

Supplementary Information for Doba et al. (Submitted)

March 13, 2026 - 10:27

Contents

1	Data management	2
1.1	Loading raw data	2
1.2	Pre-processing for missing data or errors	2
1.2.1	Double participation	2
1.2.2	Sex	3
1.2.3	Parental situation	3
1.2.4	Age	3
1.2.5	Missing SPS	3
1.2.6	Wrong scores to CTQ	3
1.3	Subset the database	3
2	Groups definition	4
2.1	Labeling data	4
2.1.1	Categories definitions	4
2.1.2	Comparison with latent profile analysis	7
2.2	Matching	8
2.2.1	Cardinality matching	8
2.2.2	Optimal pair matching	12
3	Data analysis	14
3.1	DERS	18
3.1.1	PCA description	18
3.1.2	First component analysis	19
3.1.3	Second component analysis	22
3.2	CERQ	26
3.2.1	PCA description	26
3.2.2	First component analysis	27
3.2.3	Second component analysis	30
3.3	DSI	34
3.3.1	PCA description	34
3.3.2	First component analysis	35
3.3.3	Second component analysis	38
3.4	HADS	42
3.4.1	PCA description	42
3.4.2	First component analysis	43
3.4.3	Second component analysis	46
3.5	SPS	50
3.5.1	PCA description	50
3.5.2	First component analysis	51
4	Session information	55

1 Data management

The database was saved in the file `data_converti.csv` (converted from the initial `.csv` file where the `#NULL!` were changed to `NA` using vim text editor command `:%s/#NULL!/NA`)

1.1 Loading raw data

```
data <- read.table('data/data_trauma.csv', sep=';', dec=',', header=TRUE,
                  colClasses=c(rep('factor',2), 'numeric',
                                rep('factor',2), rep("numeric", 34)))
nb.var <- ncol(data)
init.nb <- nrow(data)
print(names(data))
```

```
## [1] "Code.participant"      "Sexe"
## [3] "Age"                  "Situation.parentale"
## [5] "Niveau.etudes"        "CTQ.emotional.neglect"
## [7] "CTQ.physical.abuse"   "CTQ.emotional.abuse"
## [9] "CTQ.physical.neglect" "CTQ.sexual.abuse"
## [11] "CERQpositivefocusing" "CERQpositivereappraisal"
## [13] "CERQputtingintoperspective" "CERQrefocusonplanning"
## [15] "CERQacceptance"      "CERQrumination"
## [17] "CERQselfblame"       "CERQblamingother"
## [19] "CERQcatastrophizing" "DERSnonacceptance"
## [21] "DERSgoals"           "DERSimpulse"
## [23] "DERSawareness"       "DERSstrategies"
## [25] "DERSclarity"         "DERStotal"
## [27] "DSIemotionalreact"   "DSIemotionalcutoff"
## [29] "DSIfusionwithothers" "DSIiposition"
## [31] "DSItotal"            "HADSanxiety"
## [33] "HADSdepression"      "HADStotal"
## [35] "SPShopelessness"     "SPSsuicideideation"
## [37] "SPSnegativeselfevaluation" "SPShostility"
## [39] "SPS.total"
```

- The initial number of participants is `init.nb`: 508.
- The database has `nb.var`: 39 variables which names are given above.

1.2 Pre-processing for missing data or errors

```
attach(data)
```

The database is checked for errors that were noticed during data screening.

1.2.1 Double participation

Some participants appear twice:

```
tt <- table(Code.participant)
(twice <- as.data.frame(tt[tt!=1])$Code.participant)
```

```
## [1] C216 C364
## Levels: C216 C364
```

1.2.2 Sex

Sex variable is missing for some participants:

```
(sex.miss <- Code.participant[Sexe==''])
```

```
## [1] C1306
```

```
## 506 Levels: C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 C1009 C101 ... PD9
```

1.2.3 Parental situation

Parental situation is missing for some participants and bad encoding has been noticed:

```
(parental.miss <- Code.participant[Situation.parentale==''])
```

```
## [1] C760 C527 C135 C1220 C1141 C1321 C1132 C1134 C1242
```

```
## 506 Levels: C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 C1009 C101 ... PD9
```

```
(parental.error <- Code.participant[Situation.parentale=='15'])
```

```
## [1] C562
```

```
## 506 Levels: C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 C1009 C101 ... PD9
```

1.2.4 Age

Age is missing for some participants and some participants are older than 25 y.o.:

```
(age.miss <- Code.participant[is.na(Age)])
```

```
## [1] C1334 C1911
```

```
## 506 Levels: C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 C1009 C101 ... PD9
```

```
(age.old <- Code.participant[na.omit(Age)>25])
```

```
## [1] C744 C1133
```

```
## 506 Levels: C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 C1009 C101 ... PD9
```

1.2.5 Missing SPS

Score to SPS is missing for some participants:

```
(sps.miss <- Code.participant[is.na(SPS.total)])
```

```
## [1] C176 C704 C744 C202 C919 C1317
```

```
## 506 Levels: C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 C1009 C101 ... PD9
```

1.2.6 Wrong scores to CTQ

Some scores to the CTQ scale are erroneous:

```
(ctq.low <- Code.participant[(CTQ.sexual.abuse<5) | (CTQ.physical.abuse<5) |  
                             (CTQ.emotional.abuse<5) | (CTQ.physical.neglect<5) |  
                             (CTQ.emotional.neglect<5)])
```

```
## [1] PD31 PD32
```

```
## 506 Levels: C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 C1009 C101 ... PD9
```

1.3 Subset the database

Participants with wrong data are:

```
detach(data)
(participants.drop <- c(twice, sex.miss, age.miss, age.old, sps.miss, ctq.low,
parental.miss, parental.error))

## [1] C216 C364 C1306 C1334 C1911 C744 C1133 C176 C704 C744 C202 C919
## [13] C1317 PD31 PD32 C760 C527 C135 C1220 C1141 C1321 C1132 C1134 C1242
## [25] C562
## 506 Levels: C216 C364 C1000 C1001 C1003 C1004 C1005 C1006 C1007 C1008 ... PD9

Incomplete data are deleted from database.

trauma <- subset(data, !(Code.participant %in% participants.drop))
nb.subj <- nrow(trauma)
```

The number of subjects is now nb.subj: 482.

Some verifications and cleaning:

```
table(trauma$Sexe)

##
##      1  2
##    0 161 321

trauma$Sexe <- droplevels(trauma$Sexe)
table(trauma$Sexe)

##
##      1  2
##    161 321

summary(trauma$Age)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      14.0   17.0   17.0   17.7   18.0   25.0

table(trauma$Situation.parentale)

##
##      10 11 12 13 15
##      0 82 215 164 21  0

trauma$Situation.parentale <- droplevels(trauma$Situation.parentale)
names(trauma)[4:5] <- c('Parental', 'Education')
levels(trauma$Parental) <- c('Partnered', 'Married', 'Separated', 'Widowed')
levels(trauma$Education) <- c('Middle', 'General', 'Technical', 'Vocational',
'High')
```

2 Groups definition

2.1 Labeling data

The Childhood Trauma Questionnaire (CTQ) is used for labeling individuals according to several categories (categ variable) that are combined into groups (group variable) used in this study.

2.1.1 Categories definitions

Categories are defined using cut-off scores:

- CTQ.emotional.neglect ≥ 15 for severe emotional neglect
- CTQ.emotional.abuse ≥ 10 for severe emotional abuse
- CTQ.physical.neglect ≥ 8 for severe physical neglect
- CTQ.physical.abuse ≥ 8 for severe physical abuse
- CTQ.sexual.abuse ≥ 8 for severe sexual abuse

Score under the thresholds are considered as “low”.

The dichotomy between low and severe scores was used to get an integer code (between 0 and $2^5 - 1 = 31$) to define categories which were then grouped in:

- Low maltreatment (Low)
- Abuse (Abu)
- Neglected (Neg)
- Combination (Combi)

according to the following code (the complete explicit table of correspondences is given in Table S1):

```
ctq <- trauma[, c("CTQ.emotional.neglect", # >= 15
                 "CTQ.emotional.abuse", # >=10
                 "CTQ.physical.neglect", # >=8
                 "CTQ.physical.abuse", # >=8
                 "CTQ.sexual.abuse")] # >=8

# binary encoding of categories
ctq$categ <- (ctq[,1]>=15) * 2^0 + (ctq[,2]>=10) * 2^1 +
             (ctq[,3]>=8) * 2^2 + (ctq[,4]>=8) * 2^3 +
             (ctq[,5]>=8) * 2^4

# groups definition
ctq$group <- rep('Combi', length(ctq$categ))
ctq$group[ctq$categ==0] <- 'Low'
ctq$group[ctq$categ==1 | ctq$categ==4 | ctq$categ==5] <- 'Neg'
ctq$group[ctq$categ==2 | ctq$categ==8 | ctq$categ==10 | ctq$categ==16 |
          ctq$categ==18 | ctq$categ==24 | ctq$categ==26] <- 'Abu'

scores <- ctq
head(scores)
```

```
##   CTQ.emotional.neglect CTQ.emotional.abuse CTQ.physical.neglect
## 1                      5                      5                      5
## 2                      21                     5                      6
## 3                      6                      6                      8
## 4                      15                     17                     9
## 5                      8                      12                     13
## 6                      11                     5                      5
##   CTQ.physical.abuse CTQ.sexual.abuse categ group
## 1                      5                      5      0   Low
## 2                      5                      5      1   Neg
## 3                      5                      5      4   Neg
## 4                      6                      18     23 Combi
## 5                      14                     5     14 Combi
## 6                      5                      5      0   Low
```

Definition of factors:

```
trauma$categ <- factor(ctq$categ)
trauma$group <- factor(ctq$group, levels=c('Low', 'Neg', 'Abu', 'Combi'))
trauma$Sexe <- factor(trauma$Sexe, labels=c('M', 'F'))
```

Table S1: Definitions of the 32 categories and the four groups of subjects, together with the number of participants in each category in the initial database (i.e., before pairing)

Neglect		Abuse			Categories	Males	Females	Groups
Emotional	Physical	Emotional	Physical	Sexual				
low	low	low	low	low	0	75	133	Low maltreatment
severe	low	low	low	low	1	9	10	Neglect
low	severe	low	low	low	4	10	12	
severe	severe	low	low	low	5	5	9	
					Total	24	31	
low	low	severe	low	low	2	2	7	Abuse
low	low	low	severe	low	8	6	2	
low	low	severe	severe	low	10	2	4	
low	low	low	low	severe	16	1	17	
low	low	severe	low	severe	18	0	3	
low	low	low	severe	severe	24	1	2	
low	low	severe	severe	severe	26	0	4	
					Total	12	39	
severe	low	severe	low	low	3	4	8	Combination
low	severe	severe	low	low	6	5	6	
severe	severe	severe	low	low	7	1	11	
severe	low	low	severe	low	9	1	0	
severe	low	severe	severe	low	11	2	1	
low	severe	low	severe	low	12	1	1	
severe	severe	low	severe	low	13	0	1	
low	severe	severe	severe	low	14	3	2	
severe	severe	severe	severe	low	15	8	6	
severe	low	low	low	severe	17	2	3	
severe	low	severe	low	severe	19	1	1	
low	severe	low	low	severe	20	1	2	
severe	severe	low	low	severe	21	0	1	
low	severe	severe	low	severe	22	0	5	
severe	severe	severe	low	severe	23	0	2	
severe	low	low	severe	severe	25	0	1	
severe	low	severe	severe	severe	27	1	3	
low	severe	low	severe	severe	28	1	1	
severe	severe	low	severe	severe	29	0	0	
low	severe	severe	severe	severe	30	1	6	
severe	severe	severe	severe	severe	31	4	6	
					Total	36	67	

2.1.2 Comparison with latent profile analysis

```
library(mclust)
```

```
## Package 'mclust' version 6.1.2  
## Type 'citation("mclust")' for citing this R package in publications.
```

```
vars <- ctq[,1:5]  
names(vars)
```

```
## [1] "CTQ.emotional.neglect" "CTQ.emotional.abuse" "CTQ.physical.neglect"  
## [4] "CTQ.physical.abuse" "CTQ.sexual.abuse"
```

```
# scaled.vars <- scale(vars)  
scaled.vars <- vars # same result with or without scaling
```

Mclust provides the model with the lowest BIC in several models with different numbers of classes and different covariance models.

```
mclust.fit <- Mclust(scaled.vars, G=2:20) # tests for 2 to 20 classes  
summary(mclust.fit)
```

```
## -----  
## Gaussian finite mixture model fitted by EM algorithm  
## -----  
##  
## Mclust VEV (ellipsoidal, equal shape) model with 10 components:  
##  
## log-likelihood n df BIC ICL  
## -2855.657 482 173 -6780.098 -6789.539  
##  
## Clustering table:  
## 1 2 3 4 5 6 7 8 9 10  
## 69 41 38 46 73 21 42 52 38 62
```

```
ctq$mclass <- mclust.fit$classification  
with(ctq, table(mclass, group))
```

```
##      group  
## mclass Abu Combi Low Neg  
## 1 0 0 69 0  
## 2 21 10 9 1  
## 3 10 25 0 3  
## 4 6 39 0 1  
## 5 17 29 20 7  
## 6 0 4 9 8  
## 7 0 0 35 7  
## 8 4 16 2 30  
## 9 5 2 30 1  
## 10 0 0 60 2
```

```
nindiv <- nrow(ctq)  
print(nindiv)
```

```
## [1] 482
```

```
bxp.data <- data.frame(Group = factor(rep(ctq$group, 5)),  
                        Class = factor(rep(ctq$mclass, 5)),
```

```

Scale = factor(c(rep('NE', nindiv),
                 rep('NP', nindiv),
                 rep('AE', nindiv),
                 rep('AP', nindiv),
                 rep('AS', nindiv))),
Scores = c(ctq$CTQ.emotional.neglect,
           ctq$CTQ.physical.neglect,
           ctq$CTQ.emotional.abuse,
           ctq$CTQ.physical.abuse,
           ctq$CTQ.sexual.abuse))
summary(bxp.data)

```

```

##      Group      Class  Scale      Scores
## Abu   : 315    5      :365  AE:482  Min.   : 5.000
## Combi: 625    1      :345  AP:482  1st Qu.: 5.000
## Low   :1170   10     :310  AS:482  Median : 6.000
## Neg   : 300    8      :260  NE:482  Mean   : 7.932
##              4      :230  NP:482  3rd Qu.: 9.000
##              7      :210              Max.   :25.000
##              (Other):690

```

```

par(mfrow=c(5,2), mar=c(2,2,1,0)+0.1)
for (id.class in levels(bxp.data$Class)){
  boxplot(Scores~Scale,
          data=subset(bxp.data, Class==id.class),
          main=paste("Class", id.class, sep=' '),
          ylim=c(0,25))
}

```

```

par(mfrow=c(2,2), mar=c(2,2,1,0)+0.1)
for (id.group in levels(bxp.data$Group)){
  boxplot(Scores~Scale,
          data=subset(bxp.data, Group==id.group),
          main=id.group,
          ylim=c(0,25),
          pars=list(cex.lab=0.7, cex.sub=0.7, cex.main=0.9, cex.axis=0.8))
}

```

2.2 Matching

Since data are coming from convenience sampling, we do not expect clear representativeness or comparability. We thus improve the situation by matching participants without trauma (denoted “Control Units”) with participants with trauma (denoted “Treated Units”) in order to improve the comparability between groups.

2.2.1 Cardinality matching

Cardinality matching selects largest subgroups of “Treated Units” and “Control Units” which are best paired on a set of characteristics, here: age, sex, parental situation and education level.

```

library(MatchIt)
# The trauma group is coded 1 ("treated")
trauma$binary <- as.numeric(trauma$group != 'Low')
# Fit the cardinality matching model
mi.model <- matchit(binary ~ Sexe + Age + Parental + Education,
                    data=trauma, method="cardinality")

```

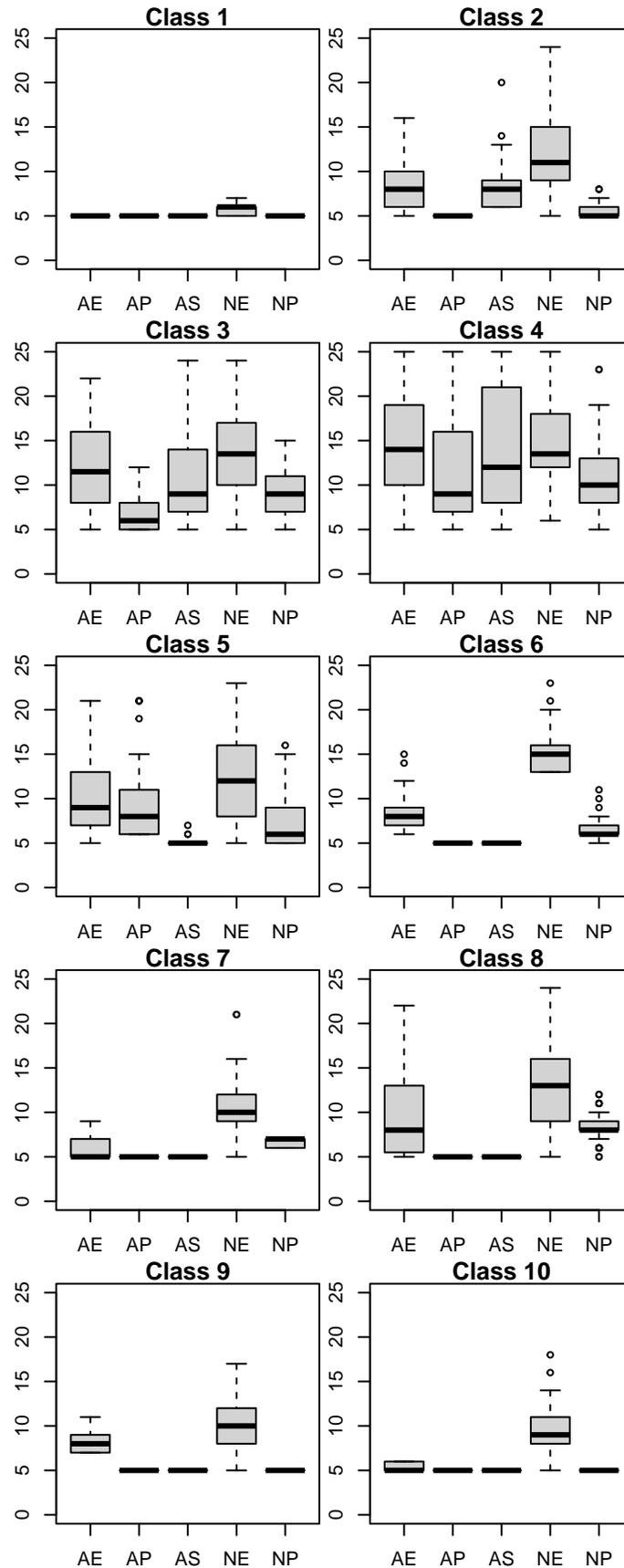


Figure S1: Scores to the CTQ sub-scales for the classes defined by latent profile analysis.

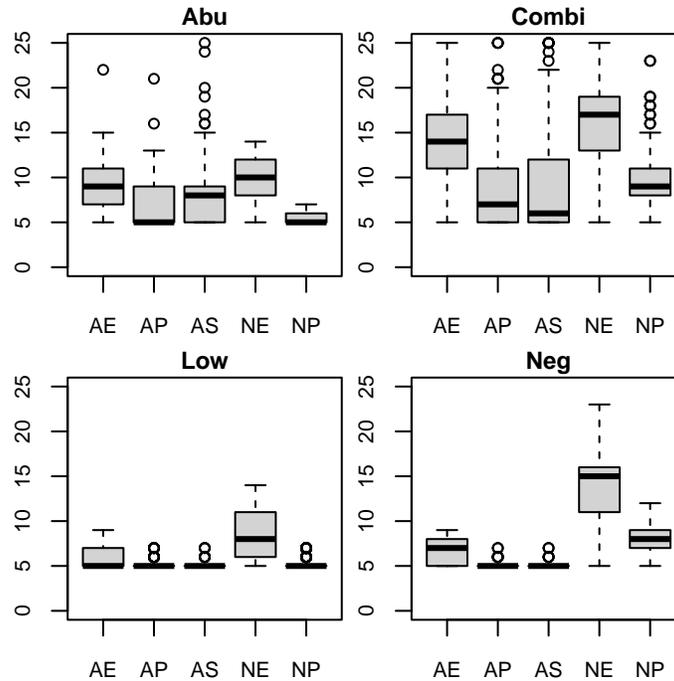


Figure S2: Scores to the CTQ sub-scales for the classes defined by categorical groups.

```
summary(mi.model)
```

```
##
## Call:
## matchit(formula = binary ~ Sexe + Age + Parental + Education,
## data = trauma, method = "cardinality")
##
## Summary of Balance for All Data:
##           Means Treated Means Control Std. Mean Diff. Var. Ratio
## SexeM           0.3185         0.3504        -0.0684          .
## SexeF           0.6815         0.6496         0.0684          .
## Age            17.9234        17.4701         0.2588        1.3841
## ParentalPartnered 0.1532         0.1880        -0.0966          .
## ParentalMarried   0.3790         0.5171        -0.2846          .
## ParentalSeparated 0.4153         0.2607         0.3138          .
## ParentalWidowed   0.0524         0.0342         0.0818          .
## EducationMiddle   0.0282         0.0043         0.1446          .
## EducationGeneral  0.7097         0.7778        -0.1500          .
## EducationTechnical 0.0282         0.0513        -0.1392          .
## EducationVocational 0.0202         0.0043         0.1130          .
## EducationHigh     0.2137         0.1624         0.1252          .
##           eCDF Mean eCDF Max
## SexeM           0.0319    0.0319
## SexeF           0.0319    0.0319
## Age            0.0384    0.1034
## ParentalPartnered 0.0348    0.0348
## ParentalMarried  0.1381    0.1381
## ParentalSeparated 0.1546    0.1546
## ParentalWidowed  0.0182    0.0182
```

```

## EducationMiddle      0.0240  0.0240
## EducationGeneral     0.0681  0.0681
## EducationTechnical   0.0231  0.0231
## EducationVocational  0.0159  0.0159
## EducationHigh       0.0513  0.0513
##
## Summary of Balance for Matched Data:
##           Means Treated Means Control Std. Mean Diff. Var. Ratio
## SexeM           0.3125      0.3317      -0.0413      .
## SexeF           0.6875      0.6683       0.0413      .
## Age            17.6490     17.5625       0.0494     1.1539
## ParentalPartnered 0.1827      0.1971      -0.0400      .
## ParentalMarried   0.4519      0.4712      -0.0396      .
## ParentalSeparated 0.3173      0.2933       0.0488      .
## ParentalWidowed   0.0481      0.0385       0.0431      .
## EducationMiddle   0.0096      0.0048       0.0290      .
## EducationGeneral   0.7885      0.8077      -0.0424      .
## EducationTechnical 0.0337      0.0385      -0.0290      .
## EducationVocational 0.0096      0.0048       0.0342      .
## EducationHigh     0.1587      0.1442       0.0352      .
##           eCDF Mean eCDF Max
## SexeM           0.0192  0.0192
## SexeF           0.0192  0.0192
## Age            0.0104  0.0433
## ParentalPartnered 0.0144  0.0144
## ParentalMarried 0.0192  0.0192
## ParentalSeparated 0.0240  0.0240
## ParentalWidowed 0.0096  0.0096
## EducationMiddle 0.0048  0.0048
## EducationGeneral 0.0192  0.0192
## EducationTechnical 0.0048  0.0048
## EducationVocational 0.0048  0.0048
## EducationHigh   0.0144  0.0144
##
## Sample Sizes:
##           Control Treated
## All           234     248
## Matched       208     208
## Unmatched     26      40
## Discarded     0       0

```

```
plot(summary(mi.model), xlim=c(0, 0.35))
```

```
# Apply the cardinality matching model to the data to select a subset of subjects
m.data <- match.data(mi.model)
```

```
not.matched <- !(trauma$Code.participant %in% m.data$Code.participant)
```

```
table(trauma$group)
```

```
##
##   Low   Neg   Abu   Combi
##   234   60   63   125
```

```
table(m.data$group)
```

```
##
```

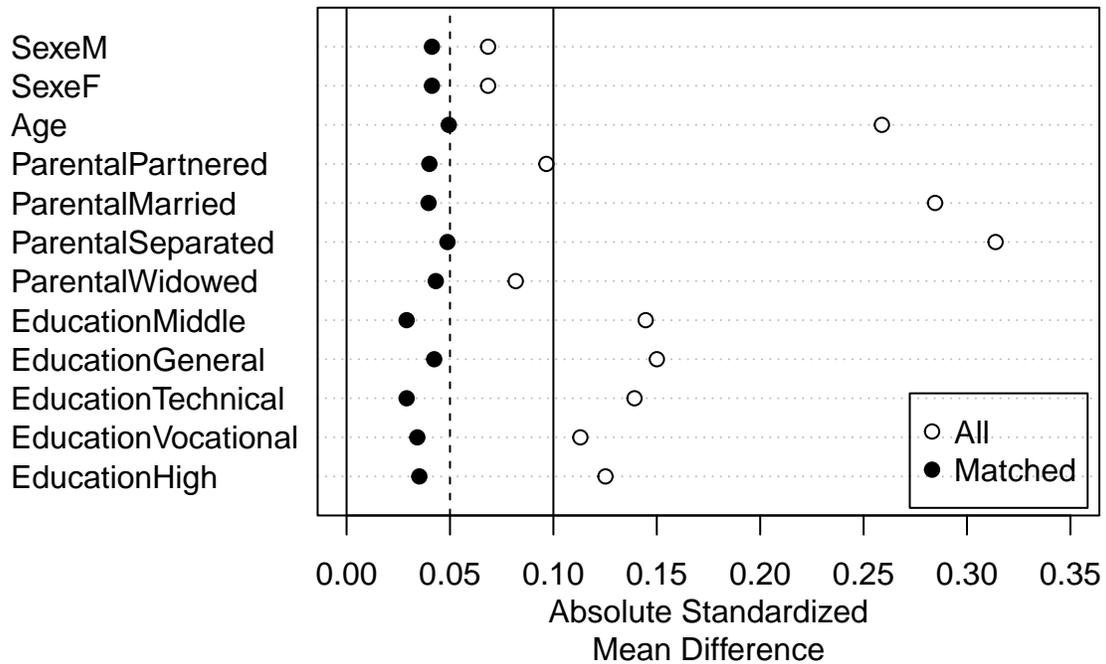


Figure S3: Absolute standardized mean difference for socio-demographic covariates before and after matching.

```
## Low Neg Abu Combi
## 208 51 53 104
```

```
table(m.data$group) / sum(table(m.data$group))
```

```
##
## Low Neg Abu Combi
## 0.5000000 0.1225962 0.1274038 0.2500000
```

```
table(m.data$group, m.data$Sex)
```

```
##
## M F
## Low 69 139
## Neg 21 30
## Abu 11 42
## Combi 33 71
```

2.2.2 Optimal pair matching

Optimal pair matching find the best pairs “treated”-“non treated” in the data.

```
# Fit the optimal pair matching
opt.model <- matchit(binary ~ Sexe + Age + Parental + Education,
                    data=m.data, method="optimal")
plot(opt.model, type = "jitter", interactive = FALSE)
```

```
# Apply the optimal pair matching model to the selected data.
opt.data <- match.data(opt.model, weights="opt.weights")
opt.data$old.group <- opt.data$group
# Create the new levels for the group factor
levels(opt.data$group) <- c('Low', 'Neg', 'Abu', 'Combi', 'LowA', 'LowN', 'LowC')
```

Distribution of Propensity Scores

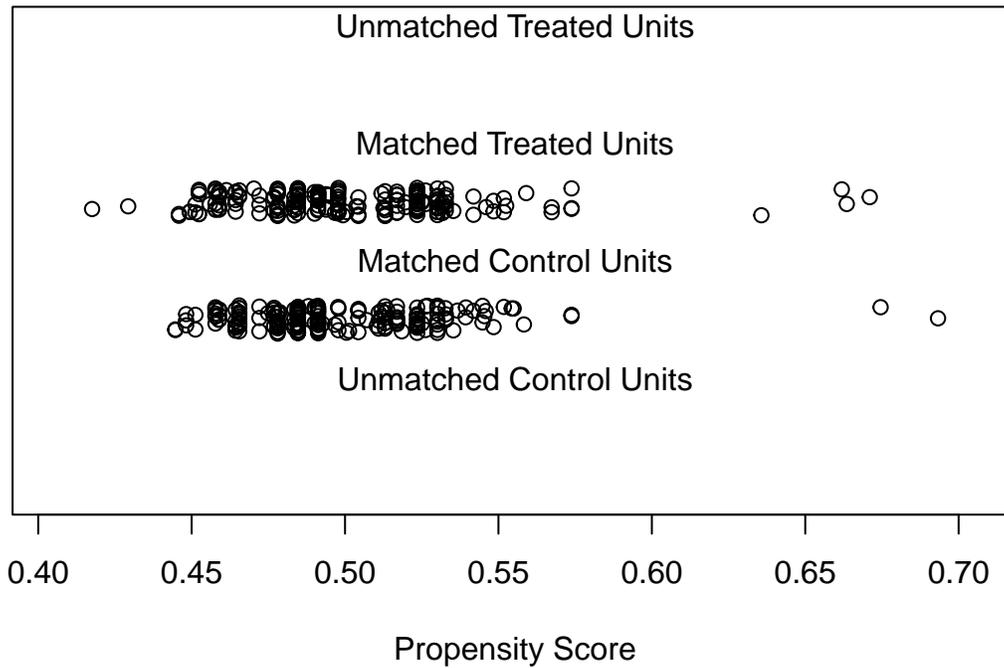


Figure S4: Distribution of the propensity score after optimal matching.

```
# The variable subclass defines the pairs of subjects with one in the trauma  
# group and the other in the no trauma group  
# Get the values for each trauma group  
subc.abu <- opt.data$subclass[opt.data$old.group=='Abu']  
subc.neg <- opt.data$subclass[opt.data$old.group=='Neg']  
subc.multi <- opt.data$subclass[opt.data$old.group=='Combi']  
# Associate the subjects in the "no trauma" group to their corresponding type of  
# trauma group  
opt.data$group[opt.data$old.group=="Low" &  
               opt.data$subclass %in% subc.abu] <- "LowA"  
opt.data$group[opt.data$old.group=="Low" &  
               opt.data$subclass %in% subc.neg] <- 'LowN'  
opt.data$group[opt.data$old.group=="Low" &  
               opt.data$subclass %in% subc.multi] <- 'LowC'  
# Delete the "Low" level for the group factor which is no more needed  
opt.data$group <- droplevels(opt.data$group)
```

The groups are now denoted:

- Low maltreatment - Abuse: LowA
- Low maltreatment - Neglect: LowN
- Low maltreatment - Combination: LowC
- Trauma - Abuse: Abu
- Trauma - Neglect: Neg
- Trauma - Combination: Combi

and the sample size of each group is:

```
table(opt.data$group)
```

```
##  
##   Neg   Abu Combi  LowA  LowN  LowC  
##   51   53  104   53   51  104
```

```
nindiv <- nrow(opt.data)  
# print(nindiv)  
# names(opt.data)  
bxp.data <- data.frame(Group = factor(rep(opt.data$group, 5)),  
                       Scale = factor(c(rep('NE', nindiv),  
                                         rep('NP', nindiv),  
                                         rep('AE', nindiv),  
                                         rep('AP', nindiv),  
                                         rep('AS', nindiv))),  
                       Scores = c(opt.data$CTQ.emotional.neglect,  
                                   opt.data$CTQ.physical.neglect,  
                                   opt.data$CTQ.emotional.abuse,  
                                   opt.data$CTQ.physical.abuse,  
                                   opt.data$CTQ.sexual.abuse))  
  
summary(bxp.data)
```

```
##   Group   Scale   Scores  
## Neg  :255  AE:416  Min.   : 5.000  
## Abu  :265  AP:416  1st Qu.: 5.000  
## Combi:520  AS:416  Median : 6.000  
## LowA :265  NE:416  Mean    : 7.849  
## LowN :255  NP:416  3rd Qu.: 9.000  
## LowC :520           Max.    :25.000
```

```
par(mfrow=c(2,3), mar=c(2,2,1,0)+0.1)
```

```
# for (id.group in levels(bxp.data$Group)){  
for (id.group in c('Neg', 'Abu', 'Combi', 'LowN', 'LowA', 'LowC')){  
boxplot(Scores~Scale,  
        data=subset(bxp.data, Group==id.group),  
        main=id.group,  
        ylim=c(0,25))  
}  
}
```

3 Data analysis

```
save(opt.data, file="output/opt.data.Robj")
```

Numerical descriptive statistics are given in the article.

Definition of functions:

```
library(FactoMineR)  
library(factoextra)  
library(emmeans)  
library(effectsize)  
library(glue)  
  
check.pca <- function(data, cutoff.thresh=0.999){
```

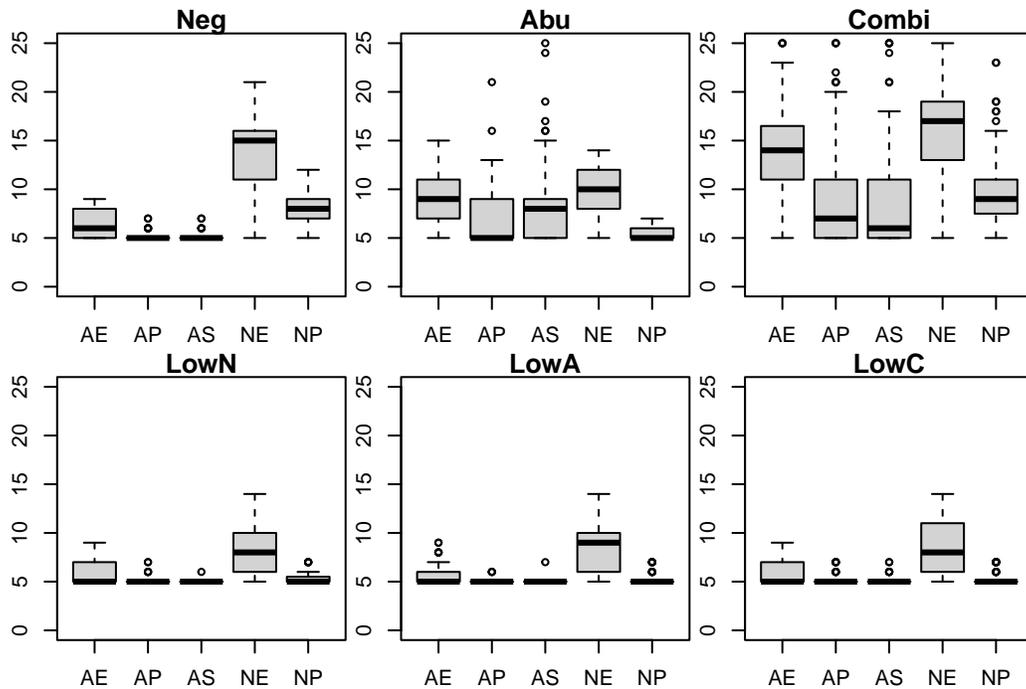


Figure S5: Distribution of scores to the CTQ for the groups Neglected, Abuse and Combination and their associated Low Maltreatment groups

```

X.scaled <- scale(data)
Sigma <- cov(X.scaled)
d2.maha <- mahalanobis(X.scaled, colMeans(X.scaled), Sigma)
cutoff <- qchisq(cutoff.thresh, df=ncol(X.scaled))
outliers.nb <- sum(d2.maha > cutoff)
outliers.idx <- if(outliers.nb!=0) which(d2.maha > cutoff) else "None"
print(glue("Nb outliers: {outliers.nb} -- {outliers.idx}"))
return(outliers.idx)
}

show.pca <- function(data, qvs, qnvs, outliers){
res.PCA <- PCA(data, quali.sup=qvs, quanti.sup=qnvs, graph=FALSE, ncp=30)
if(all(outliers!="None")){
print('----- Outliers influence -----')
var.pc <- res.PCA$eig[, 2] / 100
for (i in outliers){
pc.tot <- round(sum(res.PCA$ind$contrib[i,] * var.pc), 3)
print(glue("Contribution for outlier {i} to total variance: {pc.tot} %"))
}
}
print(' ')
}
print('----- Eigen values -----')
print(res.PCA$eig)
print(' ')
print('----- Dimension description -----')
# dimdesc: This function is designed to point out the variables and the
# categories that are the most characteristic according to each

```

```

# dimension obtained by a Factor Analysis. (From the documentation)
print(dimdesc(res.PCA, axes=1:2, proba=1))
aov.data <- data.frame(d1=res.PCA$ind$coord[,1],
                      d2=res.PCA$ind$coord[,2],
                      group=data[,qvs])

return(aov.data)
}

my.mcomp <- function(data, form){
  mod <- lm(as.formula(form), data=data)

  par(mfrow=c(1,3), mar=c(4,4,0.1,1.5))
  plot(mod, which=c(1,2,6))

  print(anova(mod))
  print(omega_squared(mod))
  print(' ')
  print(paste('----- Estimated marginal means for', form, '-----'))
  emm_groupe <- emmeans(mod, ~ group)
  print(summary(emm_groupe))

  contrasts_list <- list(
    'Neglect - Control'=c(1,0,0,0,-1,0),
    'Neglect - Abuse'=c(1,-1,0,0,0,0),
    'Neglect - Combination'=c(1,0,-1,0,0,0),
    'Abuse - Control'=c(0,1,0,-1,0,0),
    'Abuse - Combination'=c(0,1,-1,0,0,0),
    'Combination - Control'=c(0,0,1,0,0,-1))

  all_contrasts <- contrast(emm_groupe, contrasts_list)
  print(' ')
  print(paste('----- Contrasts ', form, '-----'))
  print(summary(all_contrasts))
  print(eff_size(all_contrasts,
                sigma = sigma(mod),
                edf = df.residual(mod),
                method='identity'))

  return(mod)
}

library(car)

my.lmcomp <- function(data, form){
  mod <- lm(as.formula(form), data=data)

  par(mfrow=c(1,3), mar=c(4,4,0.1,1.5))
  plot(mod, which=c(1,2,6))
  print(summary(mod))
  print(Anova(mod, type=2))
  print(omega_squared(Anova(mod, type=2)))
  return(mod)
}

```

```

mod.comp <- function(mod1, mod2){
  name1 <- deparse(substitute(mod1))
  name2 <- deparse(substitute(mod2))
  radj <- c(summary(mod1)$adj.r.squared,
            summary(mod2)$adj.r.squared)
  radj <- c(radj, radj[1]-radj[2])
  aic <- c(AIC(mod1), AIC(mod2))
  aic <- c(aic, aic[1]-aic[2])
  bic <- c(BIC(mod1), BIC(mod2))
  bic <- c(bic, bic[1]-bic[2])

  dd <- data.frame(model = c(name1, name2, 'Delta'),
                  R2adj = round(radj, 3),
                  AIC = round(aic, 3),
                  BIC = round(bic, 3))

  print(dd)
}

```

The five types of child trauma evaluated by the CTQ scale are renamed according to:

```

names(opt.data)[6:10] <- c('ctq.en', # emotional neglect
                          'ctq.pa', # physical abuse
                          'ctq.ea', # emotional abuse
                          'ctq.pn', # physical neglect
                          'ctq.sa' # sexual abuse
                          )

```

3.1 DERS

3.1.1 PCA description

```
ders <- opt.data[,c(20:25, 41, 6:10)]
pca.out <- check.pca(ders[,1:6])

## Nb outliers: 1 -- 224

aov.ders <- show.pca(ders,qvs=c(7), qnvs=c(8:12), outliers=pca.out)

## [1] "----- Outliers influence -----"
## Contribution for outlier 224 to total variance: 0.649 %
## [1] " "
## [1] "----- Eigen values -----"
##          eigenvalue percentage of variance cumulative percentage of variance
## comp 1  2.9852230          49.753716          49.75372
## comp 2  1.2136385          20.227308          69.98102
## comp 3  0.6853257          11.422095          81.40312
## comp 4  0.4451239           7.418732          88.82185
## comp 5  0.4117515           6.862525          95.68438
## comp 6  0.2589374           4.315624         100.00000
## [1] " "
## [1] "----- Dimension description -----"
## $Dim.1
##
## Link between the variable and the continuous variables (R-square)
## =====
##          correlation      p.value
## DERSstrategies    0.8850106 1.556210e-139
## DERSimpulse       0.8229383 1.044451e-103
## DERSnonacceptance 0.7389241 5.401735e-73
## DERSgoals         0.7328327 3.119255e-71
## DERSclarity       0.6527146 7.347411e-52
## ctq.ea            0.3036451 2.541023e-10
## ctq.en            0.2411606 6.438262e-07
## ctq.pn            0.1805985 2.131403e-04
## ctq.sa            0.1311125 7.412919e-03
## ctq.pa            0.1100027 2.485257e-02
## DERSawareness    -0.1251514 1.061973e-02
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##          R2      p.value
## group 0.08443648 8.273753e-07
##
## Link between variable and the categories of the categorical variables
## =====
##          Estimate      p.value
## group=Combi 0.7564038 1.192753e-06
## group=Abu   0.3524300 1.757802e-01
## group=Neg   0.1860496 5.544387e-01
## group=LowA -0.3193115 9.465136e-02
## group=LowC -0.3517619 5.883790e-03
## group=LowN -0.6238101 2.808877e-03
```

```

##
## $Dim.2
##
## Link between the variable and the continuous variables (R-square)
## =====
##           correlation      p.value
## DERSawareness  0.913891364 3.496421e-164
## DERSclarity    0.562282751 4.745983e-36
## ctq.en         0.215298380 9.419413e-06
## ctq.ea         0.194212605 6.683376e-05
## ctq.pn         0.191210896 8.691497e-05
## ctq.pa         0.111772710 2.260686e-02
## ctq.sa         0.102039896 3.749030e-02
## DERSnonacceptance 0.002143437 9.652342e-01
## DERSimpulse    -0.002924330 9.525813e-01
## DERSstrategies -0.092690020 5.890624e-02
## DERSgoals      -0.231677772 1.786063e-06
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##           R2      p.value
## group 0.03463201 0.01270863
##
## Link between variable and the categories of the categorical variables
## =====
##           Estimate      p.value
## group=Combi  0.32968174 0.0007646306
## group=Neg    0.10249492 0.5507831139
## group=LowN   0.02333811 0.9598845812
## group=Abu    -0.12620911 0.3153970591
## group=LowA   -0.13546159 0.2849131450
## group=LowC   -0.19384407 0.0248567139

```

3.1.2 First component analysis

```
ders.d1.group <- my.mcomp(aov.ders, 'd1~group')
```

3.1.2.1 Categorical model

```

## Analysis of Variance Table
##
## Response: d1
##           Df  Sum Sq Mean Sq F value    Pr(>F)
## group         5   104.86  20.9715   7.5623 8.274e-07 ***
## Residuals 410  1137.00   2.7732
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.
## # Effect Size for ANOVA
##
## Parameter | Omega2 |      95% CI
## -----

```

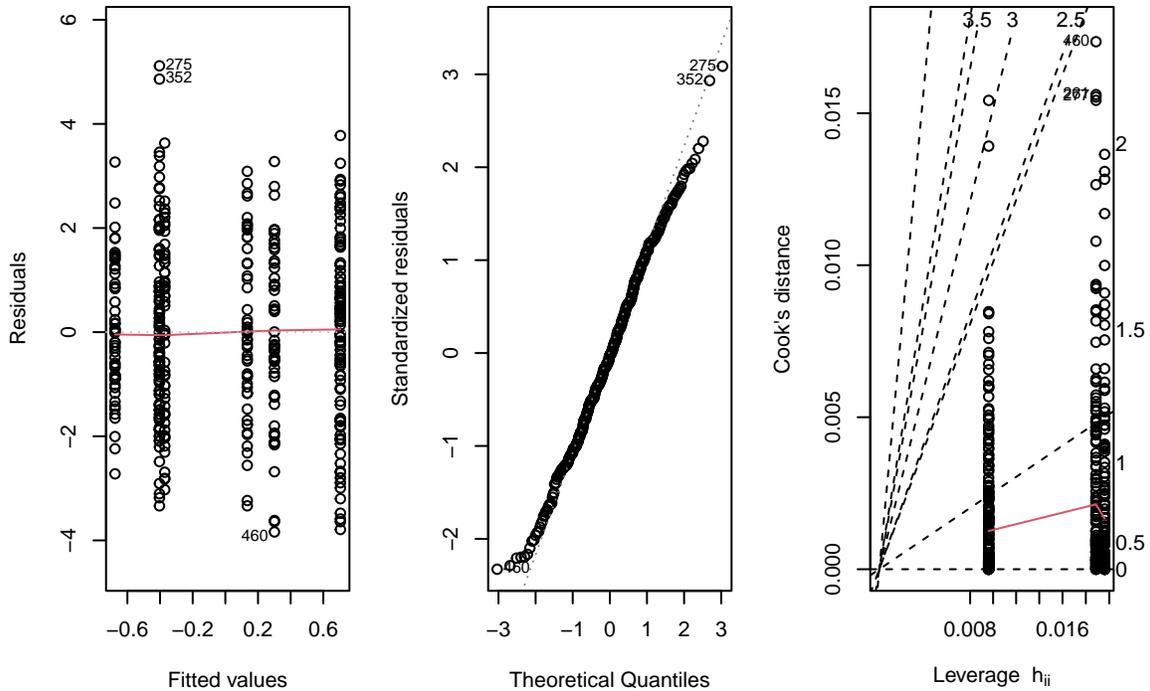


Figure S6: Diagnostic plots of the linear model $d1 \sim \text{group}$ for DERS data.

```
## group      | 0.07 | [0.03, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d1~group -----"
## group emmean SE df lower.CL upper.CL
## Neg  0.134 0.233 410 -0.324 0.5927
## Abu  0.301 0.229 410 -0.149 0.7504
## Combi 0.705 0.163 410 0.384 1.0257
## LowA -0.371 0.229 410 -0.821 0.0786
## LowN -0.676 0.233 410 -1.134 -0.2171
## LowC -0.403 0.163 410 -0.724 -0.0825
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts d1~group -----"
## contrast          estimate SE df t.ratio p.value
## Neglect - Control    0.810 0.330 410  2.456 0.0145
## Neglect - Abuse     -0.166 0.327 410 -0.509 0.6108
## Neglect - Combination -0.570 0.285 410 -2.004 0.0458
## Abuse - Control      0.672 0.323 410  2.077 0.0385
## Abuse - Combination -0.404 0.281 410 -1.437 0.1514
## Combination - Control 1.108 0.231 410  4.799 <.0001
##
## contrast          effect.size SE df lower.CL upper.CL
## (Neglect - Control)  0.4863 0.199 410  0.0956 0.87703
## (Neglect - Abuse)   -0.0999 0.196 410 -0.4856 0.28574
## (Neglect - Combination) -0.3425 0.171 410 -0.6794 -0.00563
## (Abuse - Control)    0.4034 0.195 410  0.0205 0.78625
## (Abuse - Combination) -0.2426 0.169 410 -0.5748 0.08959
```

```
## (Combination - Control)      0.6655 0.141 410   0.3890  0.94186
##
## sigma used for effect sizes: 1.665
## Confidence level used: 0.95
```

```
ders.mod <- cbind(aov.ders, opt.data[,6:10])
ders.d1.ctq <- my.lmcomp(ders.mod, 'd1~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

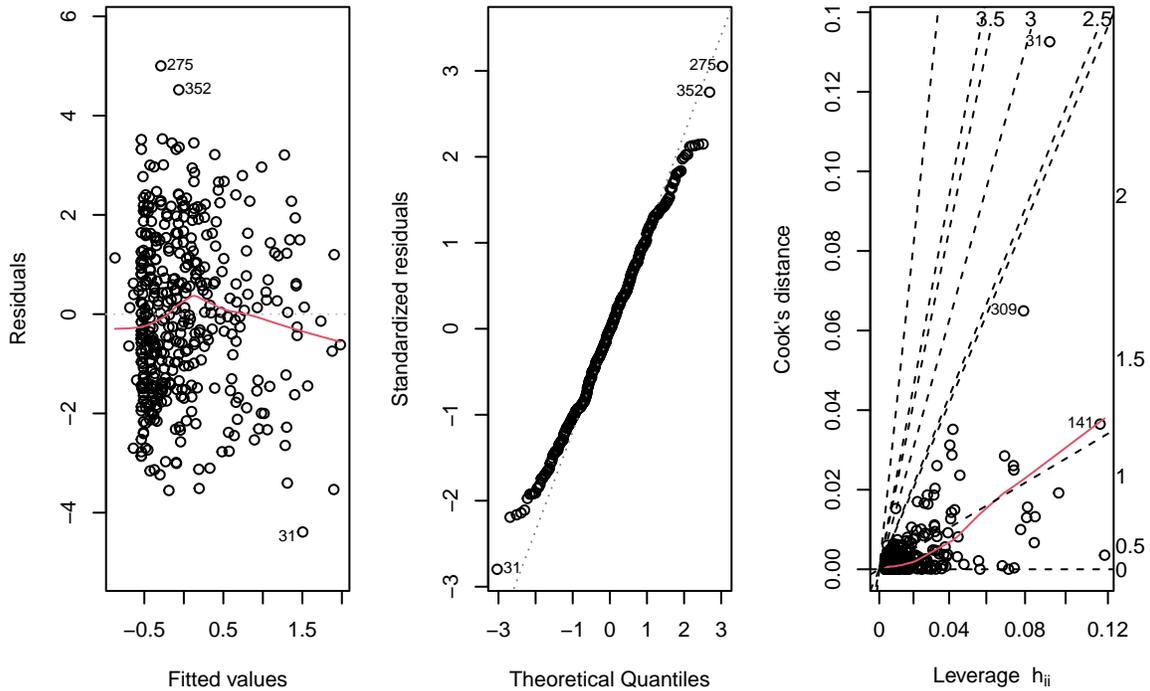


Figure S7: Diagnostic plots of the linear model $d1 \sim CTQ$ sub-scales for DERS data.

3.1.2.2 Dimensional model

```
##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.3890 -1.3354 -0.0173  1.1996  5.0027
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.070596   0.245997  -4.352 1.70e-05 ***
## ctq.en       0.027474   0.022803   1.205  0.2289
## ctq.pa      -0.059435   0.032449  -1.832  0.0677 .
## ctq.ea       0.128915   0.030057   4.289 2.24e-05 ***
## ctq.pn       0.004501   0.040881   0.110  0.9124
## ctq.sa       0.005179   0.025146   0.206  0.8369
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```

## Residual standard error: 1.647 on 410 degrees of freedom
## Multiple R-squared:  0.1046, Adjusted R-squared:  0.09371
## F-statistic: 9.582 on 5 and 410 DF,  p-value: 1.162e-08
##
## Anova Table (Type II tests)
##
## Response: d1
##           Sum Sq Df F value    Pr(>F)
## ctq.en      3.94  1  1.4517   0.22895
## ctq.pa      9.10  1  3.3550   0.06773 .
## ctq.ea     49.89  1 18.3954 2.241e-05 ***
## ctq.pn      0.03  1  0.0121   0.91239
## ctq.sa      0.12  1  0.0424   0.83692
## Residuals 1111.92 410
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----|-----|-----
## ctq.en    |          1.08e-03 | [0.00, 1.00]
## ctq.pa    |          5.63e-03 | [0.00, 1.00]
## ctq.ea    |           0.04    | [0.01, 1.00]
## ctq.pn    |           0.00    | [0.00, 1.00]
## ctq.sa    |           0.00    | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].

```

```

mod.comp(ders.d1.group, ders.d1.ctq)

```

```

##           model R2adj      AIC      BIC
## 1 ders.d1.group 0.073 1612.828 1641.043
## 2  ders.d1.ctq  0.094 1603.552 1631.767
## 3      Delta -0.020   9.276   9.276

```

3.1.3 Second component analysis

```

ders.d2.group <- my.mcomp(aov.ders, 'd2~group')

```

3.1.3.1 Categorical model

```

## Analysis of Variance Table
##
## Response: d2
##           Df Sum Sq Mean Sq F value    Pr(>F)
## group      5  17.48   3.4970  2.9417 0.01271 *
## Residuals 410 487.39   1.1888
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.
## # Effect Size for ANOVA
##
## Parameter | Omega2 |          95% CI

```

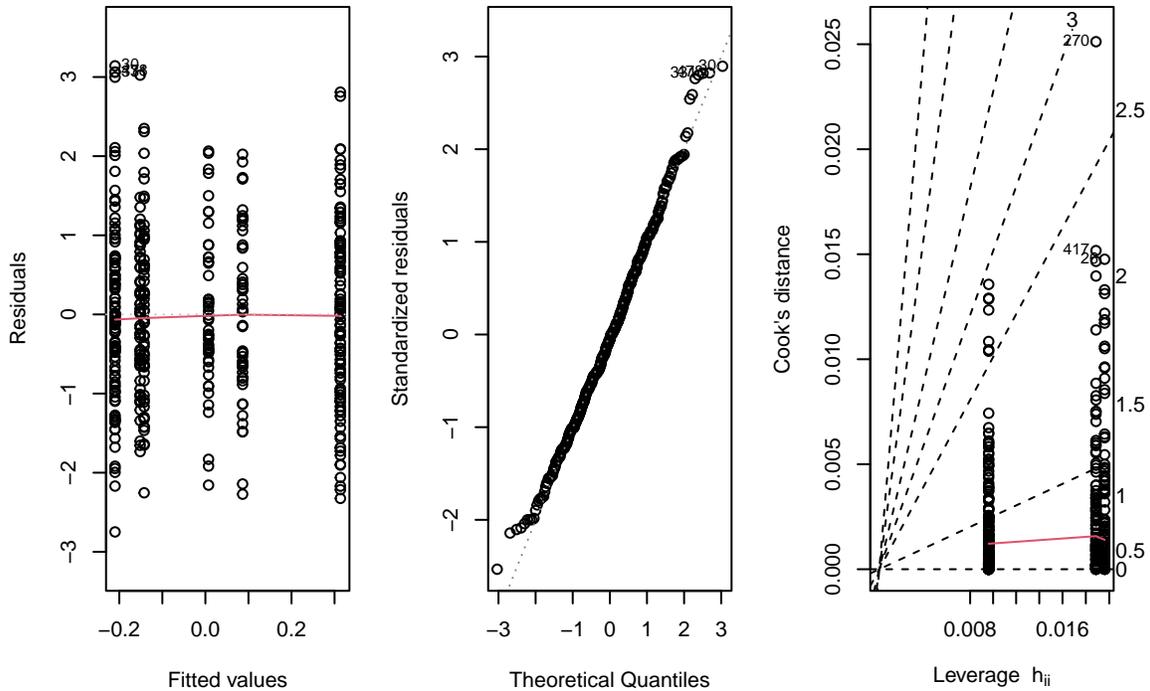


Figure S8: Diagnostic plots of the linear model $d2 \sim \text{group}$ for DERS data.

```
## -----
## group      |  0.02 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d2~group -----"
## group  emmean   SE  df lower.CL upper.CL
## Neg    0.08645  0.153 410  -0.214  0.386565
## Abu   -0.14226  0.150 410  -0.437  0.152144
## Combi  0.31363  0.107 410   0.103  0.523799
## LowA  -0.15151  0.150 410  -0.446  0.142892
## LowN   0.00729  0.153 410  -0.293  0.307408
## LowC  -0.20989  0.107 410  -0.420  0.000273
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts d2~group -----"
## contrast          estimate    SE  df t.ratio p.value
## Neglect - Control    0.07916  0.216 410   0.367  0.7141
## Neglect - Abuse      0.22870  0.214 410   1.069  0.2855
## Neglect - Combination -0.22719  0.186 410  -1.219  0.2236
## Abuse - Control      0.00925  0.212 410   0.044  0.9652
## Abuse - Combination -0.45589  0.184 410  -2.478  0.0136
## Combination - Control 0.52353  0.151 410   3.463  0.0006
##
## contrast          effect.size   SE  df lower.CL upper.CL
## (Neglect - Control)    0.07260  0.198 410  -0.317  0.4619
## (Neglect - Abuse)      0.20976  0.196 410  -0.176  0.5956
## (Neglect - Combination) -0.20837  0.171 410  -0.545  0.1280
## (Abuse - Control)      0.00849  0.194 410  -0.373  0.3904
```

```
## (Abuse - Combination)      -0.41813 0.169 410   -0.751  -0.0851
## (Combination - Control)   0.48017 0.140 410    0.206   0.7548
##
## sigma used for effect sizes: 1.09
## Confidence level used: 0.95
```

```
ders.d2.ctq <- my.lmcomp(ders.mod, 'd2~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

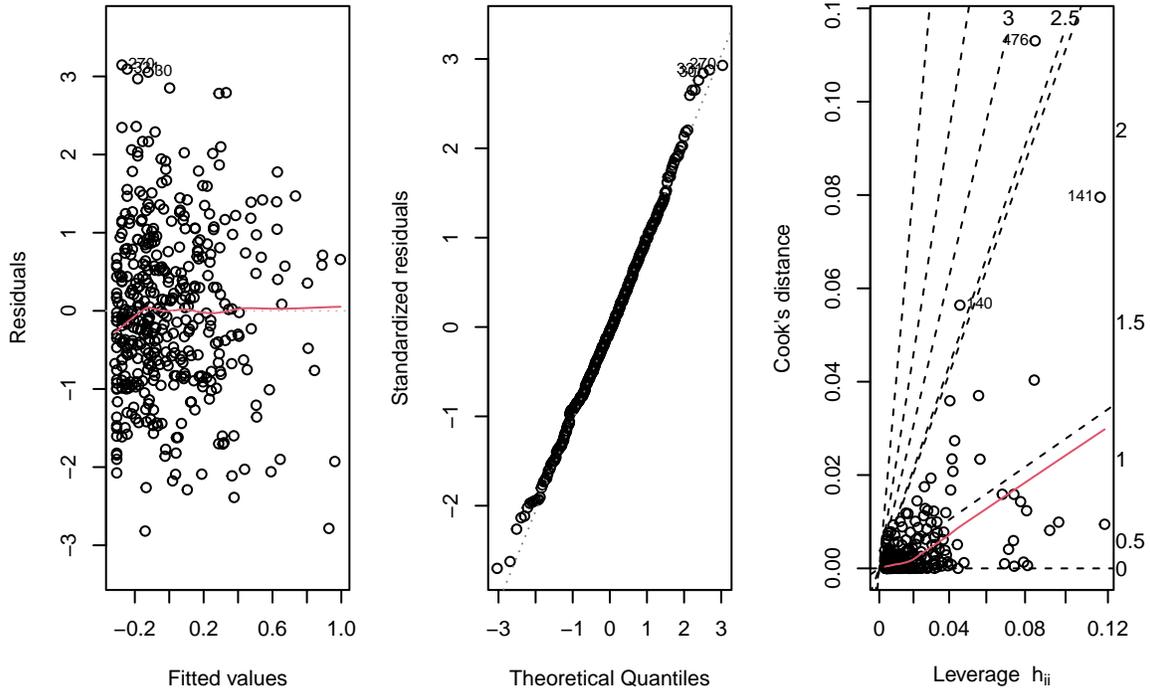


Figure S9: Diagnostic plots of the linear model $d2 \sim CTQ$ sub-scales for DERS data.

3.1.3.2 Dimensional model

```
##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.81856 -0.75512 -0.04205  0.71216  3.14663
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.694806   0.160964  -4.317 1.99e-05 ***
## ctq.en       0.030502   0.014921   2.044  0.0416 *
## ctq.pa      -0.009797   0.021232  -0.461  0.6447
## ctq.ea       0.018466   0.019667   0.939  0.3483
## ctq.pn       0.037609   0.026750   1.406  0.1605
## ctq.sa       0.001133   0.016454   0.069  0.9451
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```

## Residual standard error: 1.078 on 410 degrees of freedom
## Multiple R-squared:  0.05704,    Adjusted R-squared:  0.04554
## F-statistic:  4.96 on 5 and 410 DF,  p-value: 0.0002015
##
## Anova Table (Type II tests)
##
## Response: d2
##           Sum Sq Df F value  Pr(>F)
## ctq.en      4.85  1  4.1790 0.04157 *
## ctq.pa       0.25  1  0.2129 0.64473
## ctq.ea       1.02  1  0.8815 0.34834
## ctq.pn       2.30  1  1.9766 0.16050
## ctq.sa       0.01  1  0.0047 0.94512
## Residuals 476.07 410
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----|-----|-----
## ctq.en   |          7.58e-03 | [0.00, 1.00]
## ctq.pa   |          0.00     | [0.00, 1.00]
## ctq.ea   |          0.00     | [0.00, 1.00]
## ctq.pn   |          2.34e-03 | [0.00, 1.00]
## ctq.sa   |          0.00     | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].

```

```

mod.comp(ders.d2.group, ders.d2.ctq)

```

```

##           model R2adj      AIC      BIC
## 1 ders.d2.group  0.023 1260.442 1288.656
## 2  ders.d2.ctq   0.046 1250.671 1278.886
## 3           Delta -0.023   9.771   9.771

```

3.2 CERQ

3.2.1 PCA description

```
cerq <- opt.data[, c(11:19, 41, 6:10)]
pca.out <- check.pca(cerq[,1:9])

## Nb outliers: 2 -- 68
## Nb outliers: 2 -- 392

aov.cerq <- show.pca(cerq, qvs=c(10), qnvs=c(11:15), outliers=pca.out)

## [1] "----- Outliers influence -----"
## Contribution for outlier 68 to total variance: 0.779 %
## Contribution for outlier 392 to total variance: 0.669 %
## [1] " "
## [1] "----- Eigen values -----"
##          eigenvalue percentage of variance cumulative percentage of variance
## comp 1  2.3289202          25.876891          25.87689
## comp 2  1.8956248          21.062498          46.93939
## comp 3  1.1486817          12.763130          59.70252
## comp 4  0.8110942           9.012157          68.71468
## comp 5  0.7303331           8.114813          76.82949
## comp 6  0.6840165           7.600183          84.42967
## comp 7  0.5190504           5.767227          90.19690
## comp 8  0.4634024           5.148915          95.34582
## comp 9  0.4188766           4.654185         100.00000
## [1] " "
## [1] "----- Dimension description -----"
## $Dim.1
##
## Link between the variable and the continuous variables (R-square)
## =====
##          correlation      p.value
## CERQpositivereappraisal  0.78787559 3.421132e-89
## CERQputtingintoperspective 0.70269267 3.500404e-63
## CERQrefocusonplanning    0.69164687 1.759073e-60
## CERQpositivefocusing     0.65577527 1.716509e-52
## CERQacceptance           0.51149302 4.118721e-29
## ctq.sa                   -0.02700534 5.828367e-01
## CERQruminations          -0.04377713 3.731319e-01
## ctq.pa                   -0.04409354 3.696840e-01
## ctq.ea                   -0.09220340 6.025361e-02
## CERQblamingother        -0.09765726 4.652577e-02
## CERQselfblame           -0.12570076 1.027975e-02
## CERQcatastrophizing     -0.13076524 7.572735e-03
## ctq.pn                   -0.14898476 2.314598e-03
## ctq.en                   -0.19790067 4.813122e-05
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##          R2      p.value
## group 0.03870734 0.006180248
##
## Link between variable and the categories of the categorical variables
```

```

## =====
##           Estimate      p.value
## group=LowC  0.280052191 0.036002980
## group=Abu   0.353229108 0.078412409
## group=LowA  0.163231593 0.429756861
## group=LowN -0.008425785 0.933977471
## group=Combi -0.235129304 0.060657375
## group=Neg  -0.552957804 0.004979009
##
## $Dim.2
##
## Link between the variable and the continuous variables (R-square)
## =====
##           correlation      p.value
## CERQrumination      0.81231530 5.458564e-99
## CERQselfblame       0.71755388 5.087007e-67
## CERQcatastrophizing 0.62729433 6.924024e-47
## CERQblamingother    0.43841388 5.730055e-21
## CERQacceptance      0.31783219 3.225998e-11
## ctq.ea               0.29492301 8.562227e-10
## CERQrefocusonplanning 0.16084659 9.939650e-04
## ctq.en               0.15532574 1.483914e-03
## ctq.sa               0.15077051 2.045682e-03
## ctq.pn               0.14112096 3.925343e-03
## ctq.pa               0.10256385 3.651702e-02
## CERQpositivereappraisal 0.02297493 6.403209e-01
## CERQputtingintoperspective 0.02104638 6.686370e-01
## CERQpositivefocusing -0.08555983 8.132811e-02
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##           R2      p.value
## group 0.09128294 1.983808e-07
##
## Link between variable and the categories of the categorical variables
## =====
##           Estimate      p.value
## group=Combi  0.59700388 3.329540e-06
## group=Abu    0.49900345 1.264627e-02
## group=Neg    -0.09241639 4.035895e-01
## group=LowC  -0.14260069 8.534073e-02
## group=LowA  -0.30865835 3.759817e-02
## group=LowN  -0.55233190 6.799304e-04

```

3.2.2 First component analysis

```

cerq.mod <- cbind(aov.cerq, opt.data[,6:10])
cerq.d1.group <- my.mcomp(aov.cerq, 'd1~group')

```

3.2.2.1 Categorical model

```

## Analysis of Variance Table
##
## Response: d1

```

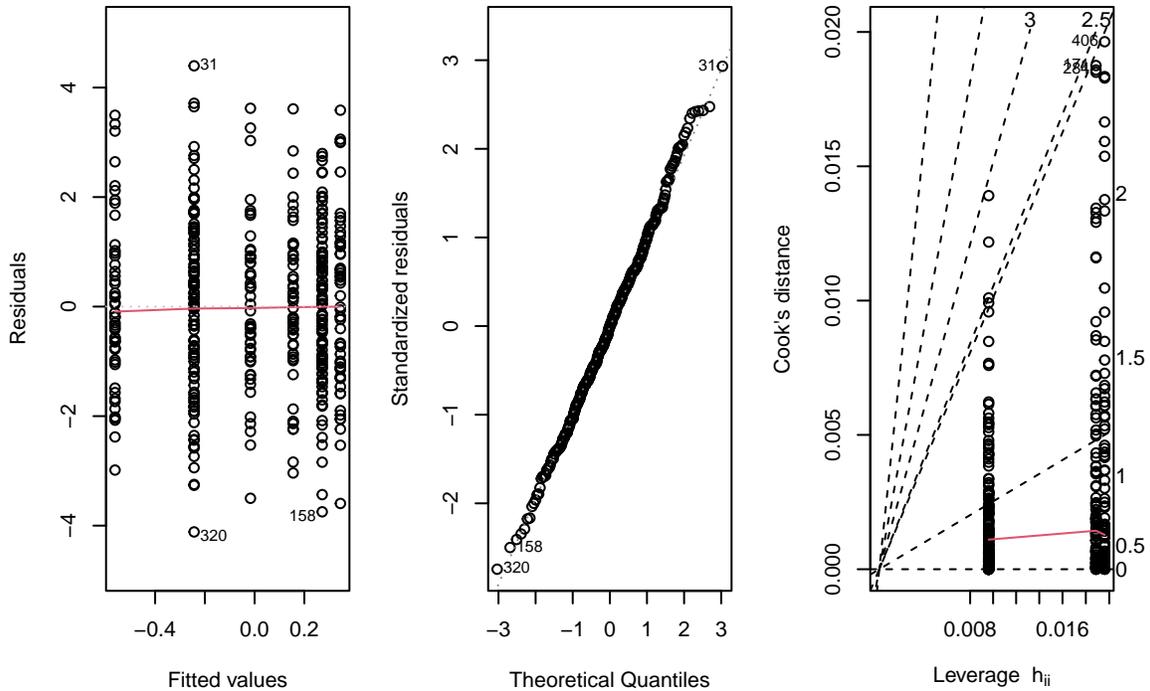


Figure S10: Diagnostic plots of the linear model $d1 \sim \text{group}$ for CERQ data.

```
##           Df Sum Sq Mean Sq F value Pr(>F)
## group      5  37.50  7.5002  3.3018 0.00618 **
## Residuals 410 931.33  2.2715
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.

## # Effect Size for ANOVA
##
## Parameter | Omega2 |      95% CI
## -----
## group     |  0.03 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d1~group -----"
## group emmean SE df lower.CL upper.CL
## Neg   -0.5612 0.211 410 -0.9760 -0.1463
## Abu    0.3450 0.207 410 -0.0619  0.7520
## Combi -0.2433 0.148 410 -0.5339  0.0472
## LowA   0.1550 0.207 410 -0.2519  0.5620
## LowN  -0.0166 0.211 410 -0.4315  0.3982
## LowC   0.2718 0.148 410 -0.0187  0.5624
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts d1~group -----"
## contrast          estimate SE df t.ratio p.value
## Neglect - Control    -0.545 0.298 410 -1.824  0.0688
```

```

## Neglect - Abuse          -0.906 0.296 410  -3.065  0.0023
## Neglect - Combination   -0.318 0.258 410  -1.234  0.2181
## Abuse - Control         0.190 0.293 410   0.649  0.5167
## Abuse - Combination     0.588 0.254 410   2.313  0.0212
## Combination - Control  -0.515 0.209 410  -2.465  0.0141
##
## contrast                effect.size   SE  df lower.CL upper.CL
## (Neglect - Control)     -0.361 0.198 410  -0.7514  0.0288
## (Neglect - Abuse)       -0.601 0.197 410  -0.9890 -0.2135
## (Neglect - Combination) -0.211 0.171 410  -0.5472  0.1255
## (Abuse - Control)       0.126 0.194 410  -0.2559  0.5080
## (Abuse - Combination)   0.390 0.169 410   0.0575  0.7232
## (Combination - Control) -0.342 0.139 410  -0.6154 -0.0682
##
## sigma used for effect sizes: 1.507
## Confidence level used: 0.95

```

```
cerq.d1.ctq <- my.lmcomp(cerq.mod, 'd1~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

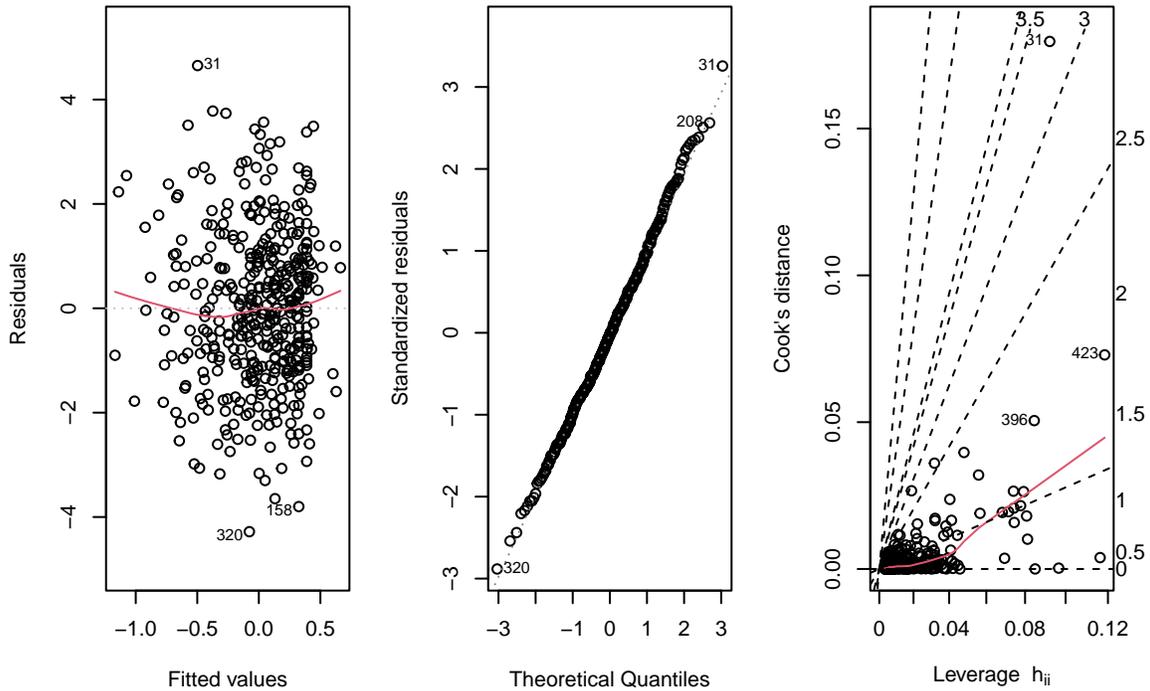


Figure S11: Diagnostic plots of the linear model $d1 \sim CTQ$ sub-scales for CERQ data.

3.2.2.2 Dimensional model

```

##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:
##   Min     1Q   Median     3Q    Max
## -4.2812 -1.0305 -0.0386  0.9567  4.6515
##

```

```

## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.77787    0.22397   3.473 0.000569 ***
## ctq.en      -0.06434    0.02076  -3.099 0.002075 **
## ctq.pa       0.01452    0.02954   0.491 0.623418
## ctq.ea       0.02516    0.02737   0.920 0.358337
## ctq.pn      -0.06759    0.03722  -1.816 0.070089 .
## ctq.sa       0.01446    0.02289   0.632 0.528017
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.499 on 410 degrees of freedom
## Multiple R-squared:  0.04866,    Adjusted R-squared:  0.03706
## F-statistic: 4.194 on 5 and 410 DF,  p-value: 0.0009955
##
## Anova Table (Type II tests)
##
## Response: d1
##           Sum Sq Df F value    Pr(>F)
## ctq.en      21.59  1  9.6044 0.002075 **
## ctq.pa       0.54  1  0.2415 0.623418
## ctq.ea       1.90  1  0.8456 0.358337
## ctq.pn       7.41  1  3.2981 0.070089 .
## ctq.sa       0.90  1  0.3989 0.528017
## Residuals 921.69 410
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----
## ctq.en    |                0.02 | [0.00, 1.00]
## ctq.pa    |                0.00 | [0.00, 1.00]
## ctq.ea    |                0.00 | [0.00, 1.00]
## ctq.pn    |             5.49e-03 | [0.00, 1.00]
## ctq.sa    |                0.00 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].

```

```
mod.comp(cerq.d1.group, cerq.d1.ctq)
```

```

##           model R2adj      AIC      BIC
## 1 cerq.d1.group 0.027 1529.823 1558.038
## 2  cerq.d1.ctq 0.037 1525.495 1553.710
## 3      Delta -0.010   4.328   4.328

```

3.2.3 Second component analysis

```
cerq.d2.group <- my.mcomp(aov.cerq, 'd2~group')
```

3.2.3.1 Categorical model

```

## Analysis of Variance Table
##
## Response: d2

```

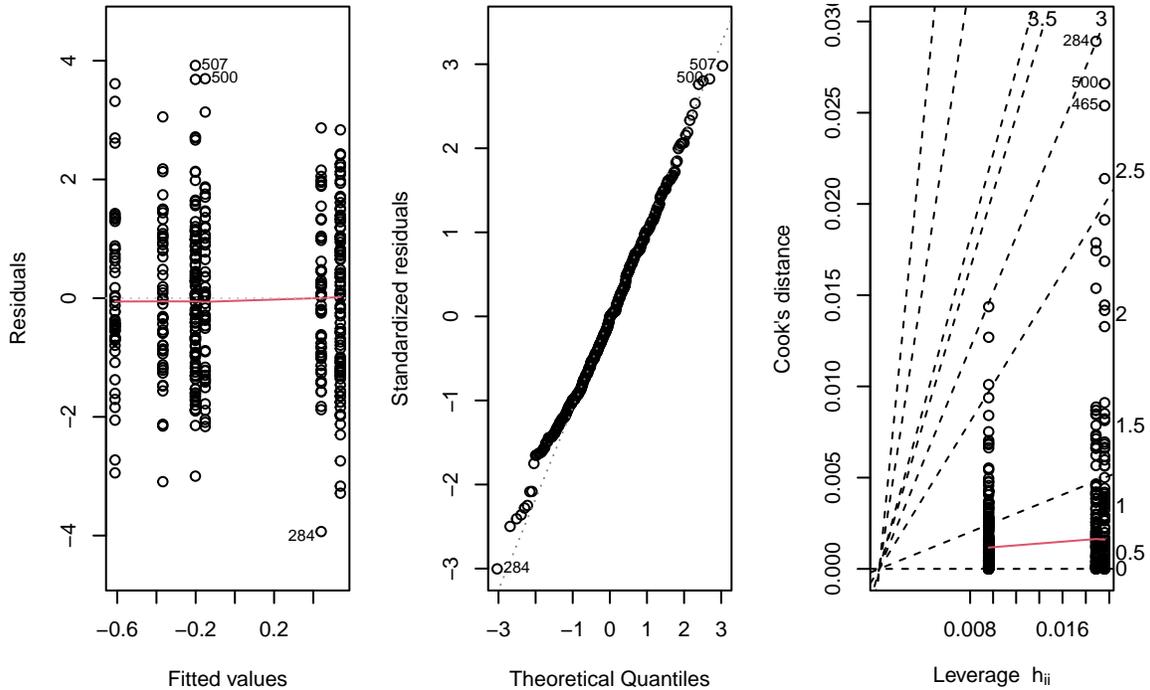


Figure S12: Diagnostic plots of the linear model $d2 \sim \text{group}$ for CERQ data.

```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## group      5  71.98  14.3968  8.2371 1.984e-07 ***
## Residuals 410 716.60  1.7478
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.
## # Effect Size for ANOVA
##
## Parameter | Omega2 |      95% CI
## -----
## group     |  0.08 | [0.04, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d2~group -----"
## group emmean  SE  df lower.CL upper.CL
## Neg   -0.151 0.185 410 -0.5151  0.2127
## Abu   0.440 0.182 410  0.0832  0.7972
## Combi 0.538 0.130 410  0.2834  0.7930
## LowA  -0.367 0.182 410 -0.7244 -0.0105
## LowN  -0.611 0.185 410 -0.9750 -0.2472
## LowC  -0.201 0.130 410 -0.4562  0.0534
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts d2~group -----"
## contrast          estimate    SE  df t.ratio p.value
## Neglect - Control      0.460 0.262 410  1.757 0.0797
```

```

## Neglect - Abuse          -0.591 0.259 410  -2.281  0.0231
## Neglect - Combination   -0.689 0.226 410  -3.051  0.0024
## Abuse - Control         0.808 0.257 410   3.145  0.0018
## Abuse - Combination     -0.098 0.223 410  -0.439  0.6607
## Combination - Control   0.740 0.183 410   4.034  0.0001
##
## contrast                effect.size    SE  df lower.CL upper.CL
## (Neglect - Control)      0.3479 0.198 410  -0.0421  0.7379
## (Neglect - Abuse)       -0.4474 0.197 410  -0.8342 -0.0605
## (Neglect - Combination) -0.5215 0.172 410  -0.8594 -0.1835
## (Abuse - Control)        0.6109 0.195 410   0.2268  0.9951
## (Abuse - Combination)   -0.0741 0.169 410  -0.4059  0.2577
## (Combination - Control) 0.5594 0.140 410   0.2841  0.8347
##
## sigma used for effect sizes: 1.322
## Confidence level used: 0.95

```

```
cerq.d2.ctq <- my.lmcomp(cerq.mod, 'd2~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

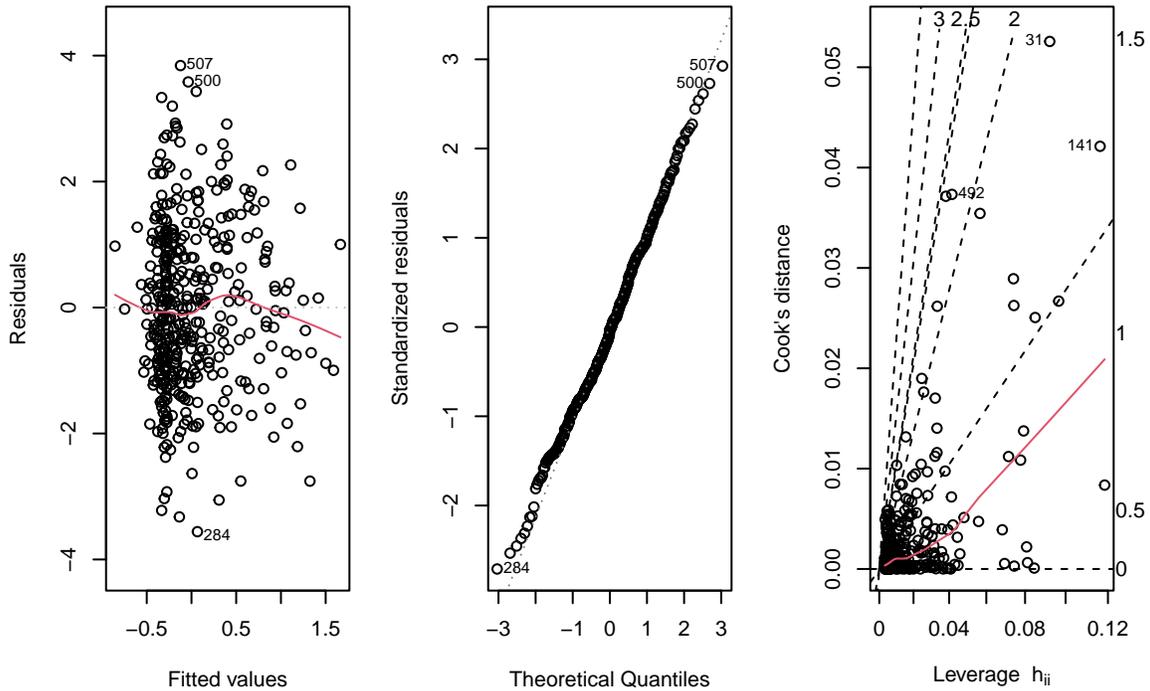


Figure S13: Diagnostic plots of the linear model d2~CTQ sub-scales for CERQ data.

3.2.3.2 Dimensional model

```

##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:
##   Min       1Q   Median       3Q      Max
## -3.5591 -0.9338 -0.0409  0.9565  3.8418
##

```

```

## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.60806    0.19654  -3.094  0.00211 **
## ctq.en      -0.01455    0.01822  -0.798  0.42508
## ctq.pa      -0.05275    0.02592  -2.035  0.04253 *
## ctq.ea       0.12803    0.02401   5.331 1.61e-07 ***
## ctq.pn      -0.01232    0.03266  -0.377  0.70622
## ctq.sa       0.01820    0.02009   0.906  0.36546
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.316 on 410 degrees of freedom
## Multiple R-squared:  0.09997,    Adjusted R-squared:  0.089
## F-statistic: 9.108 on 5 and 410 DF,  p-value: 3.15e-08
##
## Anova Table (Type II tests)
##
## Response: d2
##           Sum Sq Df F value    Pr(>F)
## ctq.en      1.10  1  0.6375  0.42508
## ctq.pa      7.17  1  4.1398  0.04253 *
## ctq.ea     49.20  1 28.4239 1.613e-07 ***
## ctq.pn      0.25  1  0.1423  0.70622
## ctq.sa      1.42  1  0.8209  0.36546
## Residuals 709.74 410
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----
## ctq.en    |                0.00 | [0.00, 1.00]
## ctq.pa    |                7.49e-03 | [0.00, 1.00]
## ctq.ea    |                0.06 | [0.03, 1.00]
## ctq.pn    |                0.00 | [0.00, 1.00]
## ctq.sa    |                0.00 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].

```

```

mod.comp(cerq.d2.group, cerq.d2.ctq)

```

```

##           model R2adj      AIC      BIC
## 1 cerq.d2.group  0.080 1420.789 1449.004
## 2  cerq.d2.ctq  0.089 1416.791 1445.006
## 3      Delta -0.009   3.997   3.997

```

3.3 DSI

3.3.1 PCA description

```
dsi <- opt.data[,c(27:30, 41, 6:10)]
pca.out <- check.pca(dsi[,1:4])

## Nb outliers: 1 -- 28

aov.dsi <- show.pca(dsi, qvs=c(5), qnvs=c(6:10), outliers=pca.out)

## [1] "----- Outliers influence -----"
## Contribution for outlier 28 to total variance: 0.612 %
## [1] " "
## [1] "----- Eigen values -----"
##          eigenvalue percentage of variance cumulative percentage of variance
## comp 1  2.1242382           53.105956           53.10596
## comp 2  1.0032656           25.081639           78.18760
## comp 3  0.5524315           13.810789           91.99838
## comp 4  0.3200646            8.001616           100.00000
## [1] " "
## [1] "----- Dimension description -----"
## $Dim.1
##
## Link between the variable and the continuous variables (R-square)
## =====
##          correlation      p.value
## DSIemotionalreact  0.884968904 1.669886e-139
## DSIfusionwithothers 0.847224077 8.584022e-116
## DSIiposition        0.777916304 1.452647e-85
## DSIemotionalcutoff  0.134632372 5.955136e-03
## ctq.pa              0.004802321 9.222068e-01
## ctq.pn              -0.083095268 9.052282e-02
## ctq.en              -0.091350684 6.267598e-02
## ctq.ea              -0.091390048 6.256241e-02
## ctq.sa              -0.105738453 3.106740e-02
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##          R2    p.value
## group 0.0181512 0.1836181
##
## Link between variable and the categories of the categorical variables
## =====
##          Estimate    p.value
## group=LowC  0.2561012 0.04123812
## group=LowA  0.1257701 0.51386391
## group=LowN  0.1035180 0.60131392
## group=Neg   -0.0464917 0.79492323
## group=Abu   -0.2123199 0.24987030
## group=Combi -0.2265777 0.06346324
##
## $Dim.2
##
## Link between the variable and the continuous variables (R-square)
```

```

## =====
## correlation p.value
## DSImotionalcutoff 0.98733951 0.000000e+00
## DSImotionalreact 0.05575911 2.564893e-01
## ctq.sa -0.04607080 3.485867e-01
## DSIfusionwithothers -0.10707125 2.899585e-02
## DSIliposition -0.11769852 1.631828e-02
## ctq.pa -0.13791694 4.832701e-03
## ctq.pn -0.21651360 8.361891e-06
## ctq.en -0.25644449 1.134956e-07
## ctq.ea -0.26274354 5.368322e-08
##
## Link between the variable and the categorical variable (1-way anova)
## =====
## R2 p.value
## group 0.06403583 5.141364e-05
##
## Link between variable and the categories of the categorical variables
## =====
## Estimate p.value
## group=LowN 0.37912695 1.784470e-03
## group=LowC 0.15536942 2.923565e-02
## group=LowA 0.05484979 5.097370e-01
## group=Abu -0.05990405 8.172204e-01
## group=Neg -0.13791879 4.130985e-01
## group=Combi -0.39152331 1.841919e-05

```

3.3.2 First component analysis

```

dsi.mod <- cbind(aov.dsi, opt.data[,6:10])
dsi.d1.group <- my.mcomp(aov.dsi, 'd1~group')

```

3.3.2.1 Categorical model

```

## Analysis of Variance Table
##
## Response: d1
## Df Sum Sq Mean Sq F value Pr(>F)
## group 5 16.04 3.2080 1.5159 0.1836
## Residuals 410 867.64 2.1162
##
## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.
## # Effect Size for ANOVA
##
## Parameter | Omega2 | 95% CI
## -----
## group | 6.16e-03 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d1~group -----"
## group emmean SE df lower.CL upper.CL
## Neg -0.0498 0.204 410 -0.4503 0.3506
## Abu -0.2157 0.200 410 -0.6085 0.1771

```

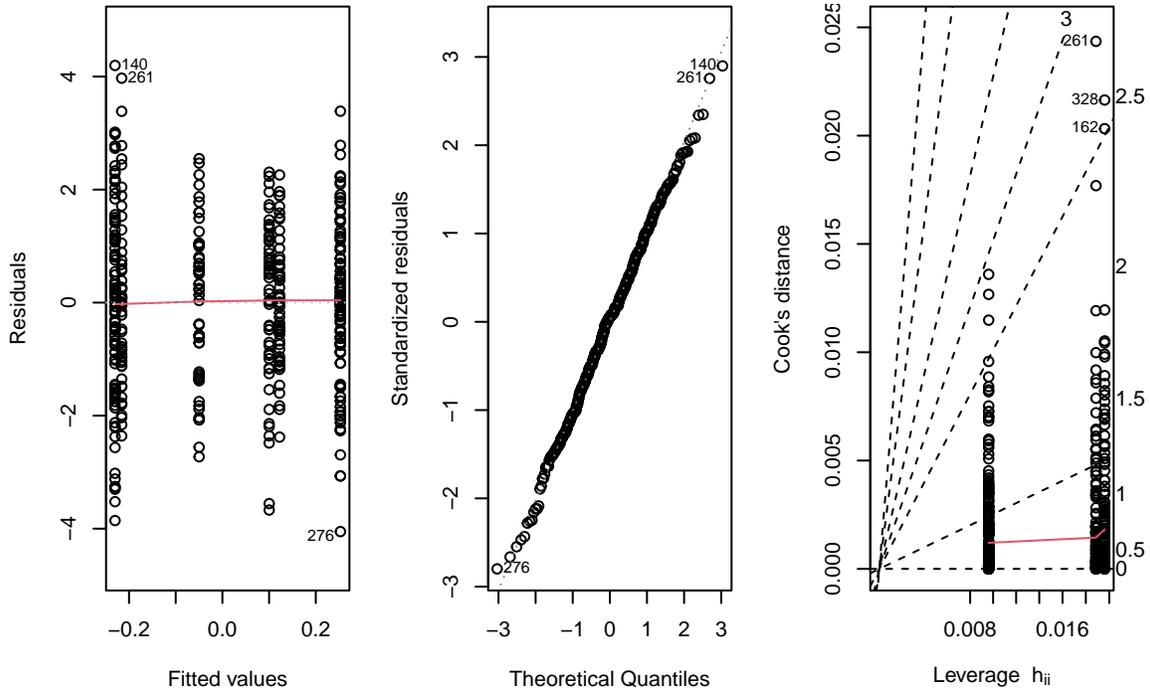


Figure S14: Diagnostic plots of the linear model $d1 \sim \text{group}$ for DSI data.

```
## Combi -0.2299 0.143 410 -0.5103 0.0505
## LowA 0.1224 0.200 410 -0.2704 0.5152
## LowN 0.1002 0.204 410 -0.3003 0.5006
## LowC 0.2528 0.143 410 -0.0277 0.5332
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts d1-group -----"
## contrast estimate SE df t.ratio p.value
## Neglect - Control -0.1500 0.288 410 -0.521 0.6028
## Neglect - Abuse 0.1658 0.285 410 0.581 0.5615
## Neglect - Combination 0.1801 0.249 410 0.724 0.4694
## Abuse - Control -0.3381 0.283 410 -1.196 0.2322
## Abuse - Combination 0.0143 0.246 410 0.058 0.9537
## Combination - Control -0.4827 0.202 410 -2.393 0.0172
##
## contrast effect.size SE df lower.CL upper.CL
## (Neglect - Control) -0.1031 0.198 410 -0.492 0.2862
## (Neglect - Abuse) 0.1140 0.196 410 -0.272 0.4997
## (Neglect - Combination) 0.1238 0.171 410 -0.212 0.4599
## (Abuse - Control) -0.2324 0.194 410 -0.615 0.1498
## (Abuse - Combination) 0.0098 0.169 410 -0.322 0.3416
## (Combination - Control) -0.3318 0.139 410 -0.605 -0.0582
##
## sigma used for effect sizes: 1.455
## Confidence level used: 0.95
```

```
dsi.d1.ctq <- my.lmcomp(dsi.mod, 'd1~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

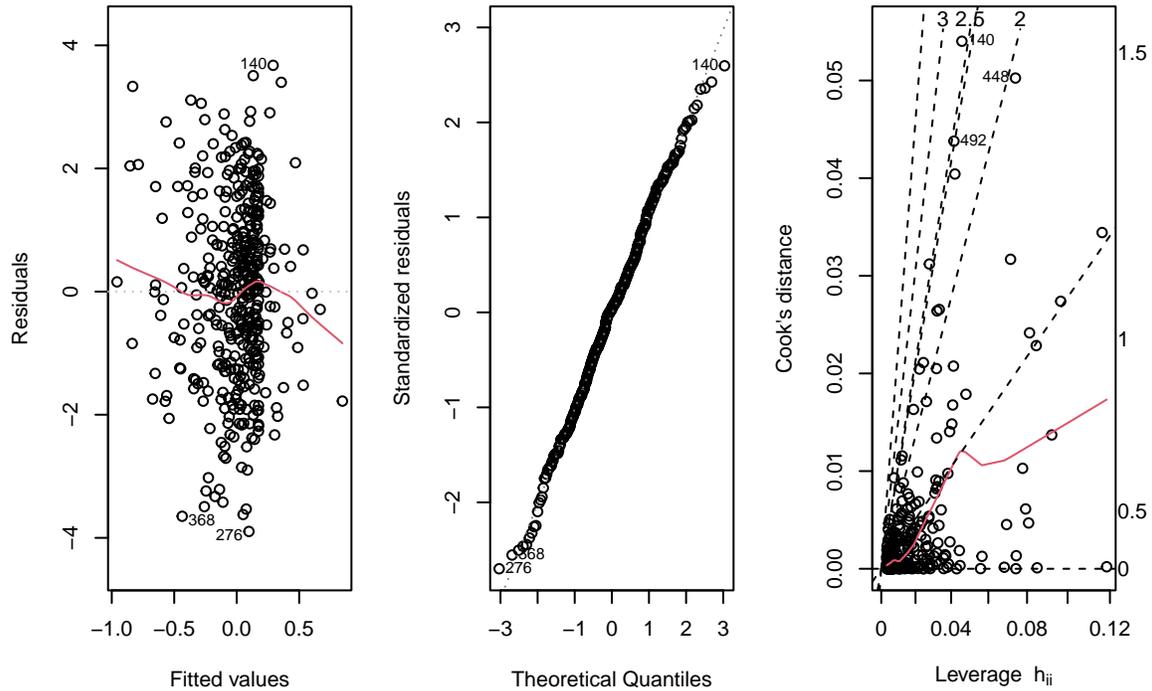


Figure S15: Diagnostic plots of the linear model $d1 \sim CTQ$ sub-scales for DSI data.

3.3.2.2 Dimensional model

```
##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.8964 -0.9874  0.0619  0.9454  3.6751
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.39471    0.21648   1.823  0.0690 .
## ctq.en      -0.01046    0.02007  -0.521  0.6024
## ctq.pa       0.05871    0.02856   2.056  0.0404 *
## ctq.ea      -0.02493    0.02645  -0.943  0.3464
## ctq.pn      -0.02883    0.03598  -0.801  0.4233
## ctq.sa      -0.03856    0.02213  -1.743  0.0822 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.449 on 410 degrees of freedom
## Multiple R-squared:  0.02556,    Adjusted R-squared:  0.01367
## F-statistic: 2.151 on 5 and 410 DF,  p-value: 0.0587
##
## Anova Table (Type II tests)
```

```
##
## Response: d1
##           Sum Sq Df F value Pr(>F)
## ctq.en      0.57  1  0.2718 0.60242
## ctq.pa      8.88  1  4.2265 0.04043 *
## ctq.ea      1.87  1  0.8885 0.34644
## ctq.pn      1.35  1  0.6423 0.42334
## ctq.sa      6.38  1  3.0363 0.08217 .
## Residuals 861.10 410
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----|-----|-----
## ctq.en    |          0.00 | [0.00, 1.00]
## ctq.pa    |       7.70e-03 | [0.00, 1.00]
## ctq.ea    |          0.00 | [0.00, 1.00]
## ctq.pn    |          0.00 | [0.00, 1.00]
## ctq.sa    |       4.87e-03 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].
```

```
mod.comp(dsi.d1.group, dsi.d1.ctq)
```

```
##           model R2adj      AIC      BIC
## 1 dsi.d1.group  0.006 1500.357 1528.571
## 2 dsi.d1.ctq   0.014 1497.207 1525.422
## 3      Delta -0.007   3.150   3.150
```

3.3.3 Second component analysis

```
dsi.d2.group <- my.mcomp(aov.dsi, 'd2~group')
```

3.3.3.1 Categorical model

```
## Analysis of Variance Table
##
## Response: d2
##           Df Sum Sq Mean Sq F value    Pr(>F)
## group      5  26.73  5.3452  5.6102 5.141e-05 ***
## Residuals 410 390.63  0.9528
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.
## # Effect Size for ANOVA
##
## Parameter | Omega2 |          95% CI
## -----|-----|-----
## group    |  0.05 | [0.02, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d2~group -----"
```

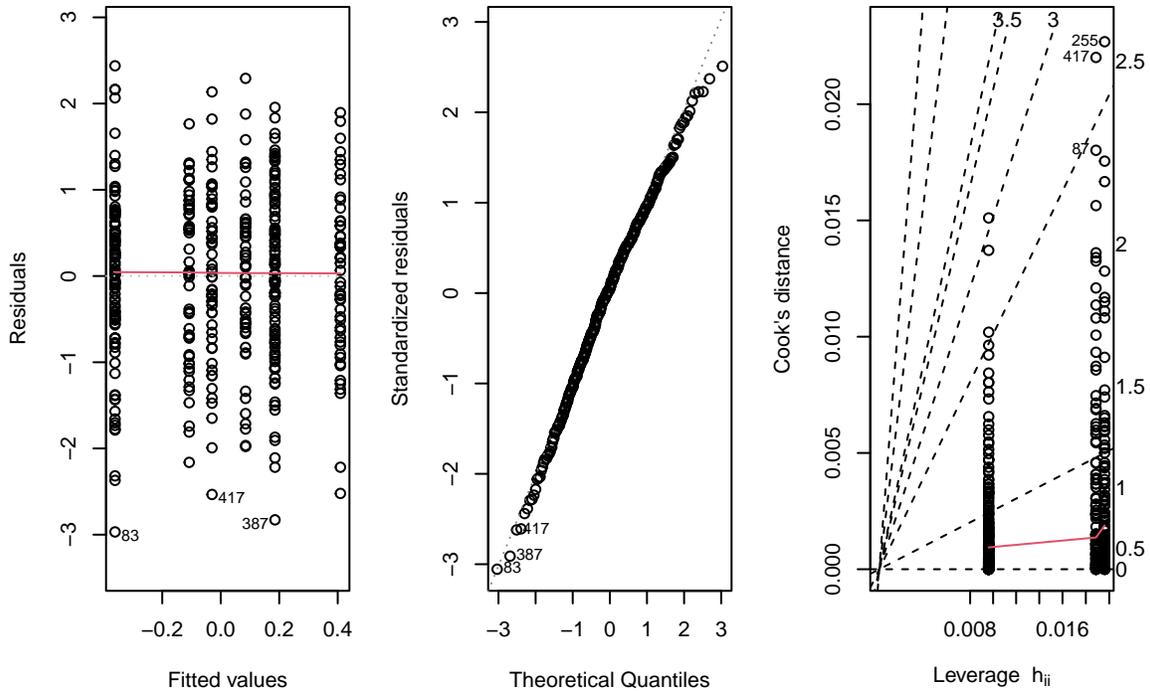


Figure S16: Diagnostic plots of the linear model $d2 \sim \text{group}$ for DSI data.

```
## group emmean SE df lower.CL upper.CL
## Neg -0.1078 0.1370 410 -0.37649 0.161
## Abu -0.0298 0.1340 410 -0.29336 0.234
## Combi -0.3614 0.0957 410 -0.54956 -0.173
## LowA 0.0850 0.1340 410 -0.17860 0.349
## LowN 0.4092 0.1370 410 0.14056 0.678
## LowC 0.1855 0.0957 410 -0.00267 0.374
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts d2-group -----"
## contrast estimate SE df t.ratio p.value
## Neglect - Control -0.517 0.193 410 -2.675 0.0078
## Neglect - Abuse -0.078 0.191 410 -0.407 0.6839
## Neglect - Combination 0.254 0.167 410 1.520 0.1293
## Abuse - Control -0.115 0.190 410 -0.605 0.5454
## Abuse - Combination 0.332 0.165 410 2.013 0.0448
## Combination - Control -0.547 0.135 410 -4.040 0.0001
##
## contrast effect.size SE df lower.CL upper.CL
## (Neglect - Control) -0.5297 0.199 410 -0.92068 -0.139
## (Neglect - Abuse) -0.0799 0.196 410 -0.46555 0.306
## (Neglect - Combination) 0.2598 0.171 410 -0.07670 0.596
## (Abuse - Control) -0.1176 0.194 410 -0.49951 0.264
## (Abuse - Combination) 0.3397 0.169 410 0.00716 0.672
## (Combination - Control) -0.5603 0.140 410 -0.83559 -0.285
##
## sigma used for effect sizes: 0.9761
## Confidence level used: 0.95
```



```

##
## Response: d2
##           Sum Sq Df F value Pr(>F)
## ctq.en      3.87  1  4.2025 0.04100 *
## ctq.pa      0.13  1  0.1370 0.71143
## ctq.ea      6.05  1  6.5615 0.01078 *
## ctq.pn      1.85  1  2.0031 0.15774
## ctq.sa      3.60  1  3.9079 0.04873 *
## Residuals 377.74 410
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----|-----|-----
## ctq.en    |          7.64e-03 | [0.00, 1.00]
## ctq.pa    |          0.00     | [0.00, 1.00]
## ctq.ea    |          0.01     | [0.00, 1.00]
## ctq.pn    |          2.41e-03 | [0.00, 1.00]
## ctq.sa    |          6.94e-03 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].

```

```

mod.comp(dsi.d2.group, dsi.d2.ctq)

```

```

##           model R2adj      AIC      BIC
## 1 dsi.d2.group  0.053 1168.383 1196.598
## 2  dsi.d2.ctq  0.084 1154.425 1182.640
## 3      Delta -0.031   13.958   13.958

```

3.4 HADS

3.4.1 PCA description

```
had <- opt.data[,c(32:33, 41, 6:10)]
pca.out <- check.pca(had[,1:2])

## Nb outliers: 0 -- None
aov.had <- show.pca(had, qvs=c(3), qnvs=c(4:8), outliers=pca.out)

## [1] "----- Eigen values -----"
##          eigenvalue percentage of variance cumulative percentage of variance
## comp 1  1.1398253          56.99126          56.99126
## comp 2  0.8601747          43.00874          100.00000
## [1] " "
## [1] "----- Dimension description -----"
## $Dim.1
##
## Link between the variable and the continuous variables (R-square)
## =====
##          correlation      p.value
## HADSanxiety      0.7549256 7.256420e-78
## HADSdepression  0.7549256 7.256420e-78
## ctq.en          0.3663352 1.169755e-14
## ctq.ea          0.3627741 2.194744e-14
## ctq.pn          0.2678635 2.878968e-08
## ctq.sa          0.1285062 8.689746e-03
## ctq.pa          0.1077429 2.799758e-02
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##          R2      p.value
## group 0.119305 4.734844e-10
##
## Link between variable and the categories of the categorical variables
## =====
##          Estimate      p.value
## group=Combi  0.629429405 7.904423e-11
## group=Neg    0.192966939 3.184321e-01
## group=Abu    0.003380728 7.184243e-01
## group=LowA   -0.255484335 2.436003e-02
## group=LowC   -0.204546742 4.439423e-03
## group=LowN   -0.365745995 2.726967e-03
##
## $Dim.2
##
## Link between the variable and the continuous variables (R-square)
## =====
##          correlation      p.value
## HADSdepression  0.65581046 1.687882e-52
## ctq.en          0.16165399 9.363800e-04
## ctq.pn          0.11110649 2.343064e-02
## ctq.pa          0.09249904 5.943201e-02
## ctq.ea          0.08950662 6.818945e-02
```

```
## ctq.sa          0.01790129 7.158232e-01
## HADSanxiety    -0.65581046 1.687882e-52
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##              R2      p.value
## group 0.06395825 5.220658e-05
##
## Link between variable and the categories of the categorical variables
## =====
##              Estimate    p.value
## group=Combi  0.23617666 0.002779429
## group=Neg    0.28815319 0.018272136
## group=LowN   0.04653701 0.709922476
## group=LowA   0.02998343 0.809151377
## group=LowC  -0.21362458 0.006312730
## group=Abu   -0.38722571 0.001056989
```

3.4.2 First component analysis

```
had.mod <- cbind(aov.had, opt.data[,6:10])
had.d1.group <- my.mcomp(aov.had, 'd1~group')
```

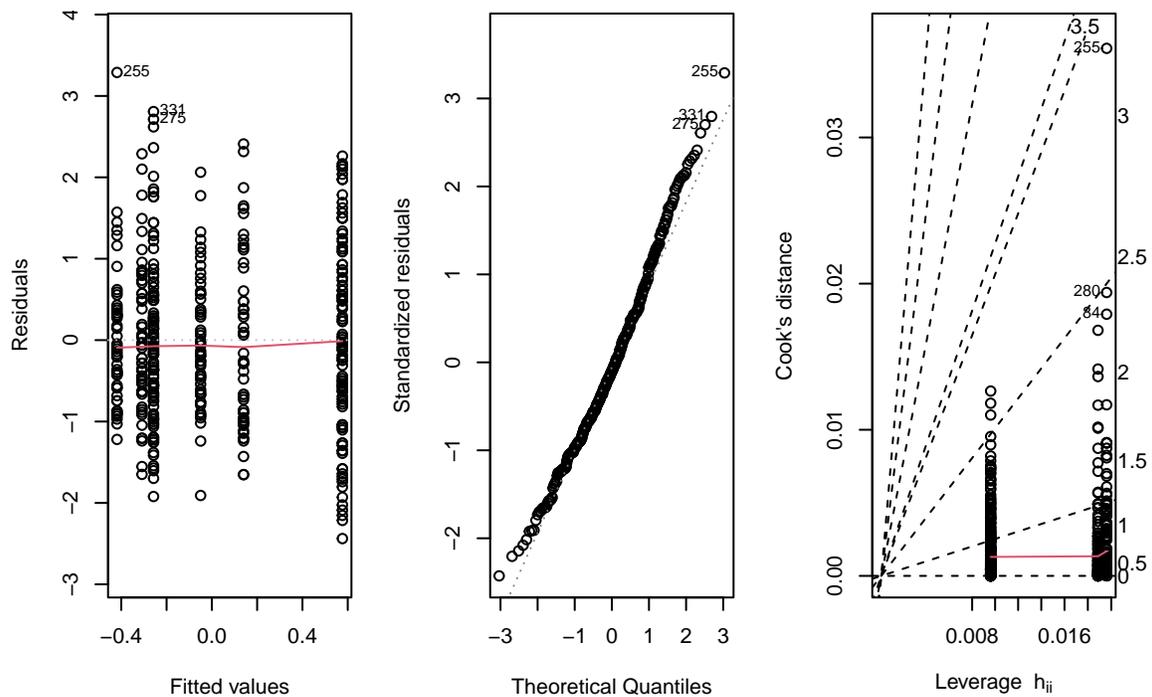


Figure S18: Diagnostic plots of the linear model $d1 \sim \text{group}$ for HAD data.

3.4.2.1 Categorical model

```
## Analysis of Variance Table
##
## Response: d1
##              Df Sum Sq Mean Sq F value    Pr(>F)
```

```

## group      5  56.57 11.3141  11.108 4.735e-10 ***
## Residuals 410 417.60  1.0185
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.

## # Effect Size for ANOVA
##
## Parameter | Omega2 |      95% CI
## -----
## group     |  0.11 | [0.06, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d1~group -----"
## group emmean  SE  df lower.CL upper.CL
## Neg    0.1400 0.141 410  -0.138  0.4178
## Abu   -0.0495 0.139 410  -0.322  0.2230
## Combi  0.5765 0.099 410   0.382  0.7710
## LowA  -0.3084 0.139 410  -0.581 -0.0359
## LowN  -0.4187 0.141 410  -0.696 -0.1409
## LowC  -0.2575 0.099 410  -0.452 -0.0629
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts  d1~group -----"
## contrast          estimate    SE  df t.ratio p.value
## Neglect - Control      0.559 0.200 410   2.796  0.0054
## Neglect - Abuse        0.190 0.198 410   0.958  0.3388
## Neglect - Combination -0.436 0.173 410  -2.530  0.0118
## Abuse - Control        0.259 0.196 410   1.320  0.1874
## Abuse - Combination   -0.626 0.170 410  -3.676  0.0003
## Combination - Control  0.834 0.140 410   5.959 <.0001
##
## contrast          effect.size    SE  df lower.CL upper.CL
## (Neglect - Control)    0.554 0.199 410   0.162  0.9447
## (Neglect - Abuse)     0.188 0.196 410  -0.198  0.5737
## (Neglect - Combination) -0.432 0.172 410  -0.770 -0.0951
## (Abuse - Control)     0.256 0.194 410  -0.126  0.6388
## (Abuse - Combination) -0.620 0.170 410  -0.955 -0.2858
## (Combination - Control) 0.826 0.142 410   0.548  1.1048
##
## sigma used for effect sizes: 1.009
## Confidence level used: 0.95

```

```
had.d1.ctq <- my.lmcomp(had.mod, 'd1~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

3.4.2.2 Dimensional model

```

##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:

```

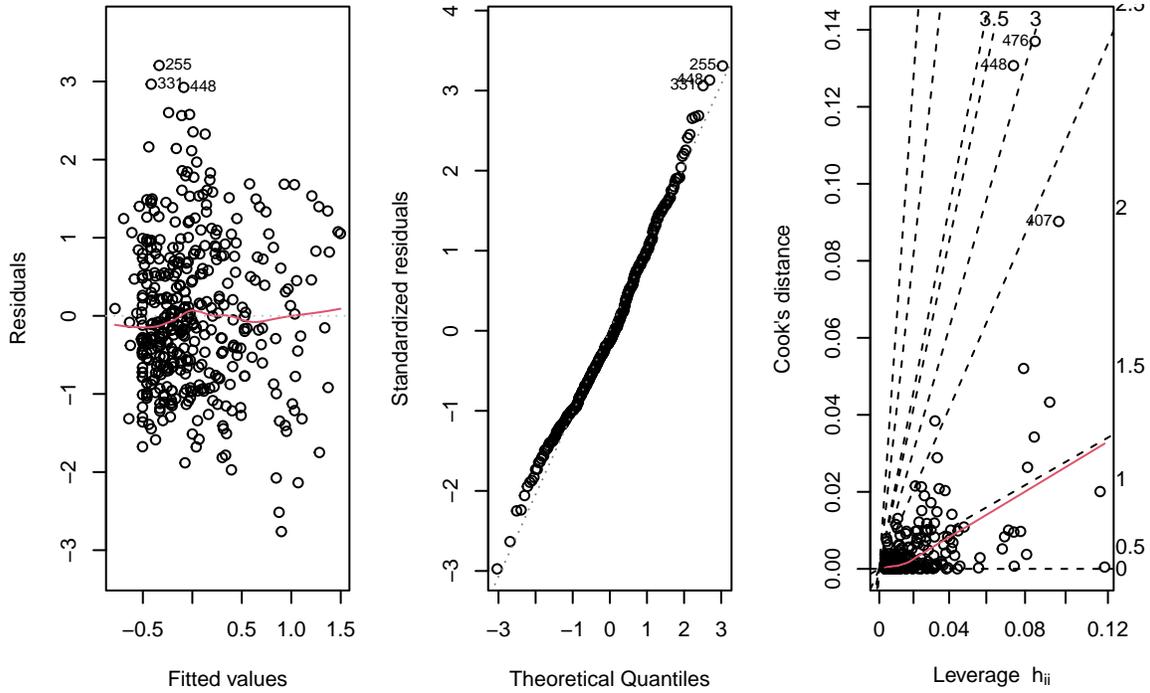


Figure S19: Diagnostic plots of the linear model $d1 \sim CTQ$ sub-scales for HAD data.

```
##      Min      1Q  Median      3Q      Max
## -2.7618 -0.6644 -0.1191  0.6713  3.2054
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.961140  0.145034  -6.627 1.08e-10 ***
## ctq.en       0.043959  0.013444   3.270 0.00117 **
## ctq.pa      -0.057926  0.019131  -3.028 0.00262 **
## ctq.ea       0.077551  0.017721   4.376 1.53e-05 ***
## ctq.pn       0.035980  0.024102   1.493 0.13626
## ctq.sa      -0.008006  0.014825  -0.540 0.58949
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9709 on 410 degrees of freedom
## Multiple R-squared:  0.1849, Adjusted R-squared:  0.1749
## F-statistic: 18.6 on 5 and 410 DF, p-value: < 2.2e-16
##
## Anova Table (Type II tests)
##
## Response: d1
##           Sum Sq Df F value    Pr(>F)
## ctq.en    10.08  1 10.6916  0.001167 **
## ctq.pa     8.64  1  9.1681  0.002618 **
## ctq.ea    18.05  1 19.1513 1.534e-05 ***
## ctq.pn     2.10  1  2.2284  0.136262
## ctq.sa     0.27  1  0.2916  0.589494
## Residuals 386.50 410
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |      95% CI
## -----|-----|-----
## ctq.en    |          0.02 | [0.00, 1.00]
## ctq.pa    |          0.02 | [0.00, 1.00]
## ctq.ea    |          0.04 | [0.02, 1.00]
## ctq.pn    |       2.94e-03 | [0.00, 1.00]
## ctq.sa    |          0.00 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].
```

```
mod.comp(had.d1.group, had.d1.ctq)
```

```
##          model R2adj      AIC      BIC
## 1 had.d1.group 0.109 1196.151 1224.365
## 2 had.d1.ctq  0.175 1163.962 1192.177
## 3      Delta -0.066   32.189   32.189
```

3.4.3 Second component analysis

```
had.d2.group <- my.mcomp(aov.had, 'd2~group')
```

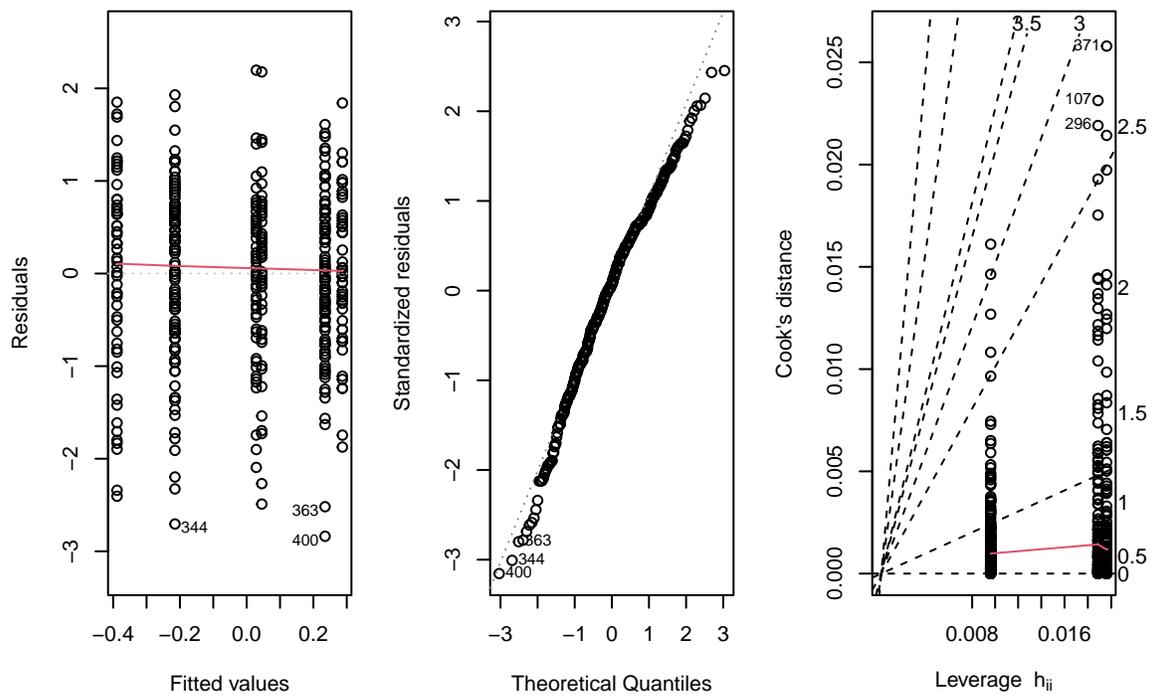


Figure S20: Diagnostic plots of the linear model $d2 \sim \text{group}$ for HAD data.

3.4.3.1 Categorical model

```
## Analysis of Variance Table
##
## Response: d2
##          Df Sum Sq Mean Sq F value    Pr(>F)
```

```

## group      5  22.89  4.5773  5.6029  5.221e-05 ***
## Residuals 410 334.95  0.8169
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.

## # Effect Size for ANOVA
##
## Parameter | Omega2 |      95% CI
## -----
## group     |  0.05 | [0.02, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d2~group -----"
## group emmean      SE df lower.CL upper.CL
## Neg    0.2870 0.1270 410  0.0382  0.5358
## Abu   -0.3884 0.1240 410 -0.6324 -0.1443
## Combi  0.2350 0.0886 410  0.0608  0.4092
## LowA   0.0288 0.1240 410 -0.2152  0.2729
## LowN   0.0454 0.1270 410 -0.2034  0.2942
## LowC  -0.2148 0.0886 410 -0.3890 -0.0406
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts d2~group -----"
## contrast          estimate      SE df t.ratio p.value
## Neglect - Control      0.242 0.179 410  1.350 0.1778
## Neglect - Abuse        0.675 0.177 410  3.809 0.0002
## Neglect - Combination  0.052 0.155 410  0.336 0.7367
## Abuse - Control       -0.417 0.176 410 -2.376 0.0180
## Abuse - Combination   -0.623 0.153 410 -4.087 0.0001
## Combination - Control  0.450 0.125 410  3.589 0.0004
##
## contrast          effect.size      SE df lower.CL upper.CL
## (Neglect - Control)      0.2673 0.198 410  -0.122  0.6570
## (Neglect - Abuse)        0.7472 0.198 410   0.358  1.1362
## (Neglect - Combination)  0.0575 0.171 410  -0.279  0.3936
## (Abuse - Control)       -0.4616 0.195 410  -0.845 -0.0784
## (Abuse - Combination)   -0.6897 0.170 410  -1.025 -0.3546
## (Combination - Control)  0.4977 0.140 410   0.223  0.7724
##
## sigma used for effect sizes: 0.9038
## Confidence level used: 0.95

```

```
had.d2.ctq <- my.lmcomp(had.mod, 'd2~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

3.4.3.2 Dimensional model

```

##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:

```

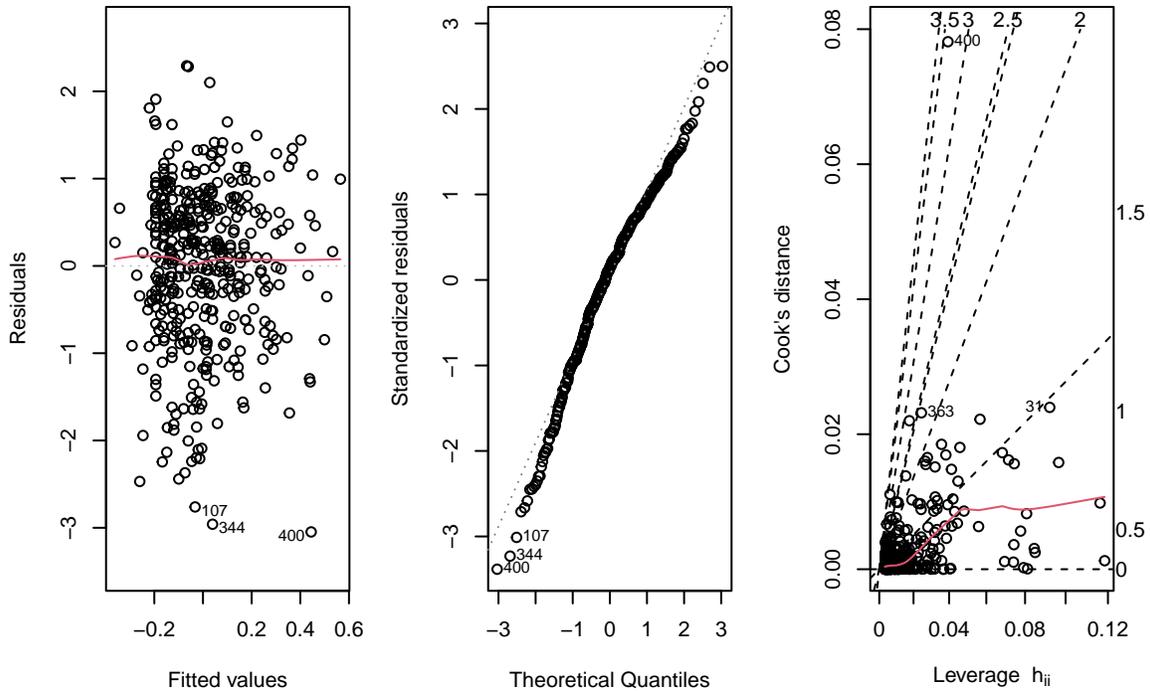


Figure S21: Diagnostic plots of the linear model $d2 \sim CTQ$ sub-scales for HAD data.

```
##      Min      1Q  Median      3Q      Max
## -3.0474 -0.5470  0.1317  0.6564  2.2923
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.39907    0.13727  -2.907  0.00385 **
## ctq.en       0.03329    0.01272   2.616  0.00922 **
## ctq.pa       0.02044    0.01811   1.129  0.25970
## ctq.ea      -0.01331    0.01677  -0.793  0.42803
## ctq.pn       0.01382    0.02281   0.606  0.54485
## ctq.sa      -0.01317    0.01403  -0.938  0.34855
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.919 on 410 degrees of freedom
## Multiple R-squared:  0.03236,    Adjusted R-squared:  0.02056
## F-statistic: 2.743 on 5 and 410 DF,  p-value: 0.01883
##
## Anova Table (Type II tests)
##
## Response: d2
##           Sum Sq Df F value  Pr(>F)
## ctq.en     5.78  1  6.8439 0.009223 **
## ctq.pa     1.08  1  1.2739 0.259702
## ctq.ea     0.53  1  0.6294 0.428029
## ctq.pn     0.31  1  0.3672 0.544855
## ctq.sa     0.74  1  0.8808 0.348547
## Residuals 346.25 410
## ---
```

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----|-----|-----
## ctq.en    |          0.01 | [0.00, 1.00]
## ctq.pa    |        6.58e-04 | [0.00, 1.00]
## ctq.ea    |          0.00 | [0.00, 1.00]
## ctq.pn    |          0.00 | [0.00, 1.00]
## ctq.sa    |          0.00 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].

```

```

mod.comp(had.d2.group, had.d2.ctq)

```

```

##          model R2adj      AIC      BIC
## 1 had.d2.group 0.053 1104.403 1132.618
## 2  had.d2.ctq 0.021 1118.213 1146.428
## 3      Delta 0.032  -13.810  -13.810

```

3.5 SPS

3.5.1 PCA description

```
sps <- opt.data[,c(35:38, 41, 6:10)]
pca.out <- check.pca(sps[,1:4])

## Nb outliers: 1 -- 396

aov.sps <- show.pca(sps, qvs=c(5), qnvs=c(6:10), outliers=pca.out)

## [1] "----- Outliers influence -----"
## Contribution for outlier 396 to total variance: 1.424 %
## [1] " "
## [1] "----- Eigen values -----"
##          eigenvalue percentage of variance cumulative percentage of variance
## comp 1  2.8829643          72.074107          72.07411
## comp 2  0.5009278          12.523196          84.59730
## comp 3  0.3835958           9.589895          94.18720
## comp 4  0.2325121           5.812802         100.00000
## [1] " "
## [1] "----- Dimension description -----"
## $Dim.1
##
## Link between the variable and the continuous variables (R-square)
## =====
##          correlation      p.value
## SPShopelessness      0.9023737 1.959510e-153
## SPSsuicideideation    0.8609233 1.397414e-123
## SPSnegativeselvaluation 0.8427008 2.180819e-113
## SPShostility          0.7857178 2.170831e-88
## ctq.en                0.5278642 3.201746e-31
## ctq.ea                0.4287662 4.941718e-20
## ctq.pn                0.3273475 7.592296e-12
## ctq.sa                0.2289904 2.366526e-06
## ctq.pa                0.2067101 2.143721e-05
##
## Link between the variable and the categorical variable (1-way anova)
## =====
##          R2      p.value
## group 0.2086644 3.184221e-19
##
## Link between variable and the categories of the categorical variables
## =====
##          Estimate      p.value
## group=Combi 1.26391195 5.384916e-17
## group=Neg   0.52201228 5.946692e-02
## group=Abu   0.08967889 9.553779e-01
## group=LowA  -0.64366224 5.898380e-04
## group=LowC  -0.44313904 1.437220e-04
## group=LowN  -0.78880183 5.659376e-05
##
## $Dim.2
##
## Link between the variable and the continuous variables (R-square)
```

```
## =====
## correlation p.value
## SPShostility 0.588795850 3.566674e-40
## ctq.en 0.105146096 3.202796e-02
## ctq.pn 0.019462415 6.922526e-01
## SPSSuicideideation -0.005720813 9.073894e-01
## ctq.ea -0.025637374 6.020783e-01
## ctq.pa -0.040128662 4.143095e-01
## ctq.sa -0.156656384 1.348870e-03
## SPShopelessness -0.182491694 1.822798e-04
## SPShopelessness -0.347723094 2.880194e-13
##
## Link between the variable and the categorical variable (1-way anova)
## =====
## R2 p.value
## group 0.007237064 0.7017036
##
## Link between variable and the categories of the categorical variables
## =====
## Estimate p.value
## group=Neg 0.116202939 0.2263527
## group=LowC 0.033085969 0.6253444
## group=Combi 0.000805077 0.9623136
## group=LowN -0.011925312 0.8671228
## group=LowA -0.025872596 0.7458111
## group=Abu -0.112296077 0.2025998
```

3.5.2 First component analysis

```
sps.mod <- cbind(aov.sps, opt.data[,6:10])
sps.d1.group <- my.mcomp(aov.sps, 'd1~group')
```

3.5.2.1 Categorical model

```
## Analysis of Variance Table
##
## Response: d1
## Df Sum Sq Mean Sq F value Pr(>F)
## group 5 250.25 50.051 21.622 < 2.2e-16 ***
## Residuals 410 949.06 2.315
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## For one-way between subjects designs, partial omega squared is
## equivalent to omega squared. Returning omega squared.
## # Effect Size for ANOVA
##
## Parameter | Omega2 | 95% CI
## -----
## group | 0.20 | [0.14, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].[1] " "
## [1] "----- Estimated marginal means for d1~group -----"
## group emmean SE df lower.CL upper.CL
```

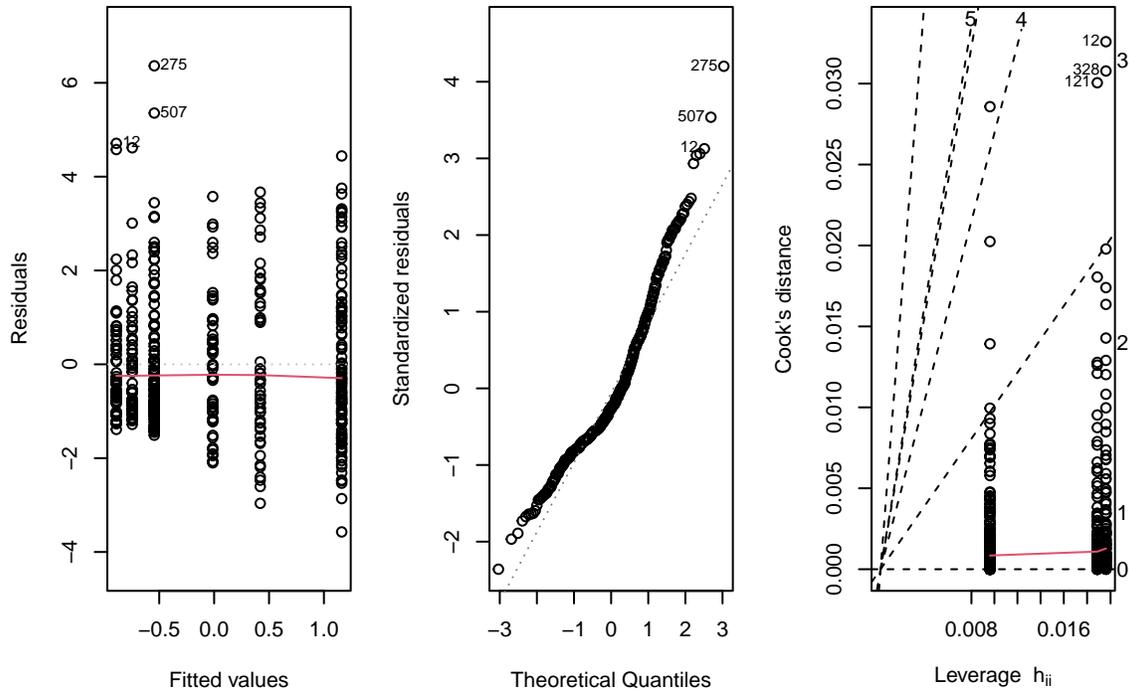


Figure S22: Diagnostic plots of the linear model $d1 \sim \text{group}$ for SPS data.

```
## Neg    0.4201 0.213 410  0.00131    0.839
## Abu   -0.0122 0.209 410 -0.42304    0.399
## Combi 1.1620 0.149 410  0.86873    1.455
## LowA  -0.7456 0.209 410 -1.15639   -0.335
## LowN  -0.8907 0.213 410 -1.30950   -0.472
## LowC  -0.5450 0.149 410 -0.83832   -0.252
##
## Confidence level used: 0.95
## [1] " "
## [1] "----- Contrasts  d1~group  -----"
## contrast          estimate      SE  df t.ratio p.value
## Neglect - Control      1.311 0.301 410   4.351 <.0001
## Neglect - Abuse         0.432 0.298 410   1.449 0.1482
## Neglect - Combination  -0.742 0.260 410  -2.853 0.0046
## Abuse - Control         0.733 0.296 410   2.481 0.0135
## Abuse - Combination    -1.174 0.257 410  -4.573 <.0001
## Combination - Control   1.707 0.211 410   8.091 <.0001
##
## contrast          effect.size    SE  df lower.CL upper.CL
## (Neglect - Control)      0.862 0.200 410   0.4678  1.255
## (Neglect - Abuse)        0.284 0.196 410  -0.1019  0.670
## (Neglect - Combination) -0.488 0.172 410  -0.8253 -0.150
## (Abuse - Control)        0.482 0.195 410   0.0987  0.865
## (Abuse - Combination)   -0.772 0.171 410  -1.1078 -0.436
## (Combination - Control)  1.122 0.144 410   0.8387  1.405
##
## sigma used for effect sizes: 1.521
## Confidence level used: 0.95
```

```
sps.d1.ctq <- my.lmcomp(sps.mod, 'd1~ctq.en+ctq.pa+ctq.ea+ctq.pn+ctq.sa')
```

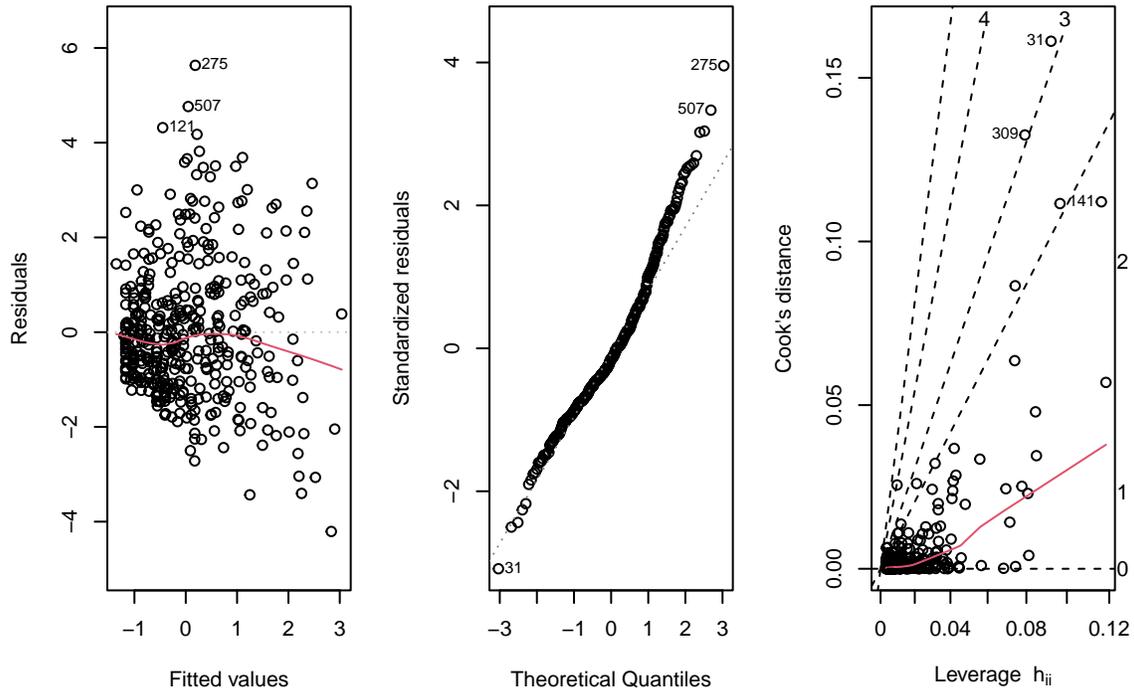


Figure S23: Diagnostic plots of the linear model $d1 \sim CTQ$ sub-scales for SPS data.

3.5.2.2 Dimensional model

```
##
## Call:
## lm(formula = as.formula(form), data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.2071 -0.9509 -0.2181  0.7375  5.6330
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.26652    0.21387  -10.598 < 2e-16 ***
## ctq.en       0.15030    0.01982   7.582  2.3e-13 ***
## ctq.pa      -0.03621    0.02821  -1.284  0.20000
## ctq.ea       0.07103    0.02613   2.718  0.00685 **
## ctq.pn       0.01135    0.03554   0.319  0.74968
## ctq.sa       0.02325    0.02186   1.064  0.28807
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.432 on 410 degrees of freedom
## Multiple R-squared:  0.2992, Adjusted R-squared:  0.2907
## F-statistic: 35.02 on 5 and 410 DF, p-value: < 2.2e-16
##
## Anova Table (Type II tests)
```

```

##
## Response: d1
##          Sum Sq Df F value  Pr(>F)
## ctq.en    117.83  1 57.4834 2.3e-13 ***
## ctq.pa     3.38  1  1.6477 0.199999
## ctq.ea    15.14  1  7.3879 0.006845 **
## ctq.pn     0.21  1  0.1019 0.749677
## ctq.sa     2.32  1  1.1315 0.288075
## Residuals 840.43 410
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## # Effect Size for ANOVA (Type II)
##
## Parameter | Omega2 (partial) |          95% CI
## -----|-----|-----
## ctq.en    |          0.12 | [0.07, 1.00]
## ctq.pa    |       1.55e-03 | [0.00, 1.00]
## ctq.ea    |          0.02 | [0.00, 1.00]
## ctq.pn    |          0.00 | [0.00, 1.00]
## ctq.sa    |       3.16e-04 | [0.00, 1.00]
##
## - One-sided CIs: upper bound fixed at [1.00].

```

```

mod.comp(sps.d1.group, sps.d1.ctq)

```

```

##          model R2adj      AIC      BIC
## 1 sps.d1.group 0.199 1537.668 1565.883
## 2 sps.d1.ctq  0.291 1487.101 1515.316
## 3      Delta -0.092  50.567  50.567

```

4 Session information

sessionInfo()

```
## R version 4.5.2 (2025-10-31)
## Platform: x86_64-pc-linux-gnu
## Running under: Debian GNU/Linux forky/sid
##
## Matrix products: default
## BLAS: /usr/lib/x86_64-linux-gnu/blas/libblas.so.3.12.1
## LAPACK: /usr/lib/x86_64-linux-gnu/lapack/liblapack.so.3.12.1; LAPACK version 3.12.0
##
## locale:
## [1] LC_CTYPE=fr_FR.UTF-8 LC_NUMERIC=C
## [3] LC_TIME=C LC_COLLATE=fr_FR.UTF-8
## [5] LC_MONETARY=fr_FR.UTF-8 LC_MESSAGES=fr_FR.UTF-8
## [7] LC_PAPER=fr_FR.UTF-8 LC_NAME=C
## [9] LC_ADDRESS=C LC_TELEPHONE=C
## [11] LC_MEASUREMENT=fr_FR.UTF-8 LC_IDENTIFICATION=C
##
## time zone: Europe/Paris
## tzcode source: system (glibc)
##
## attached base packages:
## [1] stats graphics grDevices datasets utils methods base
##
## other attached packages:
## [1] car_3.1-3 carData_3.0-5 glue_1.8.0 effectsize_1.0.1
## [5] emmeans_1.11.0 factoextra_1.0.7 ggplot2_3.5.2 FactoMineR_2.11
## [9] MatchIt_4.7.0 mclust_6.1.2 knitr_1.50
##
## loaded via a namespace (and not attached):
## [1] generics_0.1.3 renv_1.1.5 lattice_0.22-7
## [4] digest_0.6.37 magrittr_2.0.3 evaluate_1.0.3
## [7] grid_4.5.2 estimability_1.5.1 RColorBrewer_1.1-3
## [10] rlemon_0.2.1 mvtnorm_1.3-3 fastmap_1.2.0
## [13] ggrepel_0.9.6 backports_1.5.0 Formula_1.2-5
## [16] tinytex_0.57 scales_1.4.0 abind_1.4-8
## [19] cli_3.6.5 rlang_1.1.6 chk_0.10.0
## [22] scatterplot3d_0.3-44 optmatch_0.10.8 leaps_3.2
## [25] withr_3.0.2 yaml_2.3.10 datawizard_1.1.0
## [28] tools_4.5.2 multcompView_0.1-10 checkmate_2.3.4
## [31] dplyr_1.1.4 DT_0.33 bayestestR_0.16.0
## [34] flashClust_1.01-2 vctrs_0.6.5 R6_2.6.1
## [37] lifecycle_1.0.4 htmlwidgets_1.6.4 MASS_7.3-65
## [40] insight_1.3.0 cluster_2.1.8.1 pkgconfig_2.0.3
## [43] pillar_1.10.2 gtable_0.3.6 Rcpp_1.0.14
## [46] xfun_0.52 tibble_3.2.1 tidyselect_1.2.1
## [49] parameters_0.26.0 highs_1.9.0-1 farver_2.1.2
## [52] htmltools_0.5.8.1 rmarkdown_2.29 compiler_4.5.2
```