

Featback analyses: SSL

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)
- T0_SSL_TOT

```
## Read in data
library("foreign")
data <- read.spss("20210316_Featback_AllMerged_LongFormat_imputed.sav", to.data.frame = TRUE)
#names(data)
```

```

## Set appropriate variable classes
data$ID <- factor(data$ID)
data$T0_edu <- ordered(data$T0_edu)
## Abbreviate condition levels
levels(data$Condition) [levels(data$Condition)== "Waiting list"] <- "WL"
levels(data$Condition) [levels(data$Condition)== "Featback"] <- "Fb"
levels(data$Condition) [levels(data$Condition)== "Featback + expert-patient support"] <- "Fb+eps"
levels(data$Condition) [levels(data$Condition)== "expert-patient support"] <- "eps"
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$Time)
## Construct T0 variables
for (i in unique(data$ID)) {
  data$T0_EDEQ_TOT[data$ID == i] <- data$EDEQ_TOT[data$ID == i & data$Time == 1]
  data$T0_BMI[data$ID == i] <- data$BMI[data$ID == i & data$Time == 1]
  data$T0_GSES_TOT[data$ID == i] <- data$GSES_TOT[data$ID == i & data$Time == 1]
  data$T0_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$Time == 1]
  data$T0_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",
  "T0_GSES_TOT", "T0_PHQ_TOT")], function(x) table(is.na(x))))
```

##	Age.FALSE	T0_edu.FALSE	T0_edu.TRUE	T0_BMI.FALSE
##	215130	215124	6	215124
##	T0_BMI.TRUE	T0_EDEQ_TOT.FALSE	T0_EDEQ_TOT.TRUE	T0_Motiv_TOT.FALSE
##	6	215112	18	215112
##	T0_Motiv_TOT.TRUE	T0_GSES_TOT.FALSE	T0_GSES_TOT.TRUE	T0_PHQ_TOT.FALSE
##	18	215112	18	215130

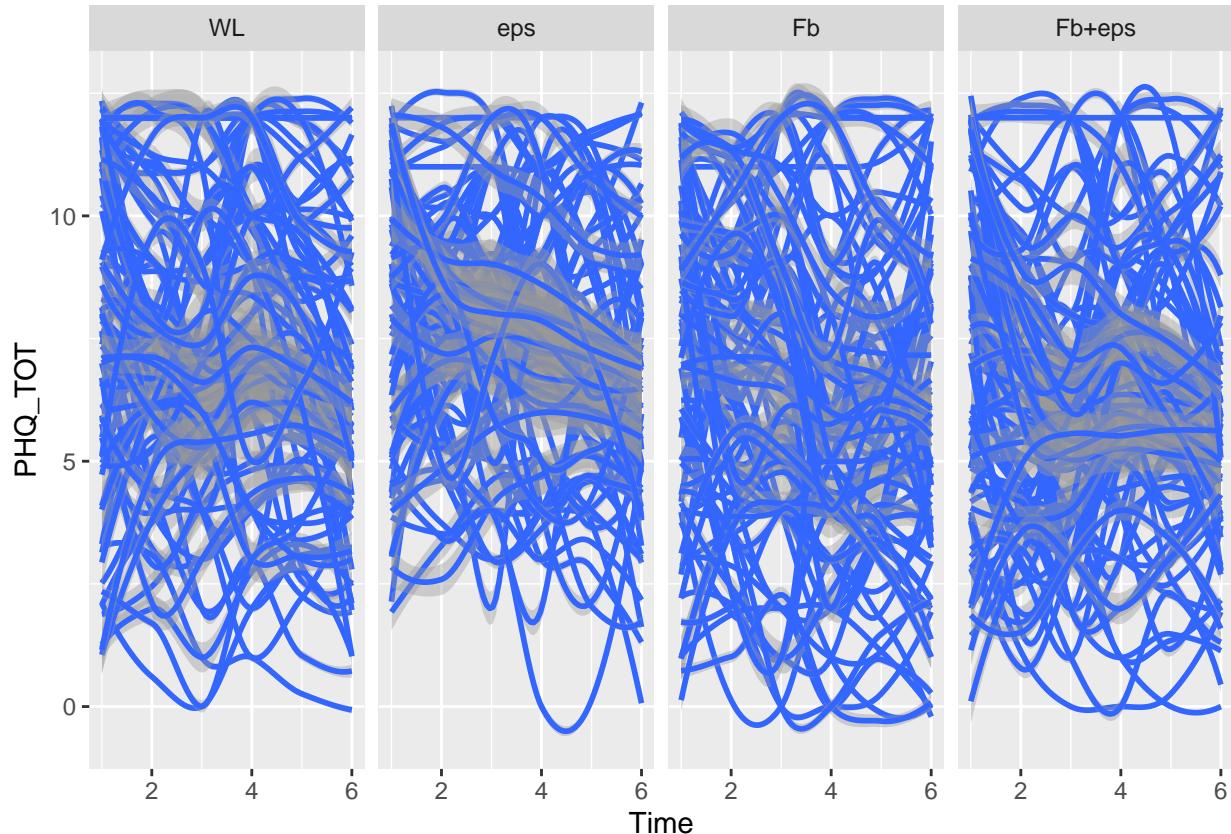
```

data <- data[!is.na(data$T0_BMI), ]
data <- data[!is.na(data$T0_EDEQ_TOT), ]
data <- data[!is.na(data$T0_edu), ]
data <- data[!is.na(data$T0_yrsED), ]
unlist(sapply(data[ , c("Age", "T0_edu", "T0_BMI", "T0_EDEQ_TOT", "T0_Motiv_TOT",
  "T0_GSES_TOT", "T0_PHQ_TOT")], function(x) table(is.na(x))))
```

##	Age.FALSE	T0_edu.FALSE	T0_BMI.FALSE	T0_EDEQ_TOT.FALSE
##	215058	215058	215058	215058
##	T0_Motiv_TOT.FALSE	T0_Motiv_TOT.TRUE	T0_GSES_TOT.FALSE	T0_PHQ_TOT.FALSE
##	215043	15	215058	215058

Exploratory plot
library("ggplot2")
ggplot(data = data, aes(x = Time, y = PHQ_TOT, group = ID)) +
 facet_grid(. ~ Condition) + geom_smooth()

`geom_smooth()` using method = 'loess' and formula 'y ~ x'
Warning: Removed 477 rows containing non-finite values (stat_smooth).



De trajecten lijken niet erg lineair.

Splits based only on treatment-time-subgroup interactions

```

library("glmertree")
library("lmerTest")
library("strucchange")
dat1 <- data[!is.na(data$SSL_TOT),]
lt1 <- lmertree(SSL_TOT ~ Time*Condition | (1|ID) | Age + T0_edu + T0_BMI + T0_EDEQ_TOT +
  T0_Motiv_TOT + T0_GSES_TOT + T0_PHQ_TOT + T0_treatment + T0_yrsED +
  T0_SSL_TOT + T0_eetbuien,
  data = dat1, cluster = ID, parm = 6:8, maxdepth = 4L)
if (length(lt1$tree) > 1L) {
  plot(lt1, which = "tree", type = "simple", 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$data))
  round(tmp$coefficients[tmp$coefficients[, "Pr(>|t|)"] < 0.05, -3], digits = 4)
} else {
  sctest(lt1$tree)
}

##          Age     T0_edu      T0_BMI T0_EDEQ_TOT T0_Motiv_TOT T0_GSES_TOT
## statistic 3.212595 14.75395 12.7361413     4.633832      2.17896    8.7160771

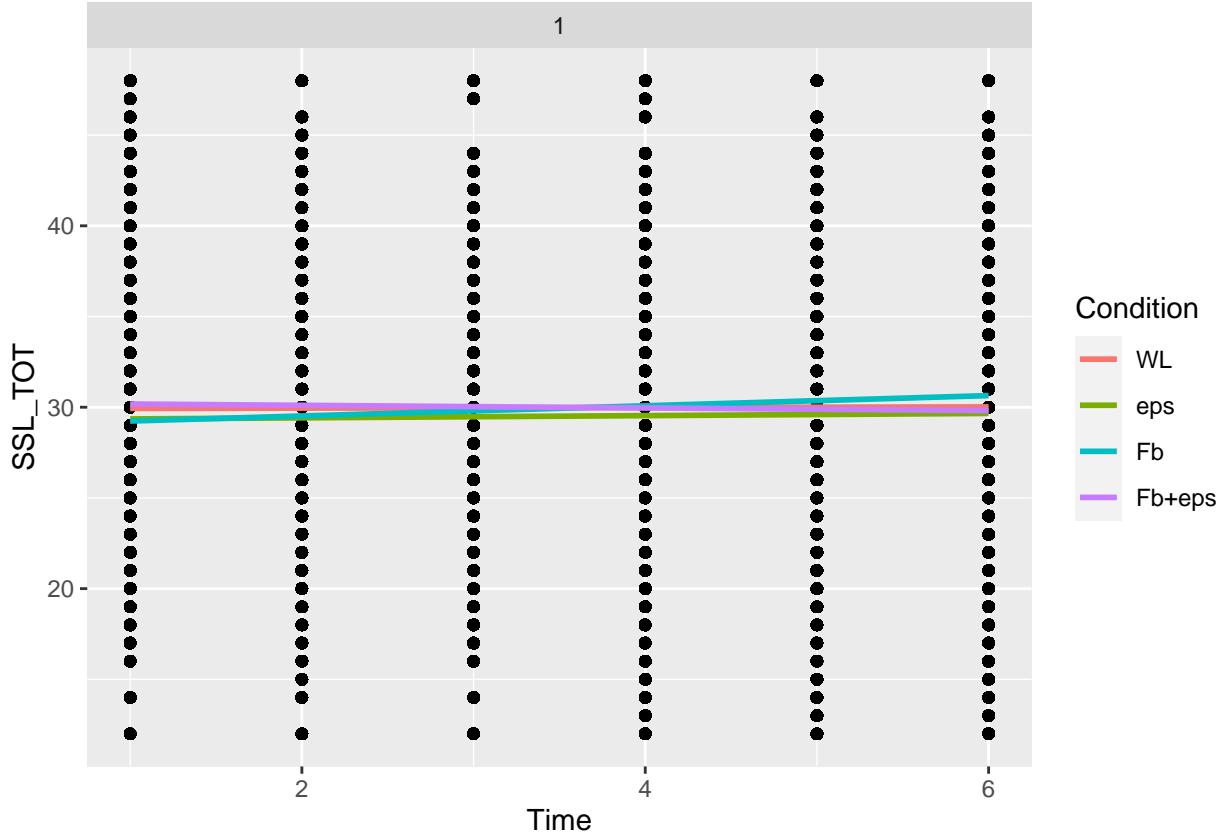
```

```

## p.value    1.000000 1.000000 0.6751708     1.000000      1.000000 0.9943708
##          T0_PHQ_TOT T0_treatment T0_yrsED T0_SSL_TOT T0_eetbuien
## statistic  6.3810384    2.9575271 3.818895    9.583726     3.379898
## p.value    0.9999964    0.9962519 1.000000    0.976356     1.000000
ggplot(data = lt1$data, aes(x = Time, y = SSL_TOT, group = ID)) + geom_point() +
  facet_grid(. ~ .tree) + geom_smooth(method = "lm", se = FALSE,
aes(x = Time, y = SSL_TOT, group = Condition, color = Condition))

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

```



```

dat2 <- dat1[dat1$Time %in% 1:2, ]
lt3 <- lmertree(SSL_TOT ~ Time*Condition | (1|ID) | Age + T0_edu + T0_BMI + T0_EDEQ_TOT +
  T0_Motiv_TOT + T0_GSES_TOT + T0_PHQ_TOT + T0_treatment + T0_yrsED +
  T0_SSL_TOT + T0_eetbuien,
  data = dat2, cluster = ID, parm = 6:8, maxdepth = 4L)

## Warning in lmertree(SSL_TOT ~ Time * Condition | (1 | ID) | Age + T0_edu + :
## 'data' contains missing values, note that listwise deletion will be employed.

if (length(lt3$tree) > 1L) {
  plot(lt3, which = "tree", fitted = "marginal", gp = gpar(cex = .7))
  #fixef(lt3)
  apply(fixef(lt3), 2, sd)
  VarCorr(lt3)
  tmp <- summary(lmer(attr(lt3$lmer, "call")$formula, data = lt3$data))
  round(tmp$coefficients[tmp$coefficients[, "Pr(>|t|)"] < 0.05, -3], digits = 4)
} else {

```

```

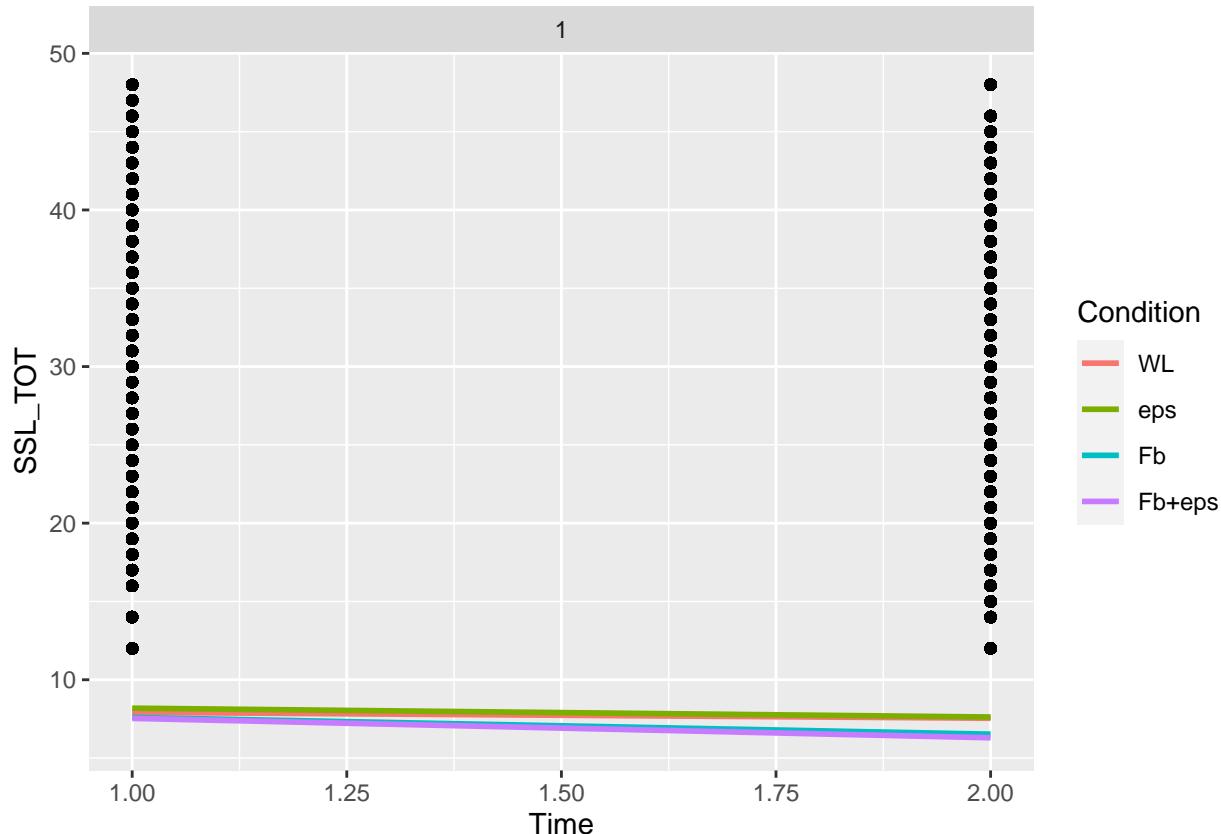
sctest(lt3$tree)
}

##          Age      T0_edu      T0_BMI TO_EDEQ_TOT TO_Motiv_TOT TO_GSES_TOT
## statistic 7.2848711 19.2704570 14.5821287 11.2719323   2.089027  7.3084053
## p.value   0.9998582  0.9999996  0.4176365  0.8617574   1.000000  0.9998466
##          T0_PHQ_TOT T0_treatment  T0_yrsED T0_SSL_TOT T0_eetbuien
## statistic  3.712911   2.6065450 11.0870537  6.8909168  9.9394895
## p.value    1.000000   0.9987738  0.8803574  0.9999661  0.9622441

ggplot(data = lt3$data, aes(x = Time, y = SSL_TOT, group = ID)) + geom_point() +
  facet_grid(. ~ .tree) +
  geom_smooth(method = "lm", se = FALSE,
              aes(x = Time, y = PHQ_TOT, group = Condition, color = Condition))

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

```



Waitlist versus all treatment groups

```

dat1$Condition2 <- factor(dat1$Condition == "WL")
lt2 <- lmertree(SSL_TOT ~ Time*Condition2 | (1|ID) | Age + T0_edu + T0_BMI + TO_EDEQ_TOT +
  TO_Motiv_TOT + TO_GSES_TOT + TO_PHQ_TOT + T0_treatment + T0_yrsED +
  T0_SSL_TOT + T0_eetbuien,
  data = dat1, cluster = ID, parm = 4, verbose = TRUE, maxdepth = 4)

## 'log Lik.' -627714.1 (df=6)

```

```

## 'log Lik.' -627714.1 (df=6)

if (length(lt2$tree) > 1L) {
  plot(lt2, which = "tree", fitted = "marginal", gp = gpar(cex = .5))
  #fixef(lt2)
  apply(fixef(lt2), 2, sd)
  VarCorr(lt2)
  tmp <- summary(lmer(attr(lt2$lmer, "call")$formula, data = lt1$data))
  round(tmp$coefficients[tmp$coefficients[, "Pr(>|t|)"] < 0.05, -3], digits = 4)
} else {
  sctest(lt2$tree)
}

##          Age      T0_edu      T0_BMI      T0_EDEQ_TOT      T0_Motiv_TOT      T0_GSES_TOT
## statistic 1.990454 11.5672224 3.655754      1.51911      1.099828      1.159244
## p.value   1.000000  0.8738927 0.999462      1.00000      1.000000      1.000000
##          T0_PHQ_TOT      T0_treatment      T0_yrsED      T0_SSL_TOT      T0_eetbuien
## statistic  4.726468    2.5428417 2.5691632     4.3731580      1.65689
## p.value    0.987215    0.7251981 0.9999992     0.9944749      1.00000
dat2 <- dat1[dat1$Time %in% 1:2, ]
lt4 <- lmertree(SSL_TOT ~ Time*Condition2 | (1|ID) | Age + T0_edu + T0_BMI + T0_EDEQ_TOT +
           T0_Motiv_TOT + T0_GSES_TOT + T0_PHQ_TOT + T0_treatment + T0_yrsED +
           T0_SSL_TOT + T0_eetbuien,
           data = dat2, cluster = ID, parm = 4, maxdepth = 4)
if (length(lt4$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$data))
  round(tmp$coefficients[tmp$coefficients[, "Pr(>|t|)"] < 0.05, -3], digits = 4)
} else {
  sctest(lt4$tree)
}

##          Age      T0_edu      T0_BMI      T0_EDEQ_TOT      T0_Motiv_TOT      T0_GSES_TOT
## statistic 8.6386290 2.992161 11.7645670     1.914564      1.629332      1.582679
## p.value   0.4939073 1.000000 0.1441475     1.000000      1.000000      1.000000
##          T0_PHQ_TOT      T0_treatment      T0_yrsED      T0_SSL_TOT      T0_eetbuien
## statistic  1.626673    2.7031307 12.3223994     4.880923     7.5871392
## p.value    1.000000    0.6867696 0.1124665     0.982407     0.6721118

```