject has a time 1
all(table(data$ID, data$Time == 1)[,2] == 1L)
## [1] FALSE
## Select only pre- and post assessments
#data <- data[data$Time %in% 1:2, ]
#data$Time <- factor(data$Time)</pre>
## Construct TO variables
for (i in unique(data$ID)) {
  data$TO EDEQ TOT[data$ID == i] <- data$EDEQ TOT[data$ID == i & data$Time == 1]</pre>
  data$TO_BMI[data$ID == i] <- data$BMI[data$ID == i & data$Time == 1]</pre>
  data$TO_GSES_TOT[data$ID == i] <- data$GSES_TOT[data$ID == i & data$Time == 1]</pre>
  data$TO_PHQ_TOT[data$ID == i] <- data$PHQ_TOT[data$ID == i & data$Time == 1]
  data$T0_SSL_TOT[data$ID == i] <- data$SSL_TOT[data$ID == i & data$Time == 1]
  data$TO_eetbuien[data$ID == i] <- data$EDEQ_ObjEet_Keer[data$ID == i & data$Time == 1]
}
## Check for missings and remove
unlist(sapply(data[ , c("Age", "T0_edu", "T0_BMI", "T0_EDEQ_TOT", "T0_Motiv_TOT",
                         "TO_GSES_TOT", "TO_PHQ_TOT")], function(x) table(is.na(x))))
##
            Age.FALSE
                             TO edu.FALSE
                                                  TO edu.TRUE
                                                                     TO BMI.FALSE
##
               215130
                                    215124
                                                                            215124
##
          TO_BMI.TRUE
                        TO_EDEQ_TOT.FALSE
                                             TO_EDEQ_TOT.TRUE TO_Motiv_TOT.FALSE
##
                     6
                                                            18
##
    TO_Motiv_TOT.TRUE
                       TO_GSES_TOT.FALSE
                                             TO_GSES_TOT.TRUE
                                                                 TO_PHQ_TOT.FALSE
##
                    18
                                    215112
                                                                            215130
data <- data[!is.na(data$T0_BMI), ]</pre>
data <- data[!is.na(data$TO_EDEQ_TOT), ]</pre>
data <- data[!is.na(data$T0_edu), ]</pre>
data <- data[!is.na(data$T0_yrsED), ]</pre>
unlist(sapply(data[ , c("Age", "T0_edu", "T0_BMI", "T0_EDEQ_TOT", "T0_Motiv_TOT",
                         "TO_GSES_TOT", "TO_PHQ_TOT")], function(x) table(is.na(x))))
##
            Age.FALSE
                             TO_edu.FALSE
                                                 TO_BMI.FALSE
                                                                TO_EDEQ_TOT.FALSE
##
               215058
                                    215058
                                                        215058
                                                                            215058
## TO_Motiv_TOT.FALSE
                       TO_Motiv_TOT.TRUE
                                            TO_GSES_TOT.FALSE
                                                                 TO_PHQ_TOT.FALSE
##
               215043
                                        15
                                                        215058
                                                                            215058
## Exploratory plot
library("ggplot2")
ggplot(data = data, aes(x = Time, y = EDEQ_TOT, group = ID)) +
 facet_grid(. ~ Condition) + geom_smooth()
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



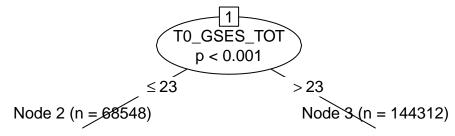


De trajecten lijken niet erg lineair.

Splits based only on treatment-time-subgroup interactions

```
library("glmertree")
library("lmerTest")
library("strucchange")
dim(data)
## [1] 215058
                  70
dat1 <- data[!is.na(data$GSES_TOT),]</pre>
dim(dat1)
## [1] 214549
## Weird: still some missing values
## Seem to occur when imputation_ is 0
## table(dat1$ID, dat1$Imputation_)[ , 1] ## not all sixes
## table(dat1$ID, dat1$Imputation_)[ , 2] ## all sixes
dat1 <- dat1[dat1$Imputation_ != 0, ]</pre>
dim(dat1)
## [1] 212980
                  70
```

```
lt1 <- lmertree(GSES_TOT ~ Time*Condition | (1|ID) | Age + TO_edu + TO_BMI + TO_EDEQ_TOT +
                  TO_Motiv_TOT + TO_GSES_TOT + TO_PHQ_TOT + TO_treatment + TO_yrsED +
                  TO_SSL_TOT + TO_eetbuien,
                data = dat1, cluster = ID, parm = 6:8, maxdepth = 4, verbose = TRUE)
## 'log Lik.' -567916.8 (df=18)
## 'log Lik.' -567916.8 (df=18)
if (length(lt1$tree) > 1L) {
  plot(lt1, type = "simple", which = "tree", fitted = "marginal", gp = gpar(cex = .5))
  #fixef(lt1)
  apply(fixef(lt1), 2, sd)
 VarCorr(lt1)
 tmp <- summary(lmer(attr(lt1$lmer, "call")$formula, data = lt1$data))</pre>
 round(tmp$coefficients[tmp$coefficients[,"Pr(>|t|)"] < 0.05, -3], digits = 4)</pre>
} else {
  sctest(lt1$tree)
}
```



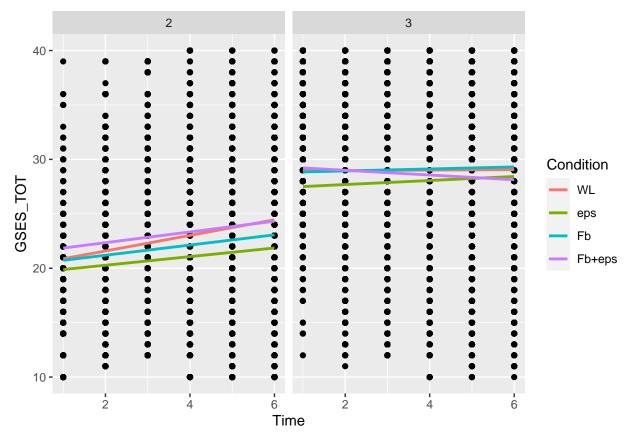
Fixed effects:
(Intercept) 24.442
Time 0.716
Conditioneps -4.956
ConditionFb -4.184
ConditionFb+eps -3.082
Time:Conditinps -0.320
Time:ConditinFb -0.248
Time:CndtnFb+ps -0.222

Fixed effects:
(Intercept) 27.059
Time 0.021
Conditioneps 0.259
ConditionFb 1.693
ConditionFb+eps 2.375
Time:Conditinps 0.164
Time:ConditinFb 0.071
Time:CndtnFb+ps -0.239

```
Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                              24.4421
                                         0.4234 57.7235 0.0000
                                          0.1826 14.3275
                                                           0.0000
## .tree3
                               2.6167
## .tree2:Time
                               0.7161
                                          0.0158 45.2135
                                                          0.0000
## .tree3:Time
                              0.0213
                                          0.0105
                                                 2.0404
                                                          0.0413
## .tree2:Conditioneps
                                          0.7590 -6.5327
                                                           0.0000
                              -4.9583
## .tree2:ConditionFb
                              -4.1844
                                          0.8388 - 4.9885
                                                           0.0000
## .tree3:ConditionFb
                              1.6941
                                          0.6410 2.6429
                                                           0.0086
```

```
0.9201 -3.3497
## .tree2:ConditionFb+eps
                                -3.0821
                                                               0.0009
## .tree3:ConditionFb+eps
                                            0.6181
                                                     3.8439
                                                               0.0001
                                 2.3761
## .tree2:Time:Conditioneps
                                -0.3199
                                            0.0209 -15.3194
                                                               0.0000
## .tree3:Time:Conditioneps
                                 0.1639
                                            0.0157 10.4425
                                                              0.0000
## .tree2:Time:ConditionFb
                                -0.2478
                                            0.0223 -11.1336
                                                              0.0000
## .tree3:Time:ConditionFb
                                 0.0708
                                            0.0150
                                                     4.7360
                                                              0.0000
## .tree2:Time:ConditionFb+eps -0.2221
                                            0.0237 -9.3677
                                                               0.0000
## .tree3:Time:ConditionFb+eps -0.2390
                                            0.0145 - 16.4870
                                                              0.0000
ggplot(data = lt1$data, aes(x = Time, y = GSES_TOT, group = ID)) +
  geom_point() + facet_grid(. ~ .tree) +
  geom_smooth(method = "lm", se = FALSE,
              aes(x = Time, y = GSES_TOT, group = Condition, color = Condition))
```

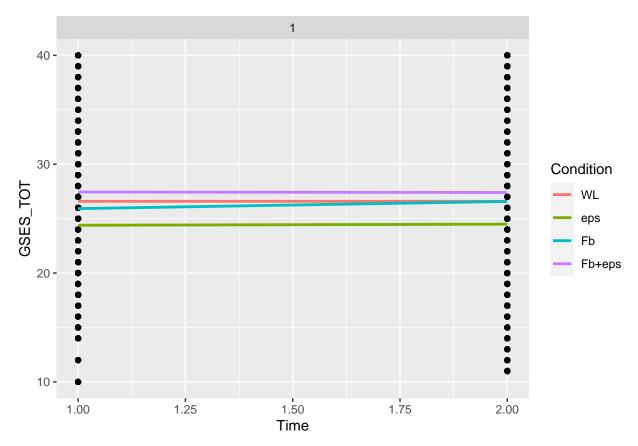
`geom_smooth()` using formula 'y ~ x'



```
## Warning in lmertree(GSES_TOT ~ Time * Condition | (1 | ID) | Age + T0_edu + :
## 'data' contains missing values, note that listwise deletion will be employed.
## 'log Lik.' -169525.9 (df=10)
## 'log Lik.' -169525.9 (df=10)
```

```
if (length(lt3$tree) > 1L) {
  plot(lt3, which = "tree", type = "simple", fitted = "marginal", gp = gpar(cex = .7))
  #fixef(lt3)
  apply(fixef(lt3), 2, sd)
  VarCorr(1t3)
  tmp <- summary(lmer(attr(lt3$lmer, "call")$formula, data = lt3$data))</pre>
 round(tmp$coefficients[tmp$coefficients[,"Pr(>|t|)"] < 0.05, -3], digits = 4)</pre>
  sctest(lt3$tree)
}
##
                       TO_edu TO_BMI TO_EDEQ_TOT TO_Motiv_TOT TO_GSES_TOT
                  Age
## statistic 3.424619 11.94041 7.346604 6.9026259
                                                       6.3652566 10.7539192
                                          0.9999645
            1.000000 1.00000 0.999826
                                                       0.9999967
                                                                   0.9101576
             TO_PHQ_TOT TO_treatment TO_yrsED TO_SSL_TOT TO_eetbuien
## statistic 8.3151584
                           1.8172927 12.0804496 10.1861647
                                                              4.527186
                                                              1.000000
                           0.9999693 0.7655654 0.9495224
              0.9975622
ggplot(data = 1t3\$data, aes(x = Time, y = GSES_TOT, group = ID)) + geom_point() +
 facet_grid(. ~ .tree) +
  geom_smooth(method = "lm", se = FALSE,
              aes(x = Time, y = GSES\_TOT, group = Condition, color = Condition))
```

`geom_smooth()` using formula 'y ~ x'



Waitlist versus all treatment groups

```
dat1$Condition2 <- factor(dat1$Condition == "WL")</pre>
lt2 <- lmertree(GSES_TOT ~ Time*Condition2 | (1|ID) | Age + TO_edu + TO_BMI + TO_EDEQ_TOT +
                  TO_Motiv_TOT + TO_GSES_TOT + TO_PHQ_TOT + TO_treatment + TO_yrsED +
                  TO_SSL_TOT + TO_eetbuien,
                data = dat1, cluster = ID, parm = 4, verbose = TRUE, maxdepth = 4)
## 'log Lik.' -569881.2 (df=6)
## 'log Lik.' -569881.2 (df=6)
if (length(lt2$tree) > 1L) {
  plot(1t2, which = "tree", fitted = "marginal", gp = gpar(cex = .5))
  #fixef(lt2)
  apply(fixef(lt2), 2, sd)
  VarCorr(1t2)
  tmp <- summary(lmer(attr(lt2$lmer, "call")$formula, data = lt1$data))</pre>
  round(tmp\$coefficients[tmp\$coefficients[,"Pr(>|t|)"] < 0.05, -3], digits = 4)
  sctest(lt2$tree)
                   Age
                        T0 edu
                                  TO_BMI TO_EDEQ_TOT TO_Motiv_TOT TO_GSES_TOT
## statistic 0.9517267 2.586386 2.219938
                                           2.4330414
                                                         1.323522 13.12622010
            1.0000000 1.000000 1.000000 0.9999998
                                                         1.000000 0.07800239
             TO_PHQ_TOT TO_treatment TO_yrsED TO_SSL_TOT TO_eetbuien
                           2.0331237 2.598200 6.9862055
## statistic 7.5711516
                                                            2.085639
              0.6748673
## p.value
                           0.8409198 0.999999 0.7720581
                                                             1.000000
dat2 <- dat1[dat1$Time %in% 1:2, ]</pre>
1t4 <- lmertree(GSES_TOT ~ Time*Condition2 | (1|ID) | Age + TO_edu + TO_BMI + TO_EDEQ_TOT +
                  TO_Motiv_TOT + TO_GSES_TOT + TO_PHQ_TOT + TO_treatment + TO_yrsED +
                  T0_SSL_TOT + T0_eetbuien,
                data = dat2, cluster = ID, parm = 4, maxdepth = 4)
if (length(lt4$tree) > 1L) {
  plot(lt4, which = "tree", fitted = "marginal", gp = gpar(cex = .7))
  #fixef(lt4)
  apply(fixef(lt4), 2, sd)
  VarCorr(lt4)
  tmp <- summary(lmer(attr(lt4$lmer, "call")$formula, data = lt3$data))</pre>
  round(tmp$coefficients[tmp$coefficients[,"Pr(>|t|)"] < 0.05, -3], digits = 4)
} else {
  sctest(lt4$tree)
}
##
                   Age
                         T0_edu
                                   TO_BMI TO_EDEQ_TOT TO_Motiv_TOT TO_GSES_TOT
## statistic 6.3555508 3.776758 2.7211261
                                            0.9556429
                                                         7.8818450
            0.8633832 1.000000 0.9999972
                                            1.0000000
                                                                      0.9999799
## p.value
                                                         0.6213518
             TO_PHQ_TOT TO_treatment TO_yrsED TO_SSL_TOT TO_eetbuien
## statistic 0.8198438 0.005647743 5.2778893 2.9712962 4.3108138
## p.value
             1.0000000 1.000000000 0.9642137 0.9999834 0.9953185
```