

# Meta-Analysis of the Big-Fish-Little-Pond-Effect in Mathematics Utilizing TIMSS Data

Supplementary Material

28 October 2021

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# 1 Data preparation

## 1.1 Install and load relevant R packages

```
# Install relevant packages
# install.packages("metafor")
# install.packages("metaSEM")
# install.packages("robumeta")
# install.packages("clubSandwich")
# install.packages("tidyverse")
# install.packages("ggplot2")
# install.packages("forestplot")
# install.packages("psych")
# install.packages("ggpubr")
# install.packages("meta")
# install.packages("metaforest")
# install.packages("caret")
# install.packages("zcurve")
# install.packages("PublicationBias")

# Load packages after every reboot of computer
library(metafor)
library(metaSEM)
library(robumeta)
library(clubSandwich)
library(tidyverse)
library(dmetar)
library(ggplot2)
library(forestplot)
library(psych)
library(ggpubr)
library(meta)
library(metaforest)
library(caret)
library(zcurve)
library(PublicationBias)
```

## 1.2 Attach the data set

```
# Read in data set
bflpedat <- read.csv2("bflpe-timss.csv", header = TRUE, dec = ",")

# Check the data
head(bflpedat, 5)
```

```
##      ESID StudyID country      country_name pdi idv mas uai  ltowvs      ivr
```

```
## 1 1 2019 AAD Abu Dhabi 80 38 53 68 23.00000 34.00000
## 2 2 2019 ADU Dubai, UAE 80 38 53 68 23.00000 34.00000
## 3 3 2019 ALB Albania NA NA NA NA 61.46096 14.50893
## 4 4 2019 ARE United Arab Emirates 80 38 53 68 23.00000 34.00000
## 5 5 2019 ARM Armenia NA NA NA NA 60.95718 NA
## year effect_size std_se pvalue se var HDI N
## 1 2019 -0.625 -7.046 0 0.089 0.00777350 0.890 4149
## 2 2019 -0.441 -7.164 0 0.062 0.00326685 0.890 4097
## 3 2019 -0.542 -5.237 0 0.104 0.01022060 0.795 2861
## 4 2019 -0.620 -12.260 0 0.051 0.00252562 0.890 4669
## 5 2019 -0.326 -4.775 0 0.068 0.00498966 0.776 1565
```

```
## Number of effect sizes
```

```
nrow(bflpedat)
```

```
## [1] 248
```

```
## List of countries
```

```
table(bflpedat$country)
```

```
##
```

```
## AAD ABA ADU ALB ARE ARM AUS AUT AZE BFL BGR BHR BIH CAB CAN CBC CHL COL COT CQU
## 3 1 4 1 3 5 5 3 2 4 2 3 1 2 2 1 3 1 5 5
## CYP CZE DEU DNK DZA EMA ENG ESP FIN FRA GEO HKG HRV HUN IDN IRL IRN ITA JOR JPN
## 3 4 4 4 1 1 5 3 3 2 4 5 3 5 1 3 5 5 1 5
## KAZ KOR KOS KWT LTU LVA MAR MDA MKD MLT MNE MNG NIR NLD NOR NOX NZL OMN PAK PHL
## 4 3 1 4 5 3 5 1 1 2 1 1 3 5 5 1 5 3 1 2
## POL PRT QAT RMO ROM RUS SAU SCO SGP SLV SRB SVK SVN SWE THA TUN TUR TWN UIN UKR
## 3 3 4 1 1 5 3 2 5 1 3 4 4 4 1 3 3 5 1 1
## UMA UMN USA YEM ZAF
## 1 1 5 3 2
```

```
## Number of studies
```

```
length(table(bflpedat$year))
```

```
## [1] 5
```

## 2 Meta-analytic baseline models

First, we evaluated the possible baseline models for their appropriateness to describe the meta-analytic data structure. The resultant baseline model formed the basis for further moderator analyses.

## 2.1 Common random-effects models (REM)

### 2.1.1 REM with variation between effect sizes

```
# Model specification
REM2e <- rma.mv(effect_size,
               var,
               random = list(~ 1 | ESID),
               data = bflpedat,
               method = "REML")

# Summarize the results
summary(REM2e, digits=4)

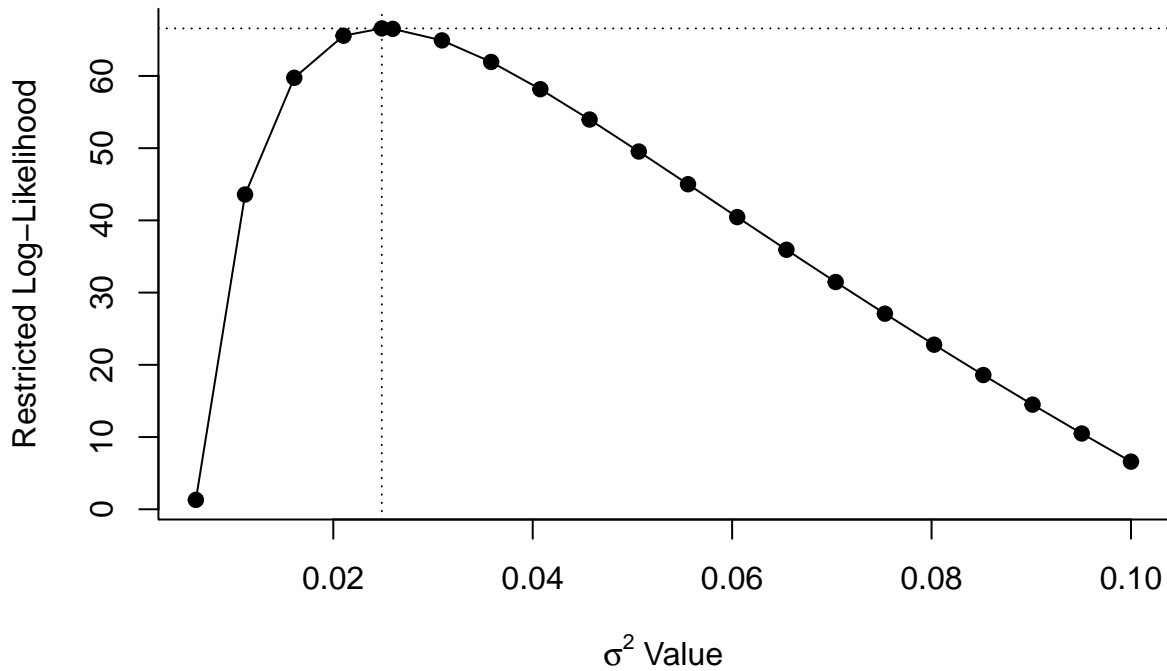
##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik   Deviance      AIC      BIC      AICc
## 66.5803 -133.1606 -129.1606 -122.1418 -129.1114
##
## Variance Components:
##
##      estim  sqrt  nlvls  fixed  factor
## sigma^2  0.0249  0.1577   248    no    ESID
##
## Test for Heterogeneity:
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4580  0.0114 -40.0998 <.0001 -0.4804 -0.4356 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Call for the 95% confidence intervals
confint(REM2e, digits=4)

##
##      estimate  ci.lb  ci.ub
## sigma^2  0.0249  0.0197  0.0314
## sigma    0.1577  0.1405  0.1773

# Profile plot
profile.rma.mv(REM2e, sigma2=1)
```

Profile Plot for  $\sigma^2$



*# Alternative specification*

```
rma(effect_size, var, data = bflpedat, method = "REML")
```

```
##
## Random-Effects Model (k = 248; tau^2 estimator: REML)
##
## tau^2 (estimated amount of total heterogeneity): 0.0249 (SE = 0.0029)
## tau (square root of estimated tau^2 value): 0.1577
## I^2 (total heterogeneity / total variability): 82.43%
## H^2 (total variability / sampling variability): 5.69
##
## Test for Heterogeneity:
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.4580  0.0114 -40.0997 <.0001 -0.4804 -0.4356 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

*# RVE standard errors*

```
conf_int(REM2e, vcov = "CR2")
```

```
##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt -0.458 0.0114 239      -0.481      -0.436
```

## 2.1.2 REM with variation between TIMSS cycles

```
# Model specification
REM2y <- rma.mv(effect_size,
               var,
               random = list(~ 1 | year),
               data = bflpedat,
               method = "REML")

# Summarize the results
summary(REM2y, digits=4)

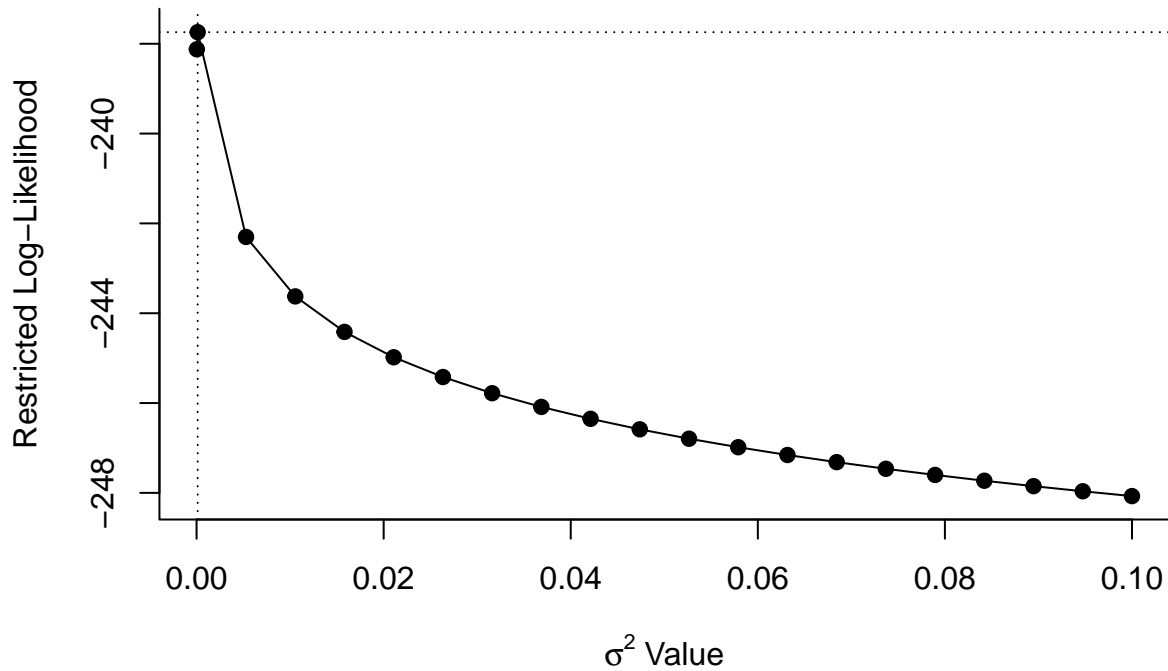
##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## -237.7424  475.4849  479.4849  486.5037  479.5341
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed  factor
## sigma^2    0.0001  0.0109     5     no   year
##
## Test for Heterogeneity:
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4412  0.0068  -64.7760 <.0001  -0.4545  -0.4278 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Call for the 95% confidence intervals
confint(REM2y, digits=4)

##
##           estimate  ci.lb  ci.ub
## sigma^2    0.0001  0.0000  0.0012
## sigma      0.0109  0.0000  0.0346

# Profile plot
profile.rma.mv(REM2y, sigma2=1)
```

## Profile Plot for $\sigma^2$



### 2.1.3 REM with variation between countries

```
# Model specification
REM2c <- rma.mv(effect_size,
               var,
               random = list(~ 1 | country),
               data = bflpedat,
               method = "REML")

# Summarize the results
summary(REM2c, digits=4)

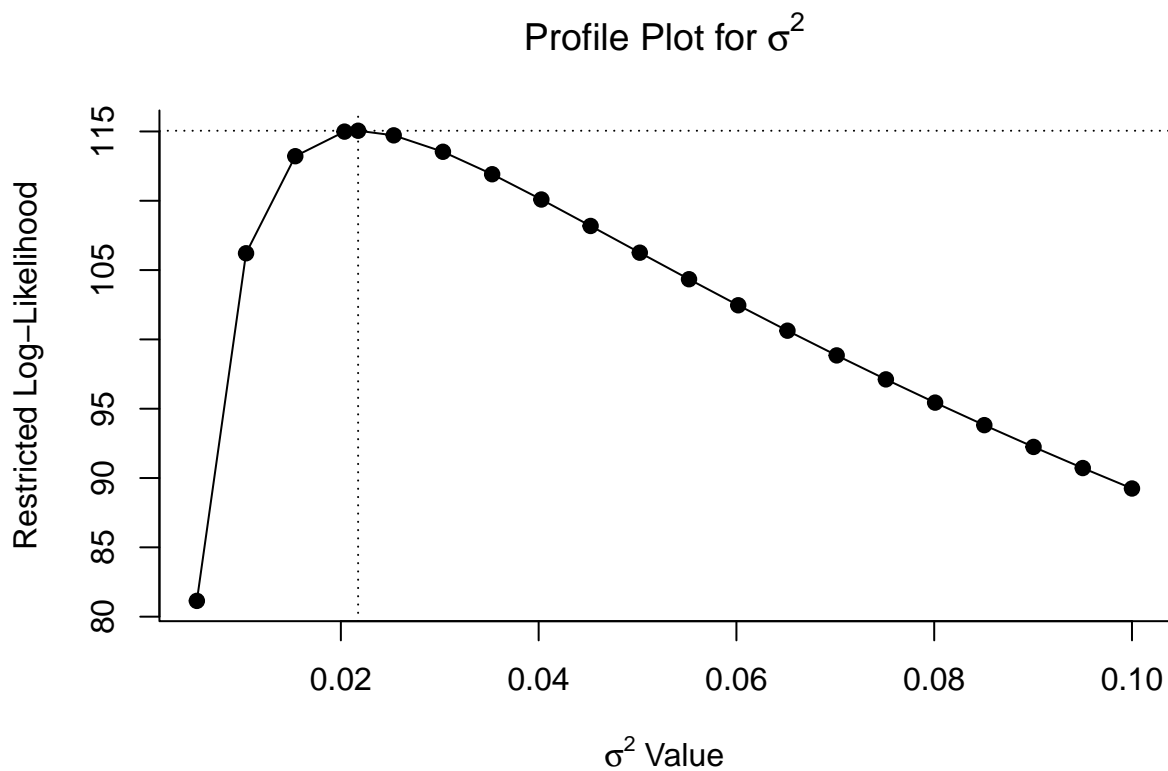
##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 115.0510 -230.1020 -226.1020 -219.0832 -226.0528
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed  factor
## sigma^2    0.0217 0.1475    85     no  country
##
## Test for Heterogeneity:
```

```
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.4482  0.0175  -25.6637 <.0001  -0.4824  -0.4140 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Call for the 95% confidence intervals
confint(REM2c, digits=4)
```

```
##
##      estimate  ci.lb  ci.ub
## sigma^2  0.0217 0.0153 0.0317
## sigma    0.1475 0.1236 0.1779
```

```
# Profile plot
profile.rma.mv(REM2c, sigma2=1)
```



#### 2.1.4 Fixed-effects model

```
# Model specification
FEM <- rma(effect_size, var, data = bflpedat, method = "FE")
```

```

# Summarize the results
summary(FEM, digits=4)

##
## Fixed-Effects Model (k = 248)
##
##   logLik   deviance      AIC      BIC      AICc
## -237.0284 1261.7650  476.0569  479.5703  476.0731
##
## I2 (total heterogeneity / total variability):  80.42%
## H2 (total variability / sampling variability):  5.11
##
## Test for Heterogeneity:
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4400  0.0046  -95.2527 <.0001  -0.4491  -0.4310 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

## 2.2 Multilevel meta-analyses with hierarchical effects

### 2.2.1 Three-level random-effects model with hierarchical nesting in TIMSS cycles

```

# Estimate the full model of the meta-analysis with three levels

# Define the levels of the meta-analysis
# Level 1: Sampling variation
# Level 2: Effect sizes (ESID)
# Level 3: TIMSS cycle (year)
REM3L <- rma.mv(effect_size,
               var,
               random = list(~ 1 | year/ESID),
               data = bflpedat,
               method = "REML")

# Summarize the results
summary(REM3L, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
##  66.5803 -133.1606 -127.1606 -116.6324 -127.0618

```

```

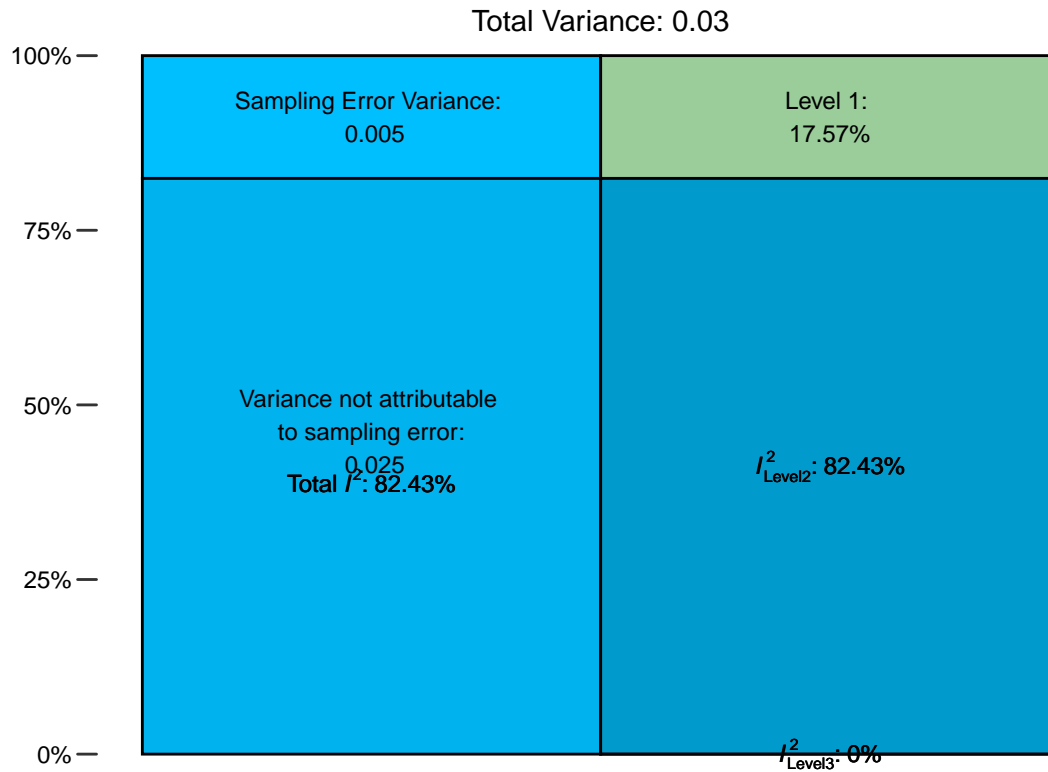
##
## Variance Components:
##
##          estim    sqrt  nlvls  fixed    factor
## sigma^2.1 0.0000  0.0000    5     no      year
## sigma^2.2 0.0249  0.1577   248    no    year/ESID
##
## Test for Heterogeneity:
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4580  0.0114  -40.0998 <.0001  -0.4804  -0.4356 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Variance distribution on levels
mlm.variance.distribution(x = REM3L)

##          % of total variance    I2
## Level 1          1.757307e+01    ---
## Level 2          8.242693e+01  82.43
## Level 3          2.086660e-08     0
## Total I2: 82.43%

res3L <- dmetar::var.comp(REM3L)
plot(res3L)

```



```
# Call for the 95% confidence intervals
confint(REM3L, digits=4)
```

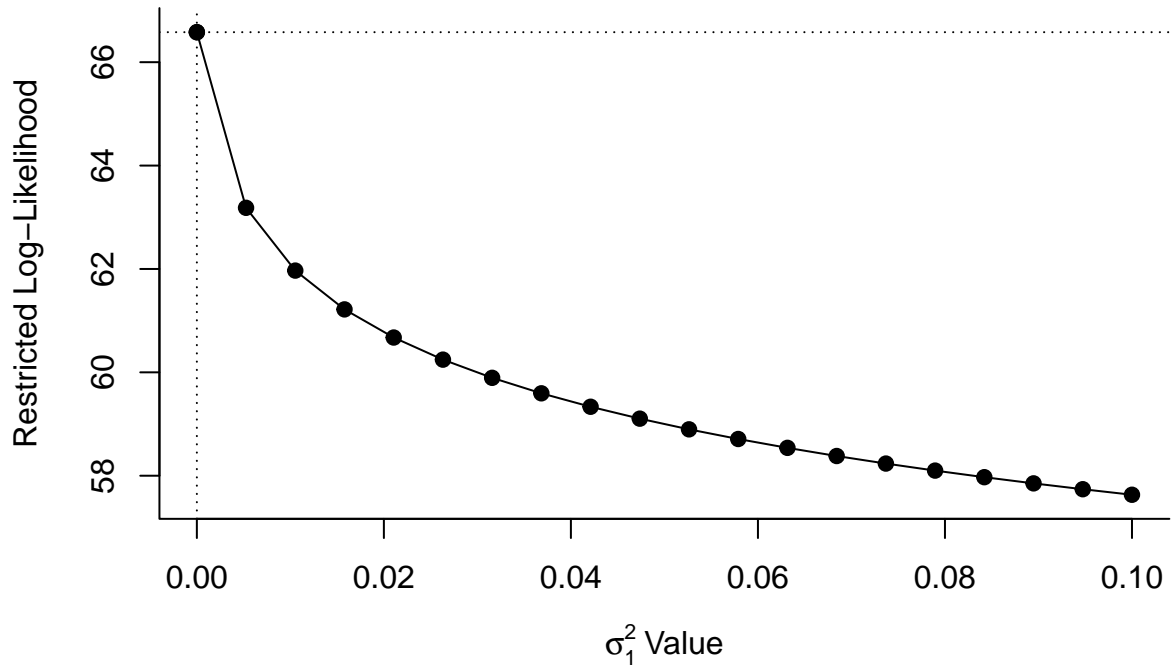
```
##
##          estimate ci.lb ci.ub
## sigma^2.1  0.0000 0.0000 0.0020
## sigma.1    0.0000 0.0000 0.0445
##
##          estimate ci.lb ci.ub
## sigma^2.2  0.0249 0.0197 0.0314
## sigma.2    0.1577 0.1405 0.1773
```

```
## RVE standard errors
conf_int(REM3L, vcov = "CR2")
```

```
##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt  -0.458 0.00903 3.66      -0.484      -0.432
```

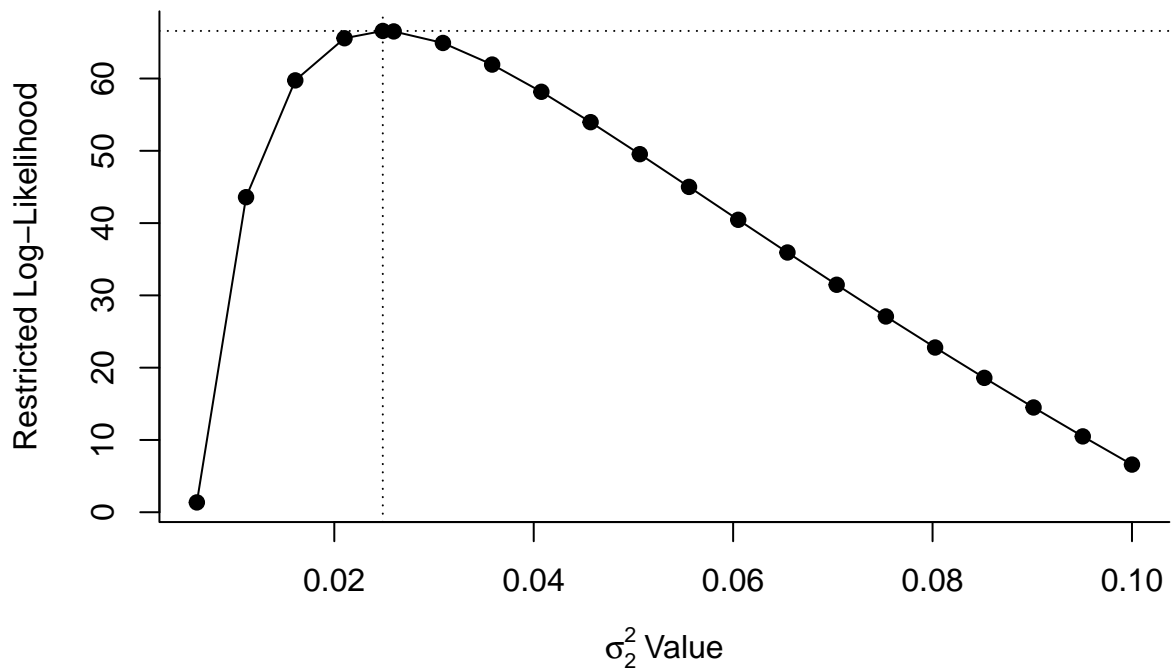
```
# Profile plots
profile.rma.mv(REM3L, sigma2=1)
```

Profile Plot for  $\sigma_1^2$



```
profile.rma.mv(REM3L,sigma2=2)
```

Profile Plot for  $\sigma_2^2$



```
# ICC[1]  
round(REM3L$sigma2[1] / sum(REM3L$sigma2), 3)
```

```

## [1] 0
round(REM3L$sigma2[2] / sum(REM3L$sigma2), 3)

## [1] 1
# Level-specific I2 values
# Source: https://www.metafor-project.org/doku.php/tips:i2_multilevel_multivariate
W <- diag(1/bflpedat$var)
X <- model.matrix(REM3L)
P <- W - W %*% X %*% solve(t(X) %*% W %*% X) %*% t(X) %*% W
## Overall I-squared
100 * sum(REM3L$sigma2) / (sum(REM3L$sigma2) + (REM3L$k-REM3L$p)/sum(diag(P)))

## [1] 82.42693
## I-squared for the different levels
100 * REM3L$sigma2 / (sum(REM3L$sigma2) + (REM3L$k-REM3L$p)/sum(diag(P)))

## [1] 2.086660e-08 8.242693e+01
## Model comparison
## Common REM2e to REM3L
anova(REM2e, REM3L)

##
##          df          AIC          BIC          AICc  logLik      LRT    pval      QE
## Full      3 -127.1606 -116.6324 -127.0618 66.5803                1261.7650
## Reduced  2 -129.1606 -122.1418 -129.1114 66.5803 0.0000 1.0000 1261.7650

```

## 2.2.2 Three-level random-effects model with hierarchical nesting in countries

```

# Estimate the full model of the meta-analysis with three levels

# Define the levels of the meta-analysis
# Level 1: Sampling variation
# Level 2: Effect sizes (ESID)
# Level 3: Countries
REM3Lc <- rma.mv(effect_size,
                var,
                random = list(~ 1 | country/ESID),
                data = bflpedat,
                method = "REML")

# Summarize the results
summary(REM3Lc, digits=4)

```

```

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##

```

```

##      logLik   Deviance      AIC      BIC      AICc
## 124.6689 -249.3379 -243.3379 -232.8097 -243.2391
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1 0.0206 0.1437    85    no      country
## sigma^2.2 0.0036 0.0597   248    no  country/ESID
##
## Test for Heterogeneity:
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4507 0.0176 -25.5958 <.0001 -0.4852 -0.4162 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

# Variance distribution on levels
mlm.variance.distribution(x = REM3Lc)

```

```

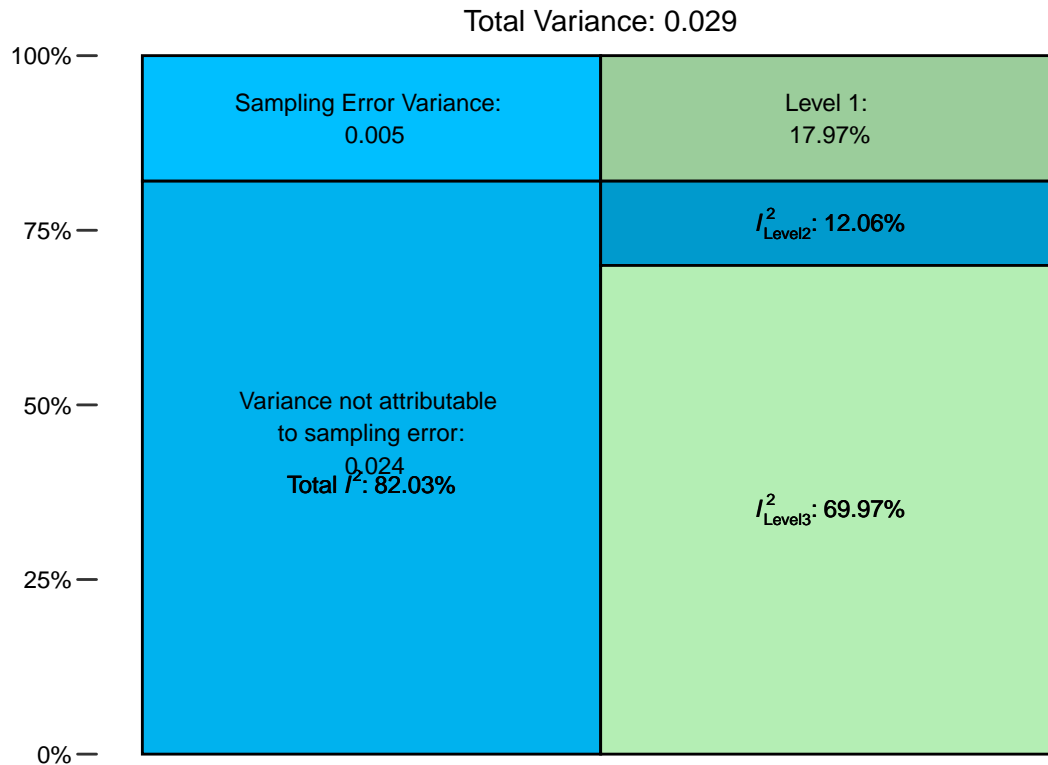
##      % of total variance    I2
## Level 1      17.96939    ---
## Level 2      12.06346  12.06
## Level 3      69.96715  69.97
## Total I2: 82.03%

```

```

res3Lc <- dmetar::var.comp(REM3Lc)
plot(res3Lc)

```



```
# Call for the 95% confidence intervals
confint(REM3Lc, digits=4)
```

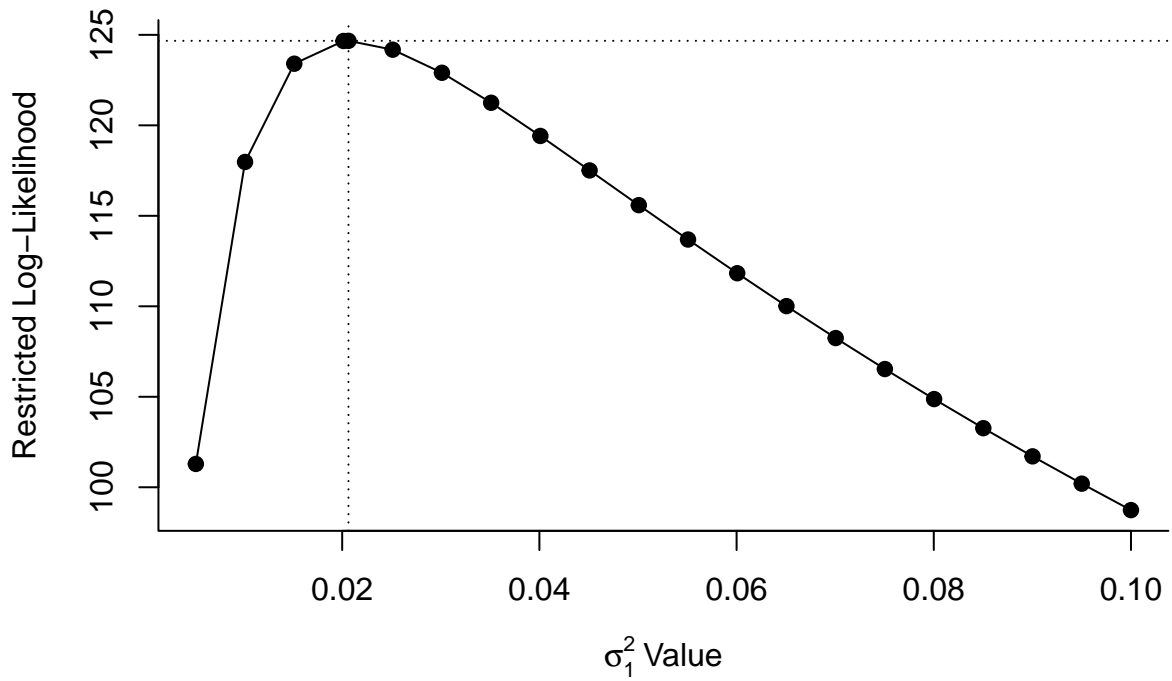
```
##
##          estimate  ci.lb  ci.ub
## sigma^2.1  0.0206 0.0141 0.0306
## sigma.1    0.1437 0.1188 0.1750
##
##          estimate  ci.lb  ci.ub
## sigma^2.2  0.0036 0.0017 0.0062
## sigma.2    0.0597 0.0408 0.0787
```

```
## RVE standard errors
conf_int(REM3Lc, vcov = "CR2")
```

```
##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt  -0.451 0.0176 80.5      -0.486      -0.416
```

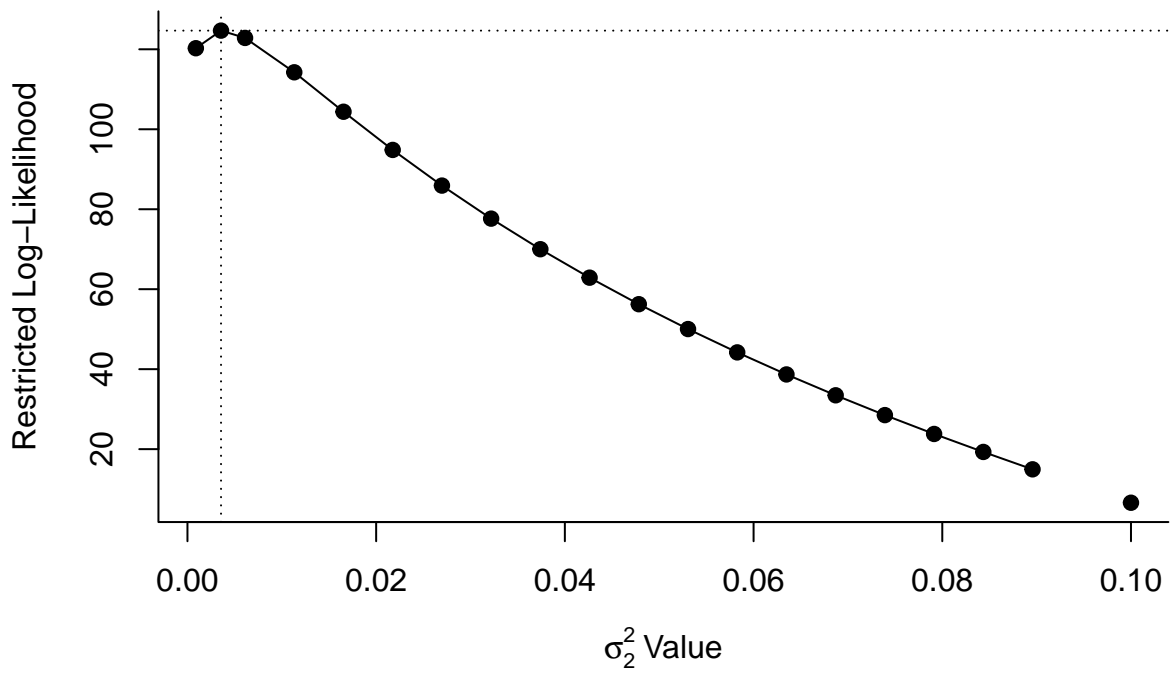
```
# Profile plots
profile.rma.mv(REM3Lc, sigma2=1)
```

Profile Plot for  $\sigma_1^2$



```
profile.rma.mv(REM3Lc, sigma2=2)
```

Profile Plot for  $\sigma_2^2$



```
# ICC[1]  
round(REM3Lc$sigma2[1] / sum(REM3Lc$sigma2), 3)
```

```

## [1] 0.853
round(REM3Lc$sigma2[2] / sum(REM3Lc$sigma2), 3)

## [1] 0.147
# Level-specific I2 values
# Source: https://www.metafor-project.org/doku.php/tips:i2_multilevel_multivariate
X1 <- model.matrix(REM3Lc)
P1 <- W - W %*% X1 %*% solve(t(X1) %*% W %*% X1) %*% t(X1) %*% W
## Overall I-squared
100 * sum(REM3Lc$sigma2) / (sum(REM3Lc$sigma2) + (REM3Lc$k-REM3Lc$p)/sum(diag(P1)))

## [1] 82.03061
## I-squared for the different levels
100 * REM3Lc$sigma2 / (sum(REM3Lc$sigma2) + (REM3Lc$k-REM3Lc$p)/sum(diag(P1)))

## [1] 69.96715 12.06346
## Model comparison
## Common REM2e to REM3L
anova(REM2e, REM3Lc)

##
##          df          AIC          BIC          AICc    logLik          LRT    pval          QE
## Full      3 -243.3379 -232.8097 -243.2391 124.6689                1261.7650
## Reduced   2 -129.1606 -122.1418 -129.1114  66.5803 116.1773 <.0001 1261.7650

```

### 2.2.3 Correlated and hierarchical effects model with RVE (CHE model)

```

#### For comparison: use cycles to describe correlated effects (CHE)
## Source code: Pustejovsky & Tipton (2021)
## https://doi.org/10.31222/osf.io/vyfcj

## Make a reasoned assumption on the correlations between effects.
## Effects over time are correlated within studies.
## We provide a constant sampling correlation (csc).

### RHO = 0.5 #####
rho <- 0.5

## Define the sampling variance-covariance matrix per study
V_mat <- impute_covariance_matrix(bflpedat$var,
                                cluster = bflpedat$country,
                                r = rho,
                                smooth_vi = TRUE)

## Estimate the full model of the meta-analysis with three levels

```

```

REM3L.csc <- rma.mv(effect_size,
                  V = V_mat,
                  random = list(~ 1 | country/ESID),
                  data = bflpedat,
                  method = "REML")

## Summarize the results
summary(REM3L.csc, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 123.1383 -246.2766 -240.2766 -229.7484 -240.1778
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed      factor
## sigma^2.1  0.0164  0.1282    85    no      country
## sigma^2.2  0.0071  0.0844   248    no  country/ESID
##
## Test for Heterogeneity:
## Q(df = 247) = 940.5129, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4555  0.0175  -26.1019 <.0001  -0.4897  -0.4213 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Confidence intervals
confint(REM3L.csc)

##
##           estimate  ci.lb  ci.ub
## sigma^2.1  0.0164  0.0102  0.0261
## sigma.1    0.1282  0.1008  0.1615
##
##           estimate  ci.lb  ci.ub
## sigma^2.2  0.0071  0.0050  0.0100
## sigma.2    0.0844  0.0707  0.1000

## RVE standard errors
clubSandwich::conf_int(REM3L.csc, vcov = "CR2")

##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt    -0.456 0.0175 79.2      -0.49      -0.421

```

```

# Level-specific I2 values
# Source: https://www.metafor-project.org/doku.php/tips:i2_multilevel_multivariate
X3 <- model.matrix(REM3L.csc)
P3 <- W - W %*% X3 %*% solve(t(X3) %*% W %*% X) %*% t(X3) %*% W
## Overall I-squared
100 * sum(REM3L.csc$sigma2) / (sum(REM3L.csc$sigma2) + (REM3L.csc$k-REM3L.csc$p)/sum(diag(P3)))

## [1] 81.63377

## I-squared for the different levels
100 * REM3L.csc$sigma2 / (sum(REM3L.csc$sigma2) + (REM3L.csc$k-REM3L.csc$p)/sum(diag(P3)))

## [1] 56.97930 24.65447

### RHO = 0.6 #####
## Define the sampling variance-covariance matrix per study
V_mat06 <- impute_covariance_matrix(bflpedat$var,
                                   cluster = bflpedat$country,
                                   r = 0.6,
                                   smooth_vi = TRUE)

## Estimate the full model of the meta-analysis with three levels
REM3L.csc06 <- rma.mv(effect_size,
                    V = V_mat06,
                    random = list(~ 1 | country/ESID),
                    data = bflpedat,
                    method = "REML")

## Summarize the results
summary(REM3L.csc06, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 122.0002 -244.0004 -238.0004 -227.4722 -237.9016
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed      factor
## sigma^2.1  0.0156  0.1249    85    no      country
## sigma^2.2  0.0082  0.0903   248    no  country/ESID
##
## Test for Heterogeneity:
## Q(df = 247) = 1029.7987, p-val < .0001
##
## Model Results:
##

```

```

## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4553  0.0174  -26.1003 <.0001  -0.4895  -0.4211  ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Confidence intervals
confint(REM3L.csc06)

##
##          estimate ci.lb ci.ub
## sigma^2.1  0.0156 0.0093 0.0252
## sigma.1    0.1249 0.0966 0.1588
##
##          estimate ci.lb ci.ub
## sigma^2.2  0.0082 0.0060 0.0111
## sigma.2    0.0903 0.0774 0.1053

## RVE standard errors
clubSandwich::conf_int(REM3L.csc06, vcov = "CR2")

##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt   -0.455 0.0174 78.9      -0.49      -0.421

### RHO = 0.7 #####
## Define the sampling variance-covariance matrix per study
V_mat07 <- impute_covariance_matrix(bflpedat$var,
                                   cluster = bflpedat$country,
                                   r = 0.7,
                                   smooth_vi = TRUE)

## Estimate the full model of the meta-analysis with three levels
REM3L.csc07 <- rma.mv(effect_size,
                    V = V_mat07,
                    random = list(~ 1 | country/ESID),
                    data = bflpedat,
                    method = "REML")

## Summarize the results
summary(REM3L.csc07, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik  Deviance      AIC      BIC      AICc
## 120.8722 -241.7443 -235.7443 -225.2162 -235.6456
##
## Variance Components:
##

```

```

##          estim    sqrt  nlvls  fixed      factor
## sigma^2.1 0.0148  0.1215    85     no      country
## sigma^2.2 0.0092  0.0961   248     no  country/ESID
##
## Test for Heterogeneity:
## Q(df = 247) = 1210.0182, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4551  0.0174  -26.0924 <.0001  -0.4892  -0.4209 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
## Confidence intervals
```

```
confint(REM3L.csc07)
```

```

##
##          estimate  ci.lb  ci.ub
## sigma^2.1  0.0148 0.0085 0.0243
## sigma.1     0.1215 0.0923 0.1560
##
##          estimate  ci.lb  ci.ub
## sigma^2.2  0.0092 0.0070 0.0122
## sigma.2     0.0961 0.0835 0.1106

```

```
## RVE standard errors
```

```
clubSandwich::conf_int(REM3L.csc07, vcov = "CR2")
```

```

##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt  -0.455 0.0174 78.6      -0.49      -0.42

```

```
### RHO = 0.8 #####
```

```
## Define the sampling variance-covariance matrix per study
```

```

V_mat08 <- impute_covariance_matrix(bflpedat$var,
                                   cluster = bflpedat$country,
                                   r = 0.8,
                                   smooth_vi = TRUE)

```

```
## Estimate the full model of the meta-analysis with three levels
```

```

REM3L.csc08 <- rma.mv(effect_size,
                    V = V_mat08,
                    random = list(~ 1 | country/ESID),
                    data = bflpedat,
                    method = "REML")

```

```
## Summarize the results
```

```
summary(REM3L.csc08, digits=4)
```

```
##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik   Deviance      AIC      BIC      AICc
## 119.7773 -239.5546 -233.5546 -223.0264 -233.4558
##
## Variance Components:
##
##      estim  sqrt  nlvls  fixed  factor
## sigma^2.1 0.0139 0.1181   85    no    country
## sigma^2.2 0.0103 0.1017  248    no  country/ESID
##
## Test for Heterogeneity:
## Q(df = 247) = 1605.5513, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.4548  0.0174  -26.0788  <.0001  -0.4890  -0.4206  ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
## Confidence intervals
```

```
confint(REM3L.csc08)
```

```
##
##      estimate  ci.lb  ci.ub
## sigma^2.1    0.0139 0.0077 0.0235
## sigma.1      0.1181 0.0879 0.1533
##
##      estimate  ci.lb  ci.ub
## sigma^2.2    0.0103 0.0080 0.0134
## sigma.2      0.1017 0.0897 0.1158
```

```
## RVE standard errors
```

```
clubSandwich::conf_int(REM3L.csc08, vcov = "CR2")
```

```
##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt    -0.455 0.0174 78.3      -0.49      -0.42
```

#### 2.2.4 Four-level cross-classified random-effects model with hierarchical clustering in TIMSS cycles and countries

This model assumes a hierarchical nesting of effect sizes in countries and TIMSS cycles; however, the country and cycle levels are independent.

```
## Cross-classified REM
```

```
CCREM4 <- rma.mv(effect_size,
```

```

var,
random = list(~ 1 | country/ESID,
              ~ 1 | year),
data = bflpedat,
method = "REML")

## Summarize the results
summary(CCREM4, digits=4)

```

```

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik   Deviance      AIC      BIC      AICc
## 125.5016 -251.0032 -243.0032 -228.9656 -242.8379
##
## Variance Components:
##
##      estim  sqrt  nlvls  fixed  factor
## sigma^2.1 0.0209 0.1447   85    no    country
## sigma^2.2 0.0032 0.0568  248    no  country/ESID
## sigma^2.3 0.0002 0.0144    5    no    year
##
## Test for Heterogeneity:
## Q(df = 247) = 1261.7650, p-val < .0001
##
## Model Results:
##
## estimate      se      zval      pval      ci.lb      ci.ub
## -0.4524 0.0189 -23.9696 <.0001 -0.4893 -0.4154 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

# Call for the 95% confidence intervals
confint(CCREM4, digits=4)

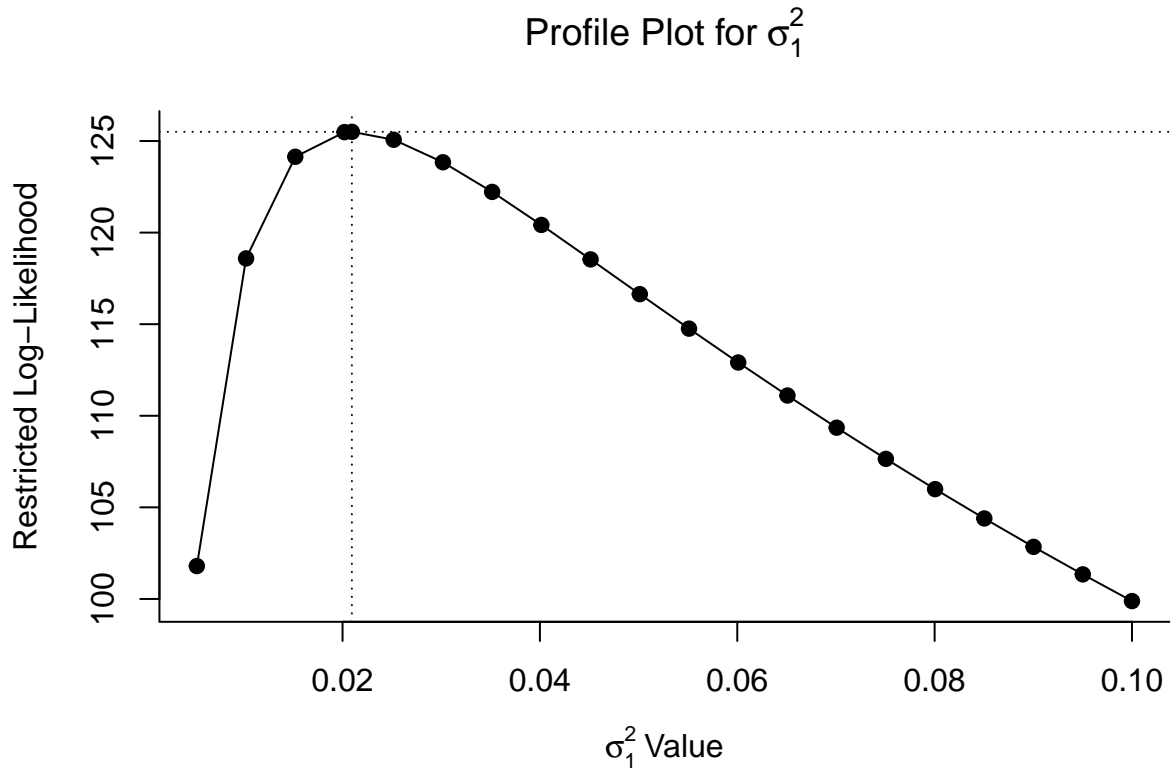
```

```

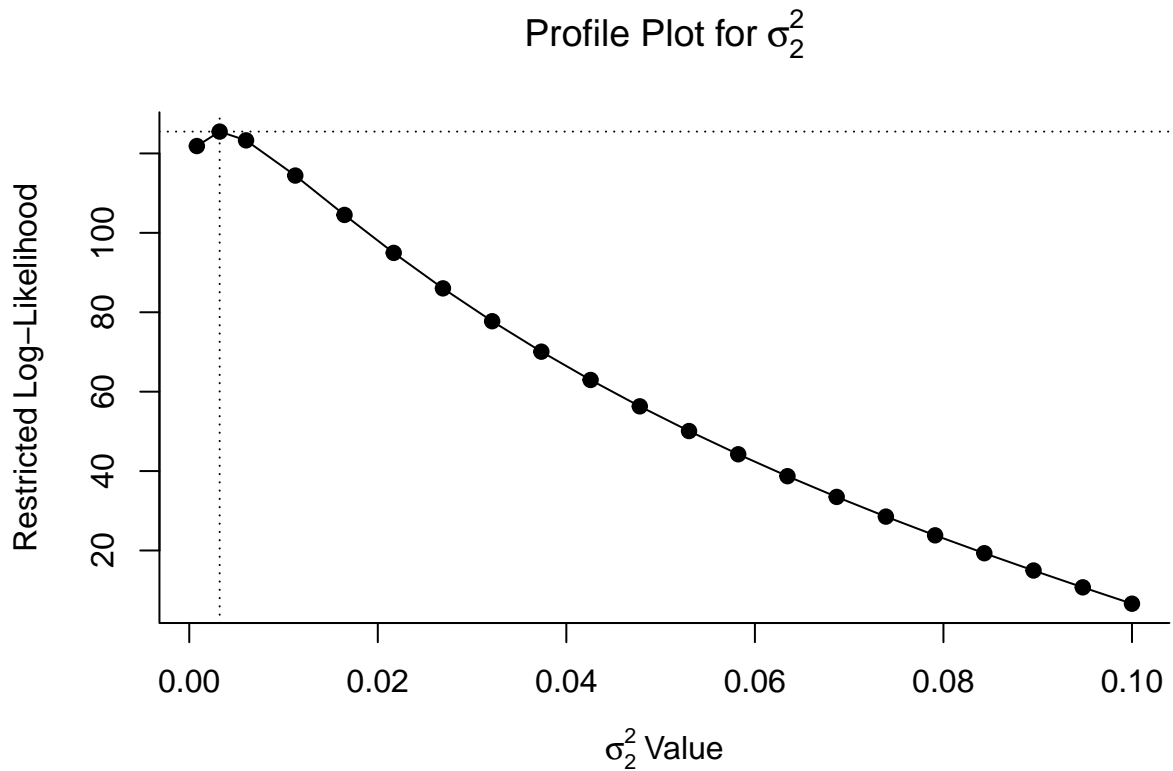
##
##      estimate  ci.lb  ci.ub
## sigma^2.1  0.0209 0.0144 0.0310
## sigma.1    0.1447 0.1198 0.1762
##
##      estimate  ci.lb  ci.ub
## sigma^2.2  0.0032 0.0014 0.0058
## sigma.2    0.0568 0.0370 0.0763
##
##      estimate  ci.lb  ci.ub
## sigma^2.3  0.0002 0.0000 0.0021
## sigma.3    0.0144 0.0000 0.0460

```

```
# Profile plots
profile.rma.mv(CCREM4, sigma2=1)
```

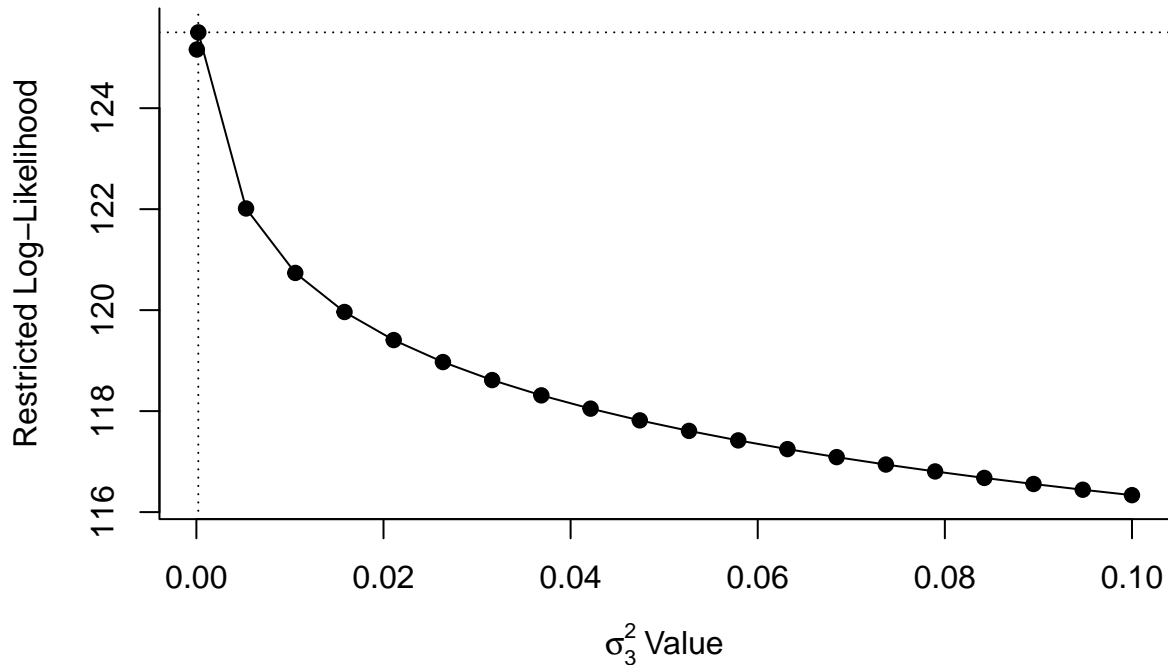


```
profile.rma.mv(CCREM4, sigma2=2)
```



```
profile.rma.mv(CCREM4, sigma2=3)
```

### Profile Plot for $\sigma_3^2$



```
# ICC[1]
round(CCREM4$sigma2[1] / sum(CCREM4$sigma2), 3)

## [1] 0.859

round(CCREM4$sigma2[2] / sum(CCREM4$sigma2), 3)

## [1] 0.132

round(CCREM4$sigma2[3] / sum(CCREM4$sigma2), 3)

## [1] 0.008

# Level-specific I2 values
X2 <- model.matrix(CCREM4)
P2 <- W - W %*% X2 %*% solve(t(X2) %*% W %*% X2) %*% t(X2) %*% W
## Overall I-squared
100 * sum(CCREM4$sigma2) / (sum(CCREM4$sigma2) + (CCREM4$k-CCREM4$p)/sum(diag(P2)))

## [1] 82.14006

## I-squared for the different levels
100 * CCREM4$sigma2 / (sum(CCREM4$sigma2) + (CCREM4$k-CCREM4$p)/sum(diag(P2)))

## [1] 70.5762190 10.8686506 0.6951873
```

```

## Model comparisons
## Common REM2e to CCREM4
anova(REM2e, CCREM4)

##
##          df          AIC          BIC          AICc    logLik          LRT    pval          QE
## Full      4 -243.0032 -228.9656 -242.8379 125.5016                1261.7650
## Reduced   2 -129.1606 -122.1418 -129.1114  66.5803 117.8426 <.0001 1261.7650

## REM3L vs. CCREM4
anova(REM3L, CCREM4)

##
##          df          AIC          BIC          AICc    logLik          LRT    pval          QE
## Full      4 -243.0032 -228.9656 -242.8379 125.5016                1261.7650
## Reduced   3 -127.1606 -116.6324 -127.0618  66.5803 117.8426 <.0001 1261.7650

## REM3Lc vs. CCREM4
anova(REM3Lc, CCREM4)

##
##          df          AIC          BIC          AICc    logLik          LRT    pval          QE
## Full      4 -243.0032 -228.9656 -242.8379 125.5016                1261.7650
## Reduced   3 -243.3379 -232.8097 -243.2391 124.6689  1.6653 0.1969 1261.7650

```

The variance between TIMSS cycles is close to zero, the corresponding confidence interval contains zero, the profile plot supports the zero variance, and the model CCREM4 does not improve the fit of the meta-analytic baseline model . This could indicate some degree of stability of the BFLPE over time.

## 2.3 Robust variance estimation (RVE) with hierarchical effects

### 2.3.1 RVE with hierarchical effects within countries

```

## Model specification and estimation
REM2.rvec <- robu(bflpedat$effect_size ~ 1,
                 data = bflpedat,
                 studynum = bflpedat$country,
                 var.eff.size = bflpedat$var,
                 modelweights = "HIER")

## Summary of the results
print(REM2.rvec)

## RVE: Hierarchical Effects Model with Small-Sample Corrections
##
## Model: bflpedat$effect_size ~ 1
##
## Number of clusters = 85

```

```

## Number of outcomes = 248 (min = 1 , mean = 2.92 , median = 3 , max = 5 )
## Omega.sq = 0
## Tau.sq = 0.02208545
##
##           Estimate StdErr t-value  dfs P(|t|>) 95% CI.L 95% CI.U Sig
## 1 X.Intercept.  -0.458 0.0192  -23.9 62.8      0  -0.496  -0.419 ***
## ---
## Signif. codes: < .01 *** < .05 ** < .10 *
## ---
## Note: If df < 4, do not trust the results

```

### 2.3.2 RVE with correlated effects in TIMSS cycles

```

## Model specification and estimation
REM2.rvet <- robu(bflpedat$effect_size ~ 1,
                 data = bflpedat,
                 studynum = bflpedat$year,
                 var.eff.size = bflpedat$var,
                 modelweights = "COR",
                 rho = 0.5)

## Summary of the results
print(REM2.rvet)

## RVE: Correlated Effects Model with Small-Sample Corrections
##
## Model: bflpedat$effect_size ~ 1
##
## Number of studies = 5
## Number of outcomes = 248 (min = 29 , mean = 49.6 , median = 55 , max = 64 )
## Rho = 0.5
## I.sq = 76.12715
## Tau.sq = 0.03544564
##
##           Estimate StdErr t-value  dfs      P(|t|>) 95% CI.L 95% CI.U Sig
## 1 X.Intercept.  -0.458 0.00879  -52.1 3.95 0.000000935  -0.483  -0.434 ***
## ---
## Signif. codes: < .01 *** < .05 ** < .10 *
## ---
## Note: If df < 4, do not trust the results

## Sensitivity analysis
robumeta::sensitivity(REM2.rvet)

## RVE: Correlated Effects Model with Small-Sample Corrections
## Model: bflpedat$effect_size ~ 1
##
## Sensitivity Analysis

```

```

##
##                               Rho = 0  Rho = 0.2  Rho = 0.4  Rho = 0.6  Rho = 0.8
## X.Intercept. Coefficient -0.45812 -0.45813 -0.45814 -0.45815 -0.45816
##                               Std. Error  0.00878  0.00879   0.00879   0.00879   0.00879
## Tau.sq      Estimate      0.03423  0.03472   0.03520   0.03569   0.03617
## Rho = 1
## -0.4582
##  0.0088
##  0.0367

```

## 2.4 Conclusions

Some highlights on the results so far:

- Variation exists between effect sizes and countries. A three-level hierarchical structure with the country level as the highest level is indicated.
- Adding the variation between the five TIMSS cycles (model CCREM4) did not result in an improved model fit (as compared to model REM3Lc). Hence, the country differences are key sources of variation in addition to the variation between effect sizes.

## 3 Moderator analyses

### 3.1 Moderation by TIMSS cycle

The model `CCREM4` already indicated a small variation between the TIMSS cycles, after controlling for variation between effect sizes and countries. Alternatively, we may refrain from adding a random component that represents the TIMSS cycles and add smart-coded variables as moderators (resulting in fixed effects) instead.

#### 3.1.1 Three-level mixed-effects meta-regression model

```
# Model specification
MEM3L.cycle <- rma.mv(effect_size ~ 1 + factor(year),
                      var,
                      random = list(~ 1 | country/ESID),
                      data = bflpedat,
                      method = "REML")

# Summarize the results
summary(MEM3L.cycle, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik   Deviance      AIC      BIC      AICc
## 123.1370 -246.2739 -232.2739 -207.8225 -231.7973
##
## Variance Components:
##
##      estim  sqrt  nlvls  fixed      factor
## sigma^2.1 0.0211 0.1453   85    no      country
## sigma^2.2 0.0033 0.0572  248    no  country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 243) = 1253.3797, p-val < .0001
##
## Test of Moderators (coefficients 2:5):
## QM(df = 4) = 7.7450, p-val = 0.1014
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt      -0.4605 0.0262 -17.5640 <.0001 -0.5119 -0.4091 ***
## factor(year)2007 -0.0002 0.0258 -0.0095 0.9925 -0.0507 0.0502
## factor(year)2011 0.0024 0.0247 0.0966 0.9230 -0.0460 0.0508
## factor(year)2015 -0.0076 0.0242 -0.3129 0.7544 -0.0551 0.0399
```

```
## factor(year)2019    0.0374  0.0238    1.5674  0.1170  -0.0094  0.0841
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Robust confidence intervals
conf_int(MEM3L.cycle, vcov = "CR2")
```

```
##              Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1          intrcpt -0.460518 0.0280 40.0    -0.5172    -0.4039
## 2 factor(year)2007 -0.000244 0.0199 27.8    -0.0410     0.0405
## 3 factor(year)2011  0.002385 0.0236 29.7    -0.0458     0.0506
## 4 factor(year)2015 -0.007583 0.0243 28.6    -0.0572     0.0421
## 5 factor(year)2019  0.037358 0.0258 28.5    -0.0156     0.0903
```

```
#### For comparison: use time as a predictor
```

```
## Define a new time variable
```

```
bflpedat$time <- NA
bflpedat$time[bflpedat$StudyID == "2003"] <- 0
bflpedat$time[bflpedat$StudyID == "2007"] <- 1
bflpedat$time[bflpedat$StudyID == "2011"] <- 2
bflpedat$time[bflpedat$StudyID == "2015"] <- 3
bflpedat$time[bflpedat$StudyID == "2019"] <- 4
bflpedat$timesq <- NA
bflpedat$timesq <- bflpedat$time^2
```

```
# Three-level mixed-effects meta-regression model
```

```
MEM3L.time <- rma.mv(effect_size ~ 1 + time,
                    var,
                    random = list(~ 1 | country/ESID),
                    data = bflpedat,
                    method = "REML")
```

```
# Summarize the results
```

```
summary(MEM3L.time, digits=4)
```

```
##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##   logLik  Deviance      AIC      BIC      AICc
## 124.8831 -249.7661 -241.7661 -227.7448 -241.6002
##
## Variance Components:
##
##      estim  sqrt  nlvls  fixed  factor
## sigma^2.1 0.0208 0.1442   85    no    country
## sigma^2.2 0.0035 0.0589  248    no  country/ESID
##
## Test for Residual Heterogeneity:
```

```

## QE(df = 246) = 1259.3479, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 3.0316, p-val = 0.0817
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt  -0.4729  0.0218 -21.6867 <.0001 -0.5157 -0.4302 ***
## time      0.0090  0.0052   1.7411  0.0817 -0.0011  0.0191 .
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Robust confidence intervals
conf_int(MEM3L.time, vcov = "CR2")

##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt -0.47294 0.02747 63.9      -0.52781      -0.4181
## 2 time     0.00899 0.00686 39.1      -0.00488      0.0229

# Predict the effects for each cycle
predict(MEM3L.time, newmods = c(0,1,2,3,4))

##
##      pred      se      ci.lb      ci.ub      pi.lb      pi.ub
## 1 -0.4729 0.0218 -0.5157 -0.4302 -0.7813 -0.1646
## 2 -0.4639 0.0192 -0.5016 -0.4262 -0.7716 -0.1563
## 3 -0.4550 0.0178 -0.4899 -0.4200 -0.7623 -0.1476
## 4 -0.4460 0.0179 -0.4810 -0.4110 -0.7533 -0.1386
## 5 -0.4370 0.0193 -0.4748 -0.3991 -0.7447 -0.1293

```

### 3.1.2 Hierarchical effects model with RVE (HE model)

```

#### For comparison: RVE
MEMRVE.cycle <- robu(bflpedat$effect_size ~ 1 + factor(year),
                    data = bflpedat,
                    studynum = bflpedat$country,
                    var.eff.size = bflpedat$var,
                    modelweights = "HIER")

## Summary of the results
print(MEMRVE.cycle)

## RVE: Hierarchical Effects Model with Small-Sample Corrections
##
## Model: bflpedat$effect_size ~ 1 + factor(year)
##
## Number of clusters = 85

```

```

## Number of outcomes = 248 (min = 1 , mean = 2.92 , median = 3 , max = 5 )
## Omega.sq = 0
## Tau.sq = 0.02216272
##
##              Estimate StdErr  t-value  dfs          P(|t|>) 95% CI.L
## 1      X.Intercept. -0.45661 0.0356 -12.8102 26.6 0.00000000000000689 -0.5298
## 2 factor.year.2007 -0.02336 0.0273  -0.8561 36.1 0.397607012405788 -0.0787
## 3 factor.year.2011 -0.00156 0.0324  -0.0482 39.3 0.961820063218260 -0.0671
## 4 factor.year.2015 -0.01243 0.0337  -0.3687 38.8 0.714328479376139 -0.0806
## 5 factor.year.2019  0.02242 0.0351   0.6395 38.7 0.526237524878630 -0.0485
## 95% CI.U Sig
## 1  -0.3834 ***
## 2   0.0320
## 3   0.0640
## 4   0.0558
## 5   0.0933
## ---
## Signif. codes: < .01 *** < .05 ** < .10 *
## ---
## Note: If df < 4, do not trust the results

```

### 3.1.3 Subgroup correlated effects model (SCE model)

```

#### For comparison: use cycles as subgroups
## Define the sampling variance-covariance matrix per study
V_mats <- impute_covariance_matrix(bflpedat$var,
                                   cluster = bflpedat$country,
                                   r = rho,
                                   smooth_vi = TRUE,
                                   subgroup = bflpedat$StudyID)

## Estimate the full model of the meta-analysis with three levels
REM3L.cscs <- rma.mv(effect_size,
                    V = V_mats,
                    random = list(~ as.factor(StudyID) | country),
                    struct = "DIAG",
                    data = bflpedat,
                    method = "REML")

## Summarize the results
summary(REM3L.cscs, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik  Deviance      AIC      BIC      AICc
##      69.5452 -139.0905 -127.0905 -106.0341 -126.7405

```

```
##
## Variance Components:
##
## outer factor: country          (nlvls = 85)
## inner factor: as.factor(StudyID) (nlvls = 5)
##
##          estim    sqrt  k.lvl  fixed  level
## tau^2.1  0.0269  0.1639   29    no   2003
## tau^2.2  0.0323  0.1797   44    no   2007
## tau^2.3  0.0231  0.1519   55    no   2011
## tau^2.4  0.0290  0.1704   56    no   2015
## tau^2.5  0.0156  0.1249   64    no   2019
##
## Test for Heterogeneity:
## Q(df = 247) = 1195.1589, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4566  0.0112  -40.9372 <.0001  -0.4785  -0.4348 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## RVE standard errors
clubSandwich::conf_int(REM3L.cscs, vcov = "CR2")
```

```
##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt    -0.457 0.0186 63.7      -0.494      -0.419
```

Once again, no further indication of significant variation across cycles.

## 3.2 Moderation by cultural orientation (cycle-invariant, country-specific)

### 3.2.1 Three-level mixed-effects meta-regression model

```
# Model specification
MEM3L.culture <- rma.mv(effect_size ~ 1 +
                        pdi + idv + mas +
                        uai + ltowvs + ivr,
                        var,
                        random = list(~ 1 | country/ESID),
                        data = bflpedat, # could also use bflpedat2 (same result)
                        method = "REML")
```

```
## Warning: Rows with NAs omitted from model fitting.
```

```
# Summarize the results
summary(MEM3L.culture, digits=4)
```

```

##
## Multivariate Meta-Analysis Model (k = 206; method: REML)
##
##      logLik      Deviance      AIC      BIC      AICc
## 117.9376 -235.8753 -217.8753 -188.2355 -216.9229
##
## Variance Components:
##
##      estim      sqrt  nlvls  fixed      factor
## sigma^2.1 0.0181 0.1346    65    no      country
## sigma^2.2 0.0035 0.0589   206    no  country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 199) = 917.6139, p-val < .0001
##
## Test of Moderators (coefficients 2:7):
## QM(df = 6) = 8.5950, p-val = 0.1977
##
## Model Results:
##
##      estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt -0.1375 0.1804 -0.7620 0.4461 -0.4911 0.2162
## pdi      -0.0024 0.0015 -1.6075 0.1080 -0.0053 0.0005
## idv      -0.0019 0.0012 -1.5869 0.1125 -0.0042 0.0004
## mas      -0.0008 0.0010 -0.8251 0.4093 -0.0027 0.0011
## uai      -0.0010 0.0010 -1.0208 0.3074 -0.0030 0.0010
## ltowvs   0.0003 0.0009 0.2818 0.7781 -0.0015 0.0020
## ivr      0.0004 0.0014 0.2599 0.7949 -0.0024 0.0031
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

*# Robust CIs*

```
conf_int(MEM3L.culture, vcov = "CR2")
```

```

##      Coef Estimate      SE  d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt -0.137489 0.174135 17.37   -0.50429   0.229315
## 2 pdi      -0.002374 0.001550 14.16   -0.00569   0.000947
## 3 idv      -0.001885 0.001240 22.78   -0.00445   0.000681
## 4 mas      -0.000802 0.001138 15.16   -0.00323   0.001622
## 5 uai      -0.001038 0.000899 9.49    -0.00306   0.000979
## 6 ltowvs   0.000256 0.000975 22.82   -0.00176   0.002273
## 7 ivr      0.000369 0.001621 18.53   -0.00303   0.003768

```

*# Variance inflation factors*

```
vif(MEM3L.culture)
```

```

##
##      pdi      idv      mas      uai ltowvs      ivr

```

```

## 2.8286 2.1597 1.0858 1.3574 1.2558 2.0918
#### Dealing with missing data in the moderators (listwise)
## Exclude effect sizes with missing data
bflpedat2 <- bflpedat[complete.cases(data.frame(bflpedat$pdv,
                                              bflpedat$idv,
                                              bflpedat$mas,
                                              bflpedat$uai,
                                              bflpedat$ltowvs,
                                              bflpedat$ivr)),]

## Re-estimate the model without the predictors to get accurate R2 values
REM3L.culture <- rma.mv(effect_size ~ 1 ,
                      var,
                      random = list(~ 1 | country/ESID),
                      data = bflpedat2,
                      method = "REML")

# Variance explanations
# Country level
(REM3L.culture$sigma2[1]-MEM3L.culture$sigma2[1])*100/REM3L.culture$sigma2[1]

## [1] 5.450629

# Effect size level
(REM3L.culture$sigma2[2]-MEM3L.culture$sigma2[2])*100/REM3L.culture$sigma2[2]

## [1] -1.057425

```

### 3.2.2 Three-level mixed-effects meta-regression model with constant sampling correlation

```

#### Dealing with missing data in the moderators (listwise)

## Generate a new sampling covariance matrix
## Define the sampling variance-covariance matrix per study
V_mat2 <- impute_covariance_matrix(bflpedat2$var,
                                  cluster = bflpedat2$country,
                                  r = rho,
                                  smooth_vi = TRUE)

## Re-estimate the model without the predictors to get accurate R2 values
REM3L.csc.culture <- rma.mv(effect_size,
                          V = V_mat2,
                          random = list(~ 1 | country/ESID),
                          data = bflpedat2,
                          method = "REML")

```

```

## Summarize the results
summary(REM3L.csc.culture, digits=4)

##
## Multivariate Meta-Analysis Model (k = 206; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 114.9548 -229.9097 -223.9097 -213.9407 -223.7903
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed      factor
## sigma^2.1  0.0162  0.1272    65     no      country
## sigma^2.2  0.0065  0.0804   206     no  country/ESID
##
## Test for Heterogeneity:
## Q(df = 205) = 786.9486, p-val < .0001
##
## Model Results:
##
## estimate      se      zval    pval    ci.lb    ci.ub
## -0.4530  0.0193  -23.4968  <.0001  -0.4908  -0.4152  ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

#### Mixed-effects meta-regression model
# Model specification
MEM3L.csc.culture <- rma.mv(effect_size ~ 1 +
                             pdi + idv + mas +
                             uai + ltowvs + ivr,
                             V = V_mat2,
                             random = list(~ 1 | country/ESID),
                             data = bflpedat2,
                             method = "REML")

```

```

# Summarize the results
summary(MEM3L.csc.culture, digits=4)

```

```

##
## Multivariate Meta-Analysis Model (k = 206; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 117.1726 -234.3452 -216.3452 -186.7055 -215.3928
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed      factor

```

```

## sigma^2.1  0.0150  0.1227    65    no    country
## sigma^2.2  0.0065  0.0805   206    no  country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 199) = 726.9204, p-val < .0001
##
## Test of Moderators (coefficients 2:7):
## QM(df = 6) = 8.9219, p-val = 0.1780
##
## Model Results:
##
##          estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt  -0.1658  0.1824  -0.9092  0.3633  -0.5233  0.1917
## pdi      -0.0024  0.0015  -1.5923  0.1113  -0.0053  0.0005
## idv      -0.0019  0.0012  -1.5831  0.1134  -0.0042  0.0004
## mas      -0.0008  0.0010  -0.8186  0.4130  -0.0027  0.0011
## uai      -0.0010  0.0010  -0.9709  0.3316  -0.0030  0.0010
## ltowvs   0.0005  0.0009   0.4943  0.6211  -0.0014  0.0023
## ivr      0.0006  0.0014   0.4182  0.6758  -0.0022  0.0034
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Robust CIs
conf_int(MEM3L.csc.culture, vcov = "CR2")

##      Coef Estimate      SE  d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt -0.165834 0.176671 17.00   -0.53858    0.206912
## 2 pdi      -0.002379 0.001563 14.21   -0.00573    0.000968
## 3 idv      -0.001871 0.001242 22.44   -0.00444    0.000701
## 4 mas      -0.000797 0.001168 15.14   -0.00328    0.001692
## 5 uai      -0.000996 0.000899  9.07   -0.00303    0.001034
## 6 ltowvs   0.000460 0.000991 23.08   -0.00159    0.002510
## 7 ivr      0.000598 0.001633 18.46   -0.00283    0.004022

# Variance inflation factors
vif(MEM3L.csc.culture)

##
## pdi idv mas uai ltowvs ivr
## 2.8254 2.1415 1.0708 1.3850 1.2861 2.1093

# Variance explanations
# Country level
(REM3L.csc.culture$sigma2[1]-MEM3L.csc.culture$sigma2[1])*100/REM3L.csc.culture$sigma2[1]

## [1] 6.946436

# Effect size level
(REM3L.csc.culture$sigma2[2]-MEM3L.csc.culture$sigma2[2])*100/REM3L.csc.culture$sigma2[2]

```

```
## [1] -0.1059797
```

### 3.3 Moderation by economic development (cycle-specific, country-specific)

#### 3.3.1 Three-level mixed-effects meta-regression model

```
# Model specification
MEM3L.hdi <- rma.mv(effect_size ~ 1 + HDI,
                    var,
                    random = list(~ 1 | country/ESID),
                    data = bflpedat,
                    method = "REML")

## Warning: Rows with NAs omitted from model fitting.

# Summarize the results
summary(MEM3L.hdi, digits=4)

##
## Multivariate Meta-Analysis Model (k = 242; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 121.0388 -242.0776 -234.0776 -220.1551 -233.9074
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed      factor
## sigma^2.1  0.0198  0.1406    83    no      country
## sigma^2.2  0.0036  0.0599   242    no  country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 240) = 1140.0634, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 5.4778, p-val = 0.0193
##
## Model Results:
##
##           estimate      se    zval    pval    ci.lb    ci.ub
## intrcpt  -0.8121  0.1545  -5.2579 <.0001  -1.1148  -0.5094 ***
## HDI       0.4256  0.1818   2.3405  0.0193   0.0692   0.7820 *
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Robust CIs
conf_int(MEM3L.hdi, vcov = "CR2")

##           Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
```

```
## 1 intrcpt -0.812 0.225 36.1 -1.2688 -0.355
## 2 HDI 0.426 0.258 39.1 -0.0955 0.947

# Variance explanations
# Country level
(REM3Lc$sigma2[1]-MEM3L.hdi$sigma2[1])*100/REM3Lc$sigma2[1]

## [1] 4.258563

# Effect size level
(REM3Lc$sigma2[2]-MEM3L.hdi$sigma2[2])*100/REM3Lc$sigma2[2]

## [1] -0.9532245

## Note: No missing values occurred in the HDI.
```

### 3.3.2 Three-level mixed-effects meta-regression model with constant sampling correlation

```
# Model specification
MEM3L.hdi.csc <- rma.mv(effect_size ~ 1 + HDI,
  V = V_mat,
  random = list(~ 1 | country/ESID),
  data = bflpedat,
  method = "REML")

## Warning: Rows with NAs omitted from model fitting.

# Summarize the results
summary(MEM3L.hdi.csc, digits=4)

##
## Multivariate Meta-Analysis Model (k = 242; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 121.4148 -242.8295 -234.8295 -220.9070 -234.6593
##
## Variance Components:
##
##           estim  sqrt  nlvls  fixed      factor
## sigma^2.1 0.0145 0.1206   83    no      country
## sigma^2.2 0.0071 0.0842  242    no  country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 240) = 872.4148, p-val < .0001
##
## Test of Moderators (coefficient 2):
## QM(df = 1) = 9.0647, p-val = 0.0026
##
## Model Results:
```

```
##
##          estimate      se      zval      pval      ci.lb      ci.ub
## intrcpt  -0.9312  0.1582  -5.8877  <.0001  -1.2412  -0.6212  ***
## HDI      0.5585  0.1855   3.0108  0.0026   0.1949   0.9221   **
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Robust CIs
```

```
conf_int(MEM3L.hdi.csc, vcov = "CR2")
```

```
##      Coef Estimate      SE d.f. Lower 95% CI Upper 95% CI
## 1 intrcpt  -0.931 0.205 42.4      -1.3455      -0.517
## 2      HDI    0.559 0.235 45.6      0.0862      1.031
```

```
# Variance explanations
```

```
# Country level
```

```
(REM3L.csc$sigma2[1]-MEM3L.hdi.csc$sigma2[1])*100/REM3L.csc$sigma2[1]
```

```
## [1] 11.55301
```

```
# Effect size level
```

```
(REM3L.csc$sigma2[2]-MEM3L.hdi.csc$sigma2[2])*100/REM3L.csc$sigma2[2]
```

```
## [1] 0.2603856
```

## 4 Selection bias

### 4.1 Egger's regression test and Precision-Effect Test (PET)

```
## Details: https://wviechtb.github.io/metafor/reference/regtest.html
## Three-level REM with standard error as the predictor
MEM3Lc.pet <- rma.mv(effect_size ~ 1 + sqrt(var),
                    var,
                    random = list(~ 1 | country/ESID),
                    data = bflpedat,
                    method = "REML",
                    tdist = TRUE)

# Summarize the results
summary(MEM3Lc.pet, digits=4)

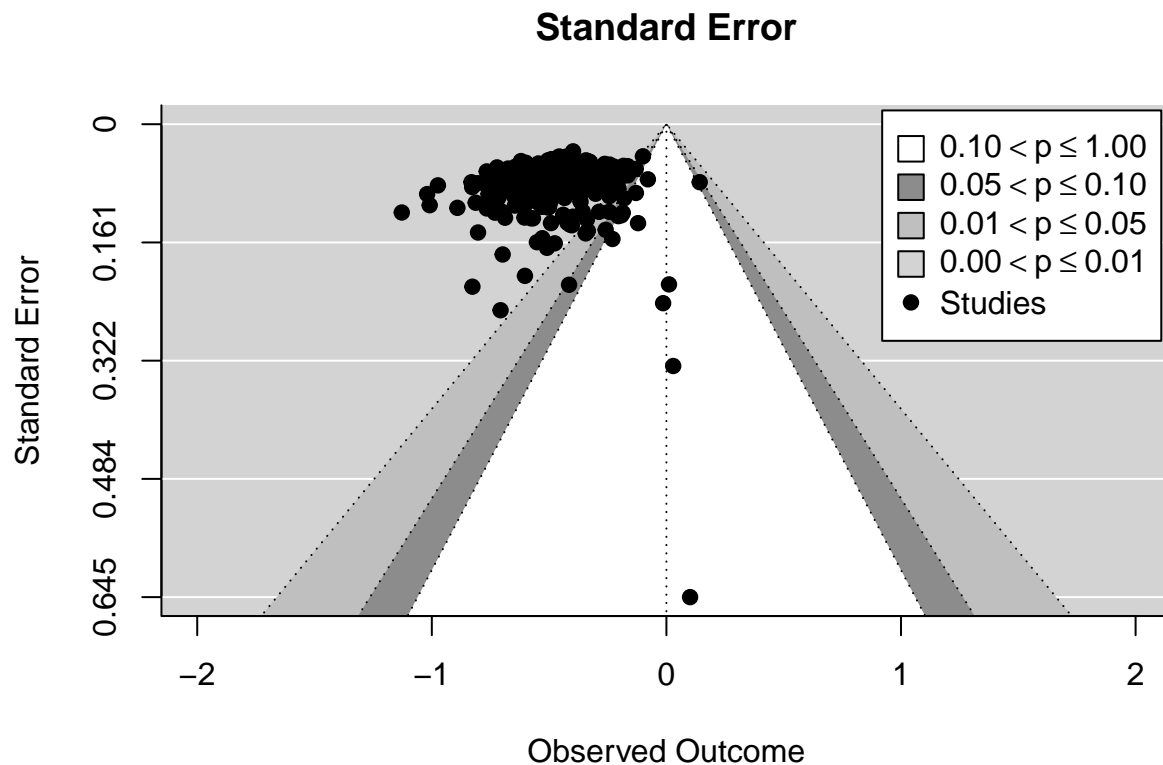
##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##   logLik   Deviance      AIC      BIC      AICc
## 128.7794 -257.5588 -249.5588 -235.5375 -249.3928
##
## Variance Components:
##
##      estim  sqrt  nlvls  fixed      factor
## sigma^2.1 0.0212 0.1457   85     no      country
## sigma^2.2 0.0028 0.0528  248     no country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 246) = 1211.7570, p-val < .0001
##
## Test of Moderators (coefficient 2):
## F(df1 = 1, df2 = 246) = 9.4099, p-val = 0.0024
##
## Model Results:
##
##      estimate      se      tval  df  pval  ci.lb  ci.ub
## intrcpt    -0.3611 0.0340 -10.6072 246 <.0001 -0.4281 -0.2940 ***
## sqrt(var)  -1.0679 0.3481  -3.0676 246 0.0024 -1.7535 -0.3822 **
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

From these mixed-effects meta-regression models, we can conclude that there is evidence on selection bias, due to the significant effect of the standard error as a moderator.

## 4.2 Funnel Plot Test

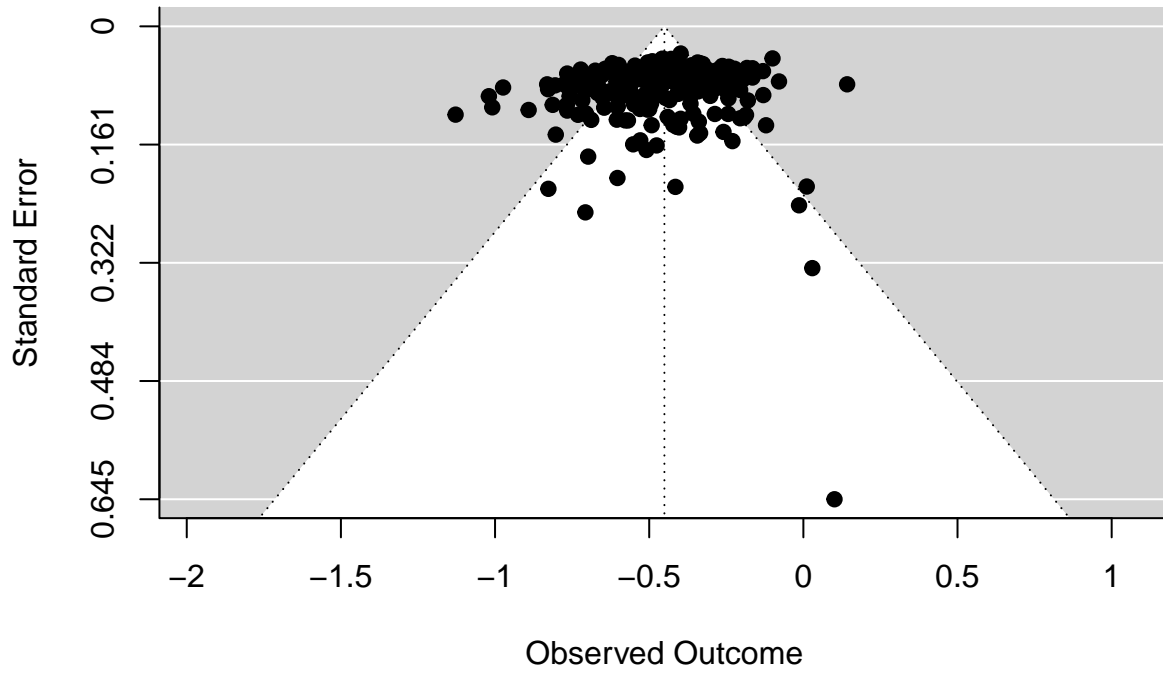
Note: In the current data set, sample sizes were missing. They were randomly generated (random numbers in Excel between 1000 and 5000) to illustrate the procedure.

```
#####  
# Contour-enhanced funnel plot  
funnel(REM3Lc, main="Standard Error",  
       level=c(90, 95, 99),  
       shade=c("white", "gray55", "gray75"),  
       refline=0,  
       legend=TRUE)
```



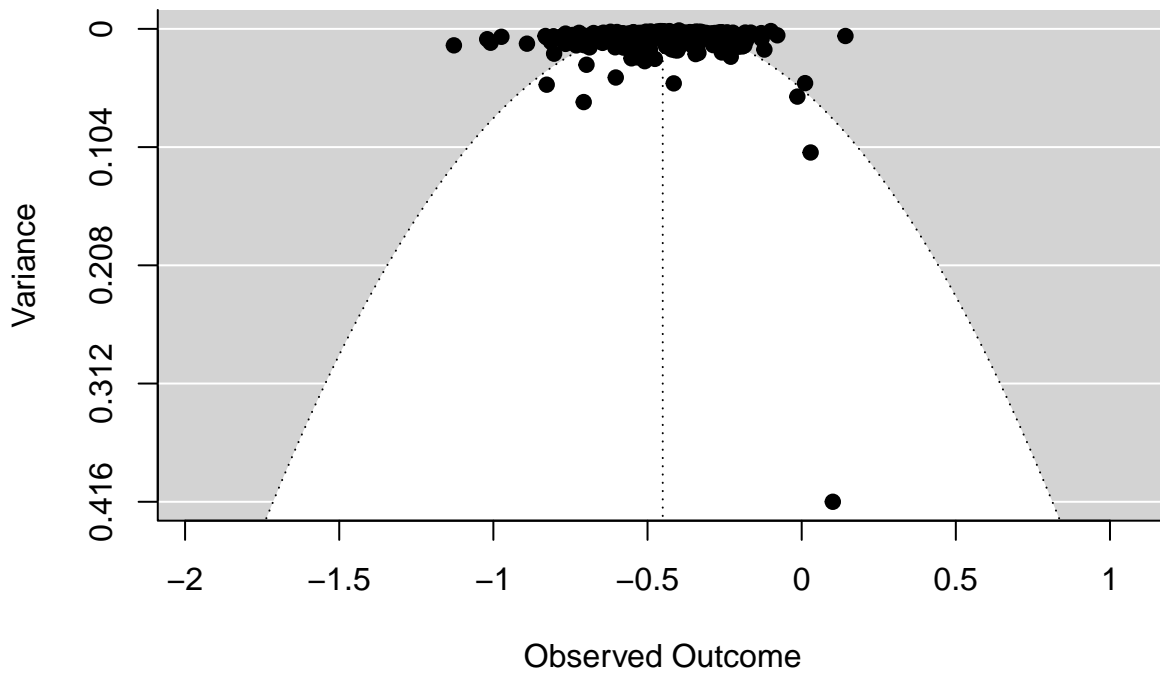
```
# Funnel plots  
funnel(REM3Lc, main="Standard Error")
```

### Standard Error



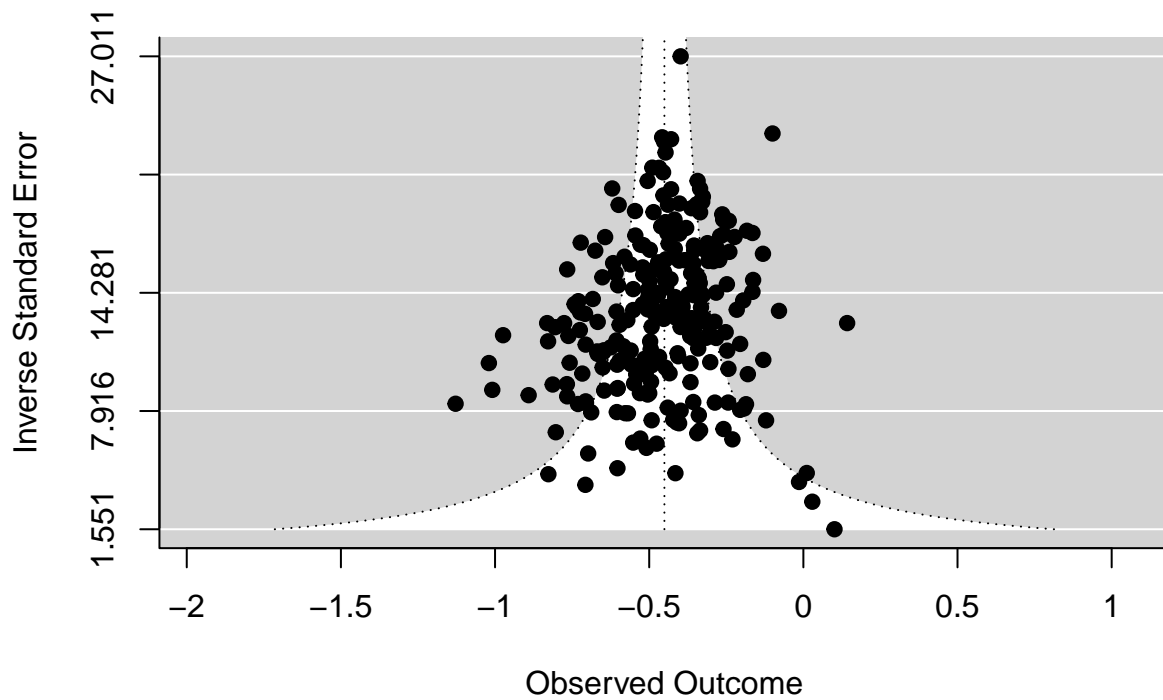
```
funnel(REM3Lc, yaxis="vi", main="Sampling Variance")
```

### Sampling Variance



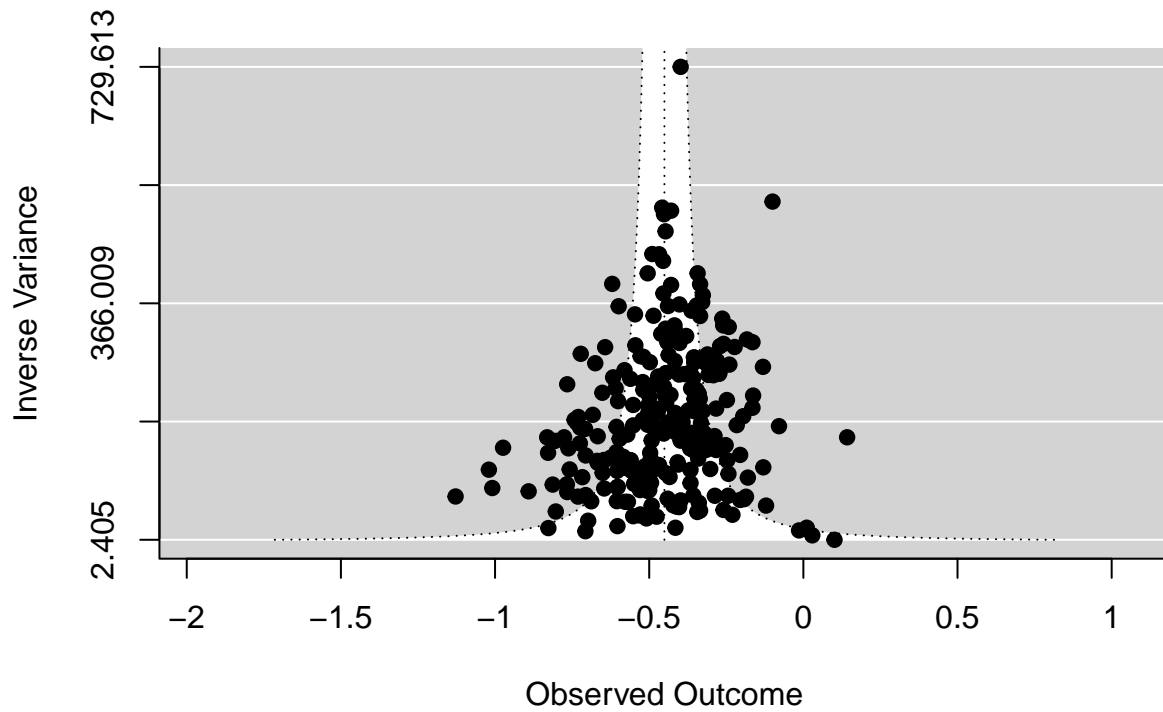
```
funnel(REM3Lc, yaxis="seinv", main="Inverse Standard Error")
```

### Inverse Standard Error



```
funnel(REM3Lc, yaxis="vinv", main="Inverse Sampling Variance")
```

### Inverse Sampling Variance



```
#####  
## Three-level REM with N as the predictor
```

```
MEM3Lc.fpt <- rma.mv(effect_size ~ 1 + bflpedat$N,
                    var,
                    random = list(~ 1 | country/ESID),
                    data = bflpedat,
                    method = "REML",
                    tdist = TRUE)

# Summarize the results
summary(MEM3Lc.fpt, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik    Deviance      AIC      BIC      AICc
## 123.8136 -247.6273 -239.6273 -225.6059 -239.4613
##
## Variance Components:
##
##           estim    sqrt  nlvls  fixed      factor
## sigma^2.1 0.0208 0.1442    85     no      country
## sigma^2.2 0.0036 0.0599   248     no  country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 246) = 1252.0498, p-val < .0001
##
## Test of Moderators (coefficient 2):
## F(df1 = 1, df2 = 246) = 0.6691, p-val = 0.4142
##
## Model Results:
##
##           estimate      se      tval  df    pval    ci.lb    ci.ub
## intrcpt      -0.4346 0.0265 -16.4067 246 <.0001 -0.4867 -0.3824 ***
## bflpedat$N   -0.0000 0.0000  -0.8180 246 0.4142 -0.0000  0.0000
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

From these mixed-effects meta-regression models, we can conclude that there is no evidence on selection bias, due to the insignificant effect of the standard error as a moderator.

### 4.3 Precision-Effect Estimate with SE (PEESE)

```
## Details: https://wviechtb.github.io/metafor/reference/regtest.html
## Three-level REM with standard error as the predictor
MEM3Lc.peese <- rma.mv(effect_size ~ 1 + var,
                      var,
                      random = list(~ 1 | country/ESID),
```

```

      data = bflpedat,
      method = "REML",
      tdist = TRUE)

# Summarize the results
summary(MEM3Lc.peese, digits=4)

##
## Multivariate Meta-Analysis Model (k = 248; method: REML)
##
##      logLik   Deviance      AIC      BIC      AICc
## 125.4660 -250.9320 -242.9320 -228.9107 -242.7661
##
## Variance Components:
##
##      estim  sqrt  nlvls  fixed      factor
## sigma^2.1 0.0209 0.1447   85     no      country
## sigma^2.2 0.0034 0.0581  248     no  country/ESID
##
## Test for Residual Heterogeneity:
## QE(df = 246) = 1245.7702, p-val < .0001
##
## Test of Moderators (coefficient 2):
## F(df1 = 1, df2 = 246) = 1.1475, p-val = 0.2851
##
## Model Results:
##
##      estimate      se      tval  df  pval  ci.lb  ci.ub
## intrcpt  -0.4404 0.0201 -21.9349 246 <.0001 -0.4799 -0.4009 ***
## var      -1.2330 1.1511  -1.0712 246 0.2851 -3.5002  1.0342
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

No evidence for selection bias.

#### 4.4 Begg's Correlation Test

Note: In the current data set, sample sizes were missing. They were randomly generated (random numbers in Excel between 1000 and 5000) to illustrate the procedure.

```

## Source code: Fernandez-Castilla et al. (2021)
## https://doi.org/10.1080/00220973.2019.1582470

## Define the precision
preci = 1/bflpedat$var
Stand_Var = bflpedat$var - (1/(sum(preci)))

```

```

## Draw from the model REM3Lc
## Define the pooled effect size from a baseline model
pooled_es <- REM3Lc$b[1]
Stand_es <- (bflpedat$effect_size - (pooled_es))/sqrt(Stand_Var)
N <- bflpedat$N
df_cor_N <- data.frame(Stand_es, N)

## Correlation coefficient and test
cor(df_cor_N, method = "kendall" )

```

```

##          Stand_es          N
## Stand_es 1.0000000 0.0252404
## N         0.0252404 1.0000000

```

```

bct <- cor.test(Stand_es, N, method = "kendall")
bct

```

```

##
## Kendall's rank correlation tau
##
## data:  Stand_es and N
## z = 0.592, p-value = 0.5538
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##      tau
## 0.0252404

```

Result: No evidence for selection bias (insignificant correlation).

## 4.5 Trim-and-fill analyses

### 4.5.1 Trim-and-fill analyses with the estimator $R_0^+$

```

## Source code: Fernandez-Castilla et al. (2021)
## https://doi.org/10.1080/00220973.2019.1582470

## Create an object d with all observed effect sizes
d <- bflpedat$effect_size

## Create a new function
## see the source code
R0_func<-function(d, pooled_d){
  d_difference1=0
  rank_positive=0
  final_rank=0
  d_difference<- d-pooled_d
  for(i in 1:length(d_difference)){
    d_difference1[i]<- if ( d_difference[i]<0) d_difference[i]*-1

```

```

    else d_difference[i] }
rank_positive<-rank(d_difference1)
for(i in 1:length(d_difference)){
  final_rank[i]<- if (d_difference[i]<0) rank_positive[i]*-1
  else rank_positive[i]}
O=length(d_difference)-(min(final_rank)*-1)
RO_ps= O-1
RO=if (RO_ps<0) 0 else RO_ps
print(RO)
}

## Obtain the RO+ estimator
RO_func(d, pooled_es)

```

```
## [1] 0
```

Given that  $R_0^+ < 3$ , there is no evidence for selection bias.

#### 4.5.2 Trim-and-fill analyses with the estimator $L_0^+$

```

## Source code: Fernandez-Castilla et al. (2021)
## https://doi.org/10.1080/00220973.2019.1582470

## Define a new function
LO_func<-function(d, pooled_d){
  d_difference1=0
  rank_positive=0
  d_difference<- d-pooled_d
  d_difference1=0
  only_positive=0
  final_rank=0
  N_ES<-length(d)
  for(i in 1:length(d_difference)){
    d_difference1[i]<- if ( d_difference[i]<0) d_difference[i]*-1
    else d_difference[i] }
  rank_positive<-rank(d_difference1)
  for(i in 1:length(d_difference)){
    final_rank[i]<- if (d_difference[i]<0) rank_positive[i]*-1
    else rank_positive[i]
    only_positive[i]<- if (final_rank[i]<0) 0
    else final_rank[i]}
  t=sum(only_positive)
  LO_ps=(4*t-N_ES*(N_ES+1))/(2*N_ES-1)
  LO= if (LO_ps<0) 0 else LO_ps
  print(LO)
}

```

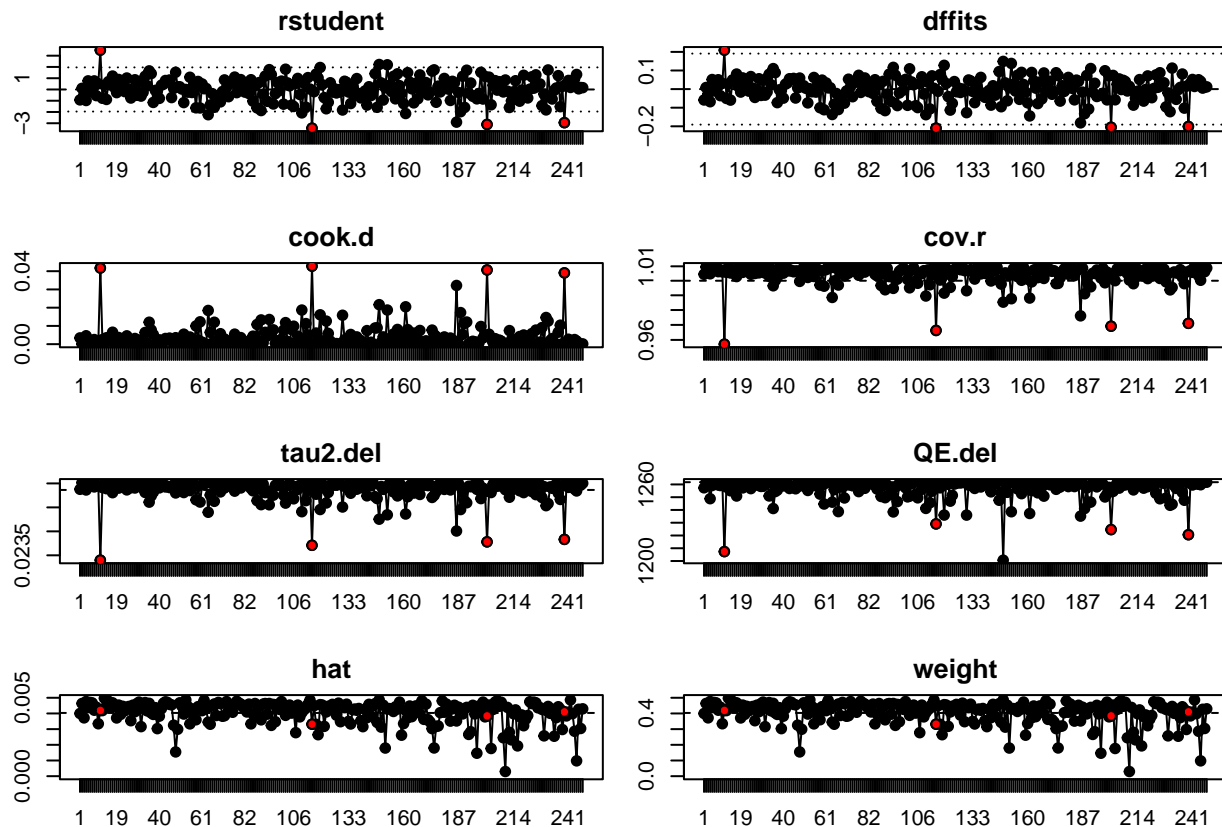
```
## Obtain the LO+ estimate
L0_func(d, pooled_es)
```

```
## [1] 1.292929
```

Given that  $L_0^+ < 2$ , there is no evidence for selection bias.

#### 4.6 Influential effect sizes

```
# Analyses based on the REM with variation between effect sizes
# Plots indicating potential influential effect sizes
infl1 <- influence.rma.uni(rma(effect_size,
                              var,
                              data = bflpedat,
                              method = "REML"))
plot.infl.rma.uni(infl1)
```



```
# Increase max print options to show all effect sizes
options(max.print = 1000000)
print(infl1)
```

```
##
##      rstudent dffits cook.d cov.r tau2.del   QE.del   hat weight  dfbs inf
## 1      -0.9259 -0.0585 0.0034 1.0044   0.0249 1257.3521 0.0040 0.3998 -0.0585
```

## 2	0.1017	0.0086	0.0001	1.0096	0.0250	1261.7647	0.0046	0.4638	0.0086
## 3	-0.4484	-0.0264	0.0007	1.0068	0.0250	1260.7458	0.0037	0.3719	-0.0264
## 4	-0.9810	-0.0679	0.0046	1.0048	0.0249	1248.8334	0.0048	0.4764	-0.0679
## 5	0.7651	0.0514	0.0026	1.0065	0.0249	1259.1472	0.0044	0.4370	0.0514
## 6	-0.1677	-0.0099	0.0001	1.0096	0.0250	1261.0269	0.0047	0.4702	-0.0099
## 7	0.0661	0.0062	0.0000	1.0097	0.0250	1261.7490	0.0047	0.4673	0.0062
## 8	0.7620	0.0502	0.0025	1.0063	0.0249	1259.5680	0.0042	0.4207	0.0502
## 9	-0.5461	-0.0352	0.0012	1.0076	0.0250	1259.1251	0.0044	0.4403	-0.0352
## 10	0.5118	0.0306	0.0009	1.0061	0.0249	1261.2775	0.0033	0.3343	0.0306
## 11	3.4910	0.2093	0.0417	0.9572	0.0234	1207.3691	0.0042	0.4194	0.2093
## 12	-0.7820	-0.0501	0.0025	1.0058	0.0249	1257.9161	0.0042	0.4181	-0.0501
## 13	0.3707	0.0278	0.0008	1.0097	0.0250	1260.4548	0.0050	0.4973	0.0278
## 14	-0.8863	-0.0592	0.0035	1.0053	0.0249	1255.0120	0.0045	0.4482	-0.0592
## 15	0.6970	0.0492	0.0024	1.0076	0.0249	1257.8482	0.0048	0.4781	0.0492
## 16	0.6839	0.0462	0.0021	1.0070	0.0249	1259.7448	0.0044	0.4373	0.0462
## 17	1.1901	0.0810	0.0065	1.0029	0.0248	1250.9250	0.0047	0.4681	0.0810
## 18	0.1580	0.0121	0.0001	1.0091	0.0250	1261.7464	0.0045	0.4453	0.0121
## 19	-0.2825	-0.0160	0.0003	1.0073	0.0250	1261.2742	0.0037	0.3712	-0.0160
## 20	0.8860	0.0602	0.0036	1.0058	0.0249	1257.2788	0.0045	0.4535	0.0602
## 21	0.2752	0.0187	0.0003	1.0079	0.0250	1261.6398	0.0039	0.3943	0.0187
## 22	0.6905	0.0471	0.0022	1.0071	0.0249	1259.4865	0.0045	0.4456	0.0471
## 23	-0.1919	-0.0110	0.0001	1.0084	0.0250	1261.3445	0.0042	0.4168	-0.0110
## 24	1.0168	0.0675	0.0046	1.0044	0.0249	1256.7076	0.0044	0.4384	0.0675
## 25	0.0587	0.0055	0.0000	1.0092	0.0250	1261.7507	0.0044	0.4447	0.0055
## 26	-0.7608	-0.0448	0.0020	1.0050	0.0249	1259.5262	0.0035	0.3537	-0.0448
## 27	-0.0762	-0.0035	0.0000	1.0093	0.0250	1261.5207	0.0045	0.4529	-0.0035
## 28	0.8603	0.0557	0.0031	1.0054	0.0249	1259.1020	0.0041	0.4112	0.0557
## 29	-0.5182	-0.0341	0.0012	1.0082	0.0250	1258.4224	0.0046	0.4630	-0.0341
## 30	0.6747	0.0474	0.0023	1.0076	0.0250	1258.5318	0.0047	0.4724	0.0475
## 31	-0.5907	-0.0326	0.0011	1.0053	0.0249	1260.6059	0.0032	0.3159	-0.0325
## 32	-0.2542	-0.0154	0.0002	1.0087	0.0250	1260.9988	0.0044	0.4364	-0.0154
## 33	1.1739	0.0801	0.0064	1.0031	0.0248	1250.9528	0.0047	0.4696	0.0801
## 34	0.3393	0.0233	0.0005	1.0082	0.0250	1261.4905	0.0042	0.4165	0.0233
## 35	1.6523	0.1102	0.0120	0.9965	0.0246	1241.1132	0.0046	0.4646	0.1101
## 36	1.4115	0.0884	0.0078	1.0001	0.0247	1254.4008	0.0040	0.4029	0.0884
## 37	-1.1510	-0.0729	0.0053	1.0024	0.0248	1255.5271	0.0039	0.3937	-0.0729
## 38	-0.0807	-0.0038	0.0000	1.0090	0.0250	1261.5598	0.0044	0.4369	-0.0038
## 39	-0.1637	-0.0080	0.0001	1.0062	0.0250	1261.6162	0.0030	0.3032	-0.0080
## 40	-0.7915	-0.0486	0.0024	1.0052	0.0249	1258.8451	0.0038	0.3827	-0.0486
## 41	0.3792	0.0263	0.0007	1.0083	0.0250	1261.3421	0.0043	0.4300	0.0263
## 42	0.4663	0.0332	0.0011	1.0087	0.0250	1260.6187	0.0047	0.4655	0.0332
## 43	0.7972	0.0557	0.0031	1.0068	0.0249	1256.8016	0.0047	0.4744	0.0557
## 44	0.7394	0.0516	0.0027	1.0071	0.0249	1257.9132	0.0047	0.4702	0.0516
## 45	0.8424	0.0576	0.0033	1.0062	0.0249	1257.5271	0.0046	0.4575	0.0576
## 46	-0.7211	-0.0480	0.0023	1.0068	0.0249	1256.5806	0.0046	0.4552	-0.0480
## 47	0.0901	0.0063	0.0000	1.0067	0.0250	1261.7650	0.0033	0.3252	0.0063
## 48	1.5301	0.0594	0.0035	0.9995	0.0248	1258.7155	0.0015	0.1546	0.0595
## 49	0.2254	0.0134	0.0002	1.0061	0.0250	1261.7199	0.0030	0.2992	0.0134

\*

## 50	0.2398	0.0180	0.0003	1.0095	0.0250	1261.6024	0.0047	0.4681	0.0181
## 51	0.0413	0.0044	0.0000	1.0093	0.0250	1261.7368	0.0045	0.4476	0.0044
## 52	-1.1656	-0.0741	0.0055	1.0022	0.0248	1255.2592	0.0040	0.3961	-0.0741
## 53	0.0003	0.0018	0.0000	1.0101	0.0250	1261.5977	0.0049	0.4866	0.0018
## 54	-0.2389	-0.0145	0.0002	1.0089	0.0250	1260.9884	0.0044	0.4444	-0.0145
## 55	1.0840	0.0625	0.0039	1.0029	0.0249	1259.0671	0.0033	0.3334	0.0625
## 56	-0.2691	-0.0162	0.0003	1.0085	0.0250	1261.0056	0.0043	0.4286	-0.0162
## 57	0.2851	0.0192	0.0004	1.0078	0.0250	1261.6278	0.0039	0.3914	0.0192
## 58	-1.6300	-0.1001	0.0100	0.9970	0.0247	1252.3751	0.0036	0.3597	-0.1002
## 59	0.7121	0.0484	0.0023	1.0069	0.0249	1259.3806	0.0044	0.4434	0.0484
## 60	-1.6442	-0.1111	0.0122	0.9962	0.0246	1244.6586	0.0043	0.4318	-0.1111
## 61	0.4927	0.0331	0.0011	1.0077	0.0250	1261.0144	0.0042	0.4177	0.0331
## 62	0.0673	0.0064	0.0000	1.0100	0.0250	1261.7416	0.0048	0.4839	0.0064
## 63	0.0922	0.0076	0.0001	1.0089	0.0250	1261.7643	0.0043	0.4301	0.0076
## 64	-2.2430	-0.1367	0.0185	0.9885	0.0244	1246.0987	0.0034	0.3446	-0.1369
## 65	-1.3793	-0.0806	0.0065	0.9999	0.0248	1255.9341	0.0033	0.3311	-0.0806
## 66	-0.3700	-0.0235	0.0006	1.0086	0.0250	1260.1513	0.0045	0.4507	-0.0235
## 67	-1.5787	-0.1101	0.0120	0.9970	0.0246	1238.6974	0.0046	0.4605	-0.1100
## 68	-0.6593	-0.0403	0.0016	1.0061	0.0249	1259.5787	0.0039	0.3870	-0.0403
## 69	-0.5284	-0.0351	0.0012	1.0082	0.0250	1257.8170	0.0047	0.4705	-0.0351
## 70	-1.1046	-0.0758	0.0057	1.0033	0.0248	1249.3861	0.0046	0.4625	-0.0757
## 71	-0.1837	-0.0101	0.0001	1.0077	0.0250	1261.4736	0.0038	0.3821	-0.0101
## 72	0.8492	0.0533	0.0028	1.0052	0.0249	1259.6402	0.0039	0.3854	0.0533
## 73	0.1755	0.0138	0.0002	1.0098	0.0250	1261.7165	0.0048	0.4762	0.0138
## 74	-0.6425	-0.0409	0.0017	1.0067	0.0249	1258.9413	0.0042	0.4214	-0.0409
## 75	0.5375	0.0375	0.0014	1.0081	0.0250	1260.3717	0.0045	0.4544	0.0375
## 76	0.5707	0.0375	0.0014	1.0071	0.0250	1260.7738	0.0041	0.4076	0.0375
## 77	1.1011	0.0741	0.0055	1.0038	0.0248	1254.3616	0.0045	0.4546	0.0741
## 78	-0.2841	-0.0181	0.0003	1.0094	0.0250	1260.0104	0.0048	0.4781	-0.0181
## 79	0.2455	0.0184	0.0003	1.0094	0.0250	1261.5911	0.0047	0.4670	0.0184
## 80	0.8748	0.0599	0.0036	1.0060	0.0249	1256.9489	0.0046	0.4603	0.0599
## 81	0.0759	0.0066	0.0000	1.0091	0.0250	1261.7597	0.0044	0.4412	0.0066
## 82	-0.2356	-0.0141	0.0002	1.0087	0.0250	1261.1111	0.0043	0.4320	-0.0141
## 83	0.5103	0.0338	0.0011	1.0074	0.0250	1261.0007	0.0041	0.4094	0.0338
## 84	-0.3617	-0.0232	0.0005	1.0088	0.0250	1259.9851	0.0046	0.4596	-0.0232
## 85	0.8657	0.0504	0.0025	1.0044	0.0249	1260.1215	0.0033	0.3334	0.0504
## 86	-1.0449	-0.0662	0.0044	1.0034	0.0248	1256.3918	0.0040	0.3971	-0.0662
## 87	0.3341	0.0243	0.0006	1.0091	0.0250	1261.3205	0.0046	0.4636	0.0243
## 88	-1.6377	-0.1042	0.0108	0.9967	0.0246	1250.5615	0.0038	0.3848	-0.1042
## 89	0.6127	0.0419	0.0018	1.0075	0.0250	1260.0975	0.0044	0.4433	0.0419
## 90	-1.8741	-0.1158	0.0133	0.9938	0.0246	1249.6006	0.0036	0.3592	-0.1159
## 91	-0.4153	-0.0269	0.0007	1.0086	0.0250	1259.5494	0.0046	0.4598	-0.0269
## 92	1.1901	0.0806	0.0065	1.0028	0.0248	1251.7466	0.0046	0.4634	0.0806
## 93	-1.0535	-0.0666	0.0044	1.0033	0.0248	1256.3911	0.0040	0.3953	-0.0666
## 94	1.7608	0.1169	0.0135	0.9947	0.0246	1238.4428	0.0046	0.4638	0.1168
## 95	1.3289	0.0749	0.0056	1.0009	0.0248	1257.7683	0.0032	0.3239	0.0749
## 96	-1.2904	-0.0887	0.0078	1.0011	0.0248	1246.5563	0.0046	0.4576	-0.0887
## 97	-0.3715	-0.0209	0.0004	1.0066	0.0250	1261.1259	0.0035	0.3475	-0.0209

## 98	0.6167	0.0429	0.0018	1.0078	0.0250	1259.7163	0.0046	0.4578	0.0429
## 99	-0.1026	-0.0052	0.0000	1.0088	0.0250	1261.5389	0.0043	0.4263	-0.0052
## 100	0.7442	0.0523	0.0027	1.0072	0.0249	1257.3653	0.0048	0.4763	0.0523
## 101	-1.3040	-0.0874	0.0076	1.0009	0.0248	1250.2339	0.0043	0.4349	-0.0874
## 102	1.7989	0.1091	0.0118	0.9951	0.0246	1250.7983	0.0039	0.3877	0.1092
## 103	0.5813	0.0385	0.0015	1.0072	0.0250	1260.6647	0.0041	0.4147	0.0385
## 104	0.3246	0.0234	0.0005	1.0090	0.0250	1261.4111	0.0045	0.4536	0.0234
## 105	-1.3824	-0.0893	0.0079	0.9999	0.0247	1252.3975	0.0040	0.4022	-0.0893
## 106	-1.5118	-0.0995	0.0098	0.9982	0.0247	1249.4927	0.0041	0.4142	-0.0995
## 107	0.5262	0.0285	0.0008	1.0050	0.0249	1261.3484	0.0028	0.2774	0.0285
## 108	0.9808	0.0663	0.0044	1.0049	0.0249	1256.1363	0.0045	0.4534	0.0663
## 109	-0.4468	-0.0265	0.0007	1.0069	0.0250	1260.7057	0.0038	0.3781	-0.0265
## 110	-2.0807	-0.1377	0.0187	0.9895	0.0244	1241.2148	0.0041	0.4051	-0.1377
## 111	0.5470	0.0368	0.0014	1.0075	0.0250	1260.7452	0.0042	0.4228	0.0368
## 112	-1.5772	-0.1066	0.0113	0.9972	0.0246	1245.6218	0.0043	0.4336	-0.1066
## 113	1.0520	0.0713	0.0051	1.0043	0.0249	1254.6888	0.0046	0.4584	0.0713
## 114	0.0369	0.0043	0.0000	1.0101	0.0250	1261.6923	0.0049	0.4859	0.0043
## 115	-3.4342	-0.2099	0.0428	0.9663	0.0237	1229.0576	0.0033	0.3315	-0.2107
## 116	-0.1991	-0.0118	0.0001	1.0091	0.0250	1261.1159	0.0045	0.4492	-0.0118
## 117	1.3736	0.0781	0.0061	1.0005	0.0248	1257.3307	0.0033	0.3308	0.0781
## 118	1.0279	0.0529	0.0028	1.0027	0.0249	1259.9617	0.0026	0.2644	0.0529
## 119	1.9523	0.1278	0.0161	0.9915	0.0245	1235.9887	0.0046	0.4564	0.1277
## 120	0.4250	0.0301	0.0009	1.0086	0.0250	1261.0096	0.0045	0.4538	0.0301
## 121	-0.7260	-0.0405	0.0016	1.0047	0.0249	1260.0824	0.0032	0.3178	-0.0405
## 122	-1.7140	-0.1129	0.0126	0.9954	0.0246	1246.8971	0.0041	0.4096	-0.1129
## 123	-1.1379	-0.0767	0.0059	1.0028	0.0248	1251.4504	0.0045	0.4462	-0.0767
## 124	0.5967	0.0406	0.0017	1.0075	0.0250	1260.3101	0.0044	0.4370	0.0406
## 125	-0.2366	-0.0145	0.0002	1.0092	0.0250	1260.8342	0.0046	0.4579	-0.0145
## 126	0.0711	0.0064	0.0000	1.0094	0.0250	1261.7557	0.0045	0.4542	0.0064
## 127	-0.3110	-0.0182	0.0003	1.0076	0.0250	1261.1167	0.0039	0.3889	-0.0182
## 128	-0.0176	0.0005	0.0000	1.0097	0.0250	1261.6239	0.0047	0.4660	0.0005
## 129	-0.2165	-0.0116	0.0001	1.0070	0.0250	1261.4816	0.0035	0.3473	-0.0116
## 130	-1.8239	-0.1267	0.0159	0.9931	0.0245	1235.9697	0.0045	0.4498	-0.1266
## 131	0.7300	0.0491	0.0024	1.0067	0.0249	1259.4227	0.0044	0.4368	0.0491
## 132	0.6296	0.0438	0.0019	1.0077	0.0250	1259.5642	0.0046	0.4593	0.0438
## 133	-0.2327	-0.0126	0.0002	1.0070	0.0250	1261.4463	0.0035	0.3496	-0.0126
## 134	-1.2504	-0.0730	0.0053	1.0011	0.0248	1256.8643	0.0033	0.3328	-0.0730
## 135	0.2509	0.0178	0.0003	1.0085	0.0250	1261.6509	0.0042	0.4232	0.0178
## 136	-0.5385	-0.0342	0.0012	1.0074	0.0250	1259.5220	0.0043	0.4282	-0.0342
## 137	0.9669	0.0630	0.0040	1.0047	0.0249	1258.0104	0.0042	0.4203	0.0630
## 138	-0.3622	-0.0225	0.0005	1.0083	0.0250	1260.5356	0.0043	0.4318	-0.0225
## 139	-0.7513	-0.0443	0.0020	1.0051	0.0249	1259.5636	0.0035	0.3545	-0.0443
## 140	-0.5462	-0.0300	0.0009	1.0054	0.0249	1260.7538	0.0032	0.3157	-0.0300
## 141	-0.1368	-0.0074	0.0001	1.0087	0.0250	1261.4596	0.0043	0.4257	-0.0074
## 142	-1.3893	-0.0863	0.0074	0.9998	0.0247	1254.1811	0.0037	0.3728	-0.0863
## 143	0.2433	0.0181	0.0003	1.0093	0.0250	1261.6168	0.0046	0.4583	0.0181
## 144	-0.5418	-0.0326	0.0011	1.0066	0.0250	1260.2517	0.0038	0.3829	-0.0326
## 145	0.0355	0.0040	0.0000	1.0093	0.0250	1261.7308	0.0045	0.4488	0.0040

\*

## 146	1.4060	0.0943	0.0089	1.0001	0.0247	1247.5963	0.0046	0.4625	0.0943
## 147	-1.5852	-0.0949	0.0090	0.9977	0.0247	1253.6997	0.0034	0.3428	-0.0949
## 148	2.2183	0.1489	0.0217	0.9853	0.0243	1200.6701	0.0049	0.4872	0.1486
## 149	0.5831	0.0335	0.0011	1.0054	0.0249	1261.1593	0.0031	0.3126	0.0334
## 150	-0.9381	-0.0631	0.0040	1.0049	0.0249	1253.8261	0.0045	0.4526	-0.0631
## 151	0.1596	0.0074	0.0001	1.0037	0.0249	1261.7519	0.0018	0.1793	0.0074
## 152	2.1930	0.1380	0.0187	0.9877	0.0243	1238.5961	0.0043	0.4276	0.1379
## 153	0.3740	0.0260	0.0007	1.0084	0.0250	1261.3536	0.0043	0.4311	0.0260
## 154	-0.1834	-0.0106	0.0001	1.0087	0.0250	1261.3078	0.0043	0.4300	-0.0106
## 155	1.1617	0.0743	0.0055	1.0028	0.0248	1256.5178	0.0041	0.4126	0.0743
## 156	-0.8336	-0.0551	0.0030	1.0057	0.0249	1256.1540	0.0044	0.4422	-0.0551
## 157	1.3926	0.0900	0.0081	1.0003	0.0247	1252.7590	0.0043	0.4284	0.0900
## 158	0.3373	0.0248	0.0006	1.0093	0.0250	1261.2335	0.0047	0.4726	0.0248
## 159	-0.3311	-0.0161	0.0003	1.0051	0.0249	1261.4274	0.0026	0.2613	-0.0161
## 160	-1.2603	-0.0739	0.0054	1.0011	0.0248	1256.7233	0.0034	0.3353	-0.0739
## 161	-2.1365	-0.1443	0.0205	0.9881	0.0244	1237.2252	0.0042	0.4193	-0.1443
## 162	1.4905	0.0895	0.0080	0.9992	0.0247	1255.1225	0.0037	0.3721	0.0895
## 163	0.6508	0.0446	0.0020	1.0074	0.0250	1259.7363	0.0045	0.4476	0.0446
## 164	0.0596	0.0051	0.0000	1.0079	0.0250	1261.7599	0.0038	0.3795	0.0051
## 165	-1.1897	-0.0780	0.0061	1.0021	0.0248	1253.4100	0.0042	0.4202	-0.0780
## 166	0.5514	0.0382	0.0015	1.0079	0.0250	1260.3983	0.0045	0.4481	0.0382
## 167	-0.8264	-0.0523	0.0027	1.0053	0.0249	1257.9725	0.0041	0.4059	-0.0523
## 168	1.2956	0.0868	0.0075	1.0016	0.0248	1250.8266	0.0046	0.4571	0.0868
## 169	-0.1939	-0.0118	0.0001	1.0097	0.0250	1260.6519	0.0048	0.4810	-0.0118
## 170	0.4853	0.0303	0.0009	1.0067	0.0250	1261.2722	0.0036	0.3623	0.0303
## 171	0.6119	0.0426	0.0018	1.0078	0.0250	1259.7055	0.0046	0.4595	0.0426
## 172	-0.4784	-0.0278	0.0008	1.0065	0.0250	1260.7144	0.0036	0.3606	-0.0278
## 173	-0.2417	-0.0131	0.0002	1.0069	0.0250	1261.4374	0.0035	0.3455	-0.0131
## 174	1.6296	0.0880	0.0077	0.9980	0.0247	1256.1627	0.0030	0.3030	0.0881
## 175	1.7439	0.0726	0.0053	0.9982	0.0247	1257.4990	0.0018	0.1798	0.0727
## 176	-1.1364	-0.0645	0.0042	1.0020	0.0248	1257.9778	0.0032	0.3173	-0.0645
## 177	0.2976	0.0180	0.0003	1.0064	0.0250	1261.6535	0.0032	0.3202	0.0179
## 178	-0.2170	-0.0125	0.0002	1.0082	0.0250	1261.3222	0.0041	0.4050	-0.0125
## 179	0.0185	0.0030	0.0000	1.0100	0.0250	1261.6676	0.0048	0.4800	0.0030
## 180	0.9805	0.0629	0.0040	1.0045	0.0249	1258.2915	0.0041	0.4079	0.0629
## 181	0.1250	0.0101	0.0001	1.0095	0.0250	1261.7623	0.0046	0.4602	0.0101
## 182	0.0288	0.0034	0.0000	1.0088	0.0250	1261.7370	0.0042	0.4223	0.0034
## 183	-0.9767	-0.0580	0.0034	1.0035	0.0249	1258.3103	0.0035	0.3511	-0.0580
## 184	0.7241	0.0484	0.0023	1.0066	0.0249	1259.6321	0.0043	0.4298	0.0484
## 185	0.5953	0.0412	0.0017	1.0078	0.0250	1260.0229	0.0045	0.4523	0.0412
## 186	-2.9014	-0.1815	0.0322	0.9761	0.0240	1235.1741	0.0035	0.3521	-0.1819
## 187	-0.2668	-0.0163	0.0003	1.0087	0.0250	1260.9080	0.0044	0.4399	-0.0163
## 188	-1.9774	-0.1324	0.0173	0.9911	0.0245	1241.1185	0.0042	0.4164	-0.1324
## 189	1.1668	0.0732	0.0054	1.0027	0.0248	1257.1059	0.0040	0.3968	0.0732
## 190	-1.5934	-0.1072	0.0114	0.9970	0.0246	1245.9786	0.0043	0.4298	-0.1072
## 191	1.7107	0.1107	0.0121	0.9958	0.0246	1246.1562	0.0044	0.4388	0.1106
## 192	-0.3206	-0.0157	0.0002	1.0052	0.0249	1261.4372	0.0027	0.2661	-0.0157
## 193	0.5739	0.0313	0.0010	1.0050	0.0249	1261.2433	0.0028	0.2834	0.0313

## 194	-0.6064	-0.0399	0.0016	1.0075	0.0250	1258.0887	0.0045	0.4520	-0.0399
## 195	-0.5627	-0.0346	0.0012	1.0068	0.0250	1259.9634	0.0040	0.3969	-0.0346
## 196	-0.8341	-0.0318	0.0010	1.0018	0.0249	1260.6570	0.0015	0.1463	-0.0318
## 197	-0.3569	-0.0225	0.0005	1.0086	0.0250	1260.3064	0.0045	0.4475	-0.0225
## 198	1.5248	0.0987	0.0097	0.9986	0.0247	1250.1655	0.0043	0.4340	0.0987
## 199	-0.8494	-0.0581	0.0034	1.0060	0.0249	1252.5599	0.0047	0.4722	-0.0581
## 200	1.1605	0.0711	0.0050	1.0026	0.0248	1257.7263	0.0038	0.3782	0.0711
## 201	-3.1071	-0.2050	0.0407	0.9691	0.0238	1224.6105	0.0038	0.3844	-0.2053
## 202	1.0943	0.0736	0.0054	1.0038	0.0248	1254.5367	0.0045	0.4539	0.0736
## 203	-1.3586	-0.0577	0.0033	1.0000	0.0248	1258.7124	0.0018	0.1765	-0.0577
## 204	0.3773	0.0260	0.0007	1.0083	0.0250	1261.3646	0.0043	0.4257	0.0260
## 205	0.7840	0.0547	0.0030	1.0069	0.0249	1257.1088	0.0047	0.4731	0.0548
## 206	-0.0603	-0.0025	0.0000	1.0099	0.0250	1261.4166	0.0048	0.4809	-0.0025
## 207	0.6906	0.0471	0.0022	1.0071	0.0249	1259.4798	0.0045	0.4458	0.0471
## 208	0.5715	0.0405	0.0016	1.0082	0.0250	1259.6432	0.0047	0.4713	0.0405
## 209	-0.2209	-0.0101	0.0001	1.0049	0.0249	1261.5975	0.0024	0.2449	-0.0101
## 210	0.8423	0.0145	0.0002	1.0004	0.0249	1261.0609	0.0003	0.0296	0.0145
## 211	-0.1504	-0.0084	0.0001	1.0090	0.0250	1261.3671	0.0044	0.4387	-0.0084
## 212	-1.6019	-0.0864	0.0074	0.9979	0.0247	1255.7146	0.0028	0.2796	-0.0864
## 213	-1.0116	-0.0488	0.0024	1.0021	0.0249	1259.6538	0.0023	0.2313	-0.0488
## 214	-0.2041	-0.0115	0.0001	1.0080	0.0250	1261.3817	0.0040	0.3982	-0.0115
## 215	0.7752	0.0512	0.0026	1.0062	0.0249	1259.4350	0.0042	0.4227	0.0512
## 216	-0.5578	-0.0241	0.0006	1.0033	0.0249	1261.1435	0.0019	0.1930	-0.0241
## 217	-1.0432	-0.0646	0.0042	1.0033	0.0248	1257.0681	0.0038	0.3797	-0.0646
## 218	-0.0544	-0.0020	0.0000	1.0081	0.0250	1261.6730	0.0039	0.3908	-0.0020
## 219	1.2591	0.0709	0.0050	1.0015	0.0248	1258.2428	0.0032	0.3216	0.0709
## 220	-0.1896	-0.0102	0.0001	1.0073	0.0250	1261.5022	0.0036	0.3628	-0.0102
## 221	-0.7522	-0.0463	0.0021	1.0055	0.0249	1259.0119	0.0039	0.3869	-0.0463
## 222	0.0245	0.0034	0.0000	1.0098	0.0250	1261.6901	0.0047	0.4747	0.0034
## 223	1.1231	0.0762	0.0058	1.0036	0.0248	1253.0259	0.0046	0.4628	0.0762
## 224	0.1146	0.0096	0.0001	1.0097	0.0250	1261.7646	0.0047	0.4723	0.0096
## 225	1.2987	0.0880	0.0077	1.0016	0.0248	1248.7414	0.0047	0.4677	0.0880
## 226	0.2431	0.0176	0.0003	1.0088	0.0250	1261.6503	0.0044	0.4360	0.0176
## 227	0.4926	0.0339	0.0012	1.0080	0.0250	1260.8559	0.0044	0.4374	0.0339
## 228	-1.4347	-0.0956	0.0091	0.9992	0.0247	1249.3058	0.0043	0.4257	-0.0956
## 229	-0.4173	-0.0205	0.0004	1.0048	0.0249	1261.2804	0.0026	0.2571	-0.0204
## 230	-1.8162	-0.1215	0.0146	0.9937	0.0245	1243.6476	0.0042	0.4193	-0.1215
## 231	1.7350	0.1129	0.0126	0.9954	0.0246	1244.4762	0.0044	0.4449	0.1129
## 232	-0.6988	-0.0436	0.0019	1.0061	0.0249	1259.0632	0.0040	0.4008	-0.0436
## 233	-0.8479	-0.0552	0.0030	1.0054	0.0249	1256.8979	0.0043	0.4272	-0.0552
## 234	-0.0793	-0.0031	0.0000	1.0052	0.0249	1261.7159	0.0025	0.2547	-0.0031
## 235	0.1286	0.0092	0.0001	1.0077	0.0250	1261.7613	0.0037	0.3731	0.0092
## 236	1.2277	0.0812	0.0066	1.0023	0.0248	1253.7130	0.0044	0.4428	0.0811
## 237	-1.5293	-0.1023	0.0104	0.9979	0.0247	1247.6046	0.0043	0.4267	-0.1023
## 238	0.2630	0.0154	0.0002	1.0060	0.0250	1261.6926	0.0030	0.2975	0.0154
## 239	-2.9478	-0.2010	0.0391	0.9710	0.0238	1220.5969	0.0041	0.4102	-0.2010
## 240	0.4997	0.0321	0.0010	1.0070	0.0250	1261.1608	0.0038	0.3842	0.0321
## 241	0.8009	0.0519	0.0027	1.0058	0.0249	1259.5594	0.0041	0.4081	0.0519

\*

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```
## 242 0.1773 0.0141 0.0002 1.0100 0.0250 1261.7023 0.0049 0.4863 0.0141
## 243 0.6476 0.0419 0.0018 1.0066 0.0249 1260.5128 0.0040 0.3988 0.0419
## 244 0.9334 0.0502 0.0025 1.0034 0.0249 1260.1816 0.0029 0.2862 0.0502
## 245 1.3335 0.0415 0.0017 1.0003 0.0248 1259.7402 0.0010 0.0977 0.0415
## 246 0.0575 0.0053 0.0000 1.0088 0.0250 1261.7538 0.0043 0.4266 0.0053
## 247 0.1741 0.0107 0.0001 1.0062 0.0250 1261.7469 0.0030 0.3042 0.0107
## 248 0.1610 0.0121 0.0001 1.0088 0.0250 1261.7466 0.0043 0.4297 0.0121
```

*# Statistics*

```
cooks.distance.rma.uni(rma(effect_size,
                           var,
                           data = bflpedat,
                           method = "REML"), 4)
```

```
##          1          2          3          4          5          6
## 3.424865e-03 7.437663e-05 6.965572e-04 4.607171e-03 2.646992e-03 9.815705e-05
##          7          8          9         10         11         12
## 3.883290e-05 2.526729e-03 1.245043e-03 9.372300e-04 4.174962e-02 2.517461e-03
##         13         14         15         16         17         18
## 7.747812e-04 3.505497e-03 2.431130e-03 2.141033e-03 6.548330e-03 1.482215e-04
##         19         20         21         22         23         24
## 2.578376e-04 3.629684e-03 3.493561e-04 2.222701e-03 1.208759e-04 4.554880e-03
##         25         26         27         28         29         30
## 3.070258e-05 2.012969e-03 1.244084e-05 3.109297e-03 1.169823e-03 2.258075e-03
##         31         32         33         34         35         36
## 1.061485e-03 2.371151e-04 6.402034e-03 5.451367e-04 1.202784e-02 7.784439e-03
##         37         38         39         40         41         42
## 5.305308e-03 1.434878e-05 6.371951e-05 2.364683e-03 6.932601e-04 1.108433e-03
##         43         44         45         46         47         48
## 3.109633e-03 2.672519e-03 3.328154e-03 2.310798e-03 4.017341e-05 3.530741e-03
##         49         50         51         52         53         54
## 1.800336e-04 3.274736e-04 1.934683e-05 5.480045e-03 3.191007e-06 2.100540e-04
##         55         56         57         58         59         60
## 3.910823e-03 2.645723e-04 3.699370e-04 9.967371e-03 2.344663e-03 1.224275e-02
##         61         62         63         64         65         66
## 1.098676e-03 4.163422e-05 5.810244e-05 1.847039e-02 6.480808e-03 5.550275e-04
##         67         68         69         70         71         72
## 1.201819e-02 1.631333e-03 1.238564e-03 5.729988e-03 1.014807e-04 2.841327e-03
##         73         74         75         76         77         78
## 1.917187e-04 1.680663e-03 1.412356e-03 1.413786e-03 5.492482e-03 3.297446e-04
##         79         80         81         82         83         84
## 3.407982e-04 3.597351e-03 4.434005e-05 1.984652e-04 1.149551e-03 5.384234e-04
##         85         86         87         88         89         90
## 2.545660e-03 4.376347e-03 5.935219e-04 1.078367e-02 1.763398e-03 1.329112e-02
##         91         92         93         94         95         96
## 7.249373e-04 6.482850e-03 4.430981e-03 1.351413e-02 5.599028e-03 7.840128e-03
##         97         98         99        100        101        102
## 4.372208e-04 1.844966e-03 2.707044e-05 2.741402e-03 7.609470e-03 1.181020e-02
```

##	103	104	105	106	107	108
##	1.490585e-03	5.497935e-04	7.936874e-03	9.849474e-03	8.155018e-04	4.399594e-03
##	109	110	111	112	113	114
##	7.022256e-04	1.869894e-02	1.355315e-03	1.127282e-02	5.080675e-03	1.889249e-05
##	115	116	117	118	119	120
##	4.275551e-02	1.405593e-04	6.087036e-03	2.800944e-03	1.609608e-02	9.073416e-04
##	121	122	123	124	125	126
##	1.642267e-03	1.264070e-02	5.879516e-03	1.652630e-03	2.111095e-04	4.167301e-05
##	127	128	129	130	131	132
##	3.317084e-04	2.309360e-07	1.349284e-04	1.585109e-02	2.420456e-03	1.924888e-03
##	133	134	135	136	137	138
##	1.594048e-04	5.315541e-03	3.184120e-04	1.176377e-03	3.968582e-03	5.089477e-04
##	139	140	141	142	143	144
##	1.965748e-03	9.001916e-04	5.558291e-05	7.421455e-03	3.283065e-04	1.068640e-03
##	145	146	147	148	149	150
##	1.612775e-05	8.856360e-03	8.955669e-03	2.165234e-02	1.121005e-03	3.986169e-03
##	151	152	153	154	155	156
##	5.505581e-05	1.870682e-02	6.772849e-04	1.119656e-04	5.517455e-03	3.042581e-03
##	157	158	159	160	161	162
##	8.063370e-03	6.164440e-04	2.607221e-04	5.444614e-03	2.048417e-02	7.975444e-03
##	163	164	165	166	167	168
##	1.996028e-03	2.564049e-05	6.070683e-03	1.460885e-03	2.743188e-03	7.506265e-03
##	169	170	171	172	173	174
##	1.402585e-04	9.211682e-04	1.824394e-03	7.753266e-04	1.713565e-04	7.715572e-03
##	175	176	177	178	179	180
##	5.263698e-03	4.156003e-03	3.231517e-04	1.558018e-04	9.172982e-06	3.954844e-03
##	181	182	183	184	185	186
##	1.033701e-04	1.161972e-05	3.361142e-03	2.344246e-03	1.704633e-03	3.217134e-02
##	187	188	189	190	191	192
##	2.659026e-04	1.730803e-02	5.350646e-03	1.141226e-02	1.213546e-02	2.476961e-04
##	193	194	195	196	197	198
##	9.835130e-04	1.593734e-03	1.198827e-03	1.011350e-03	5.100440e-04	9.684953e-03
##	199	200	201	202	203	204
##	3.378581e-03	5.049359e-03	4.068030e-02	5.419414e-03	3.322582e-03	6.795763e-04
##	205	206	207	208	209	210
##	3.004450e-03	6.113311e-06	2.224638e-03	1.646550e-03	1.025276e-04	2.115665e-04
##	211	212	213	214	215	216
##	7.160204e-05	7.435247e-03	2.378536e-03	1.336927e-04	2.623745e-03	5.805181e-04
##	217	218	219	220	221	222
##	4.170255e-03	4.026656e-06	5.017624e-03	1.041519e-04	2.149033e-03	1.164531e-05
##	223	224	225	226	227	228
##	5.803704e-03	9.229630e-05	7.713607e-03	3.105099e-04	1.151926e-03	9.084439e-03
##	229	230	231	232	233	234
##	4.188645e-04	1.460940e-02	1.262546e-02	1.907538e-03	3.045942e-03	9.608799e-06
##	235	236	237	238	239	240
##	8.519000e-05	6.570893e-03	1.039955e-02	2.376143e-04	3.907651e-02	1.034672e-03
##	241	242	243	244	245	246
##	2.693184e-03	1.998917e-04	1.756664e-03	2.521346e-03	1.717866e-03	2.826617e-05

```
##          247          248
## 1.149175e-04 1.470180e-04
```

```
dfbetas.rma.uni(rma(effect_size,
                    var,
                    data = bflpedat,
                    method = "REML"), 4)
```

```
##
## intrcpt
## 1 -0.0585
## 2 0.0086
## 3 -0.0264
## 4 -0.0679
## 5 0.0514
## 6 -0.0099
## 7 0.0062
## 8 0.0502
## 9 -0.0352
## 10 0.0306
## 11 0.2093
## 12 -0.0501
## 13 0.0278
## 14 -0.0592
## 15 0.0492
## 16 0.0462
## 17 0.0810
## 18 0.0121
## 19 -0.0160
## 20 0.0602
## 21 0.0187
## 22 0.0471
## 23 -0.0110
## 24 0.0675
## 25 0.0055
## 26 -0.0448
## 27 -0.0035
## 28 0.0557
## 29 -0.0341
## 30 0.0475
## 31 -0.0325
## 32 -0.0154
## 33 0.0801
## 34 0.0233
## 35 0.1101
## 36 0.0884
## 37 -0.0729
## 38 -0.0038
```

## 39 -0.0080  
## 40 -0.0486  
## 41 0.0263  
## 42 0.0332  
## 43 0.0557  
## 44 0.0516  
## 45 0.0576  
## 46 -0.0480  
## 47 0.0063  
## 48 0.0595  
## 49 0.0134  
## 50 0.0181  
## 51 0.0044  
## 52 -0.0741  
## 53 0.0018  
## 54 -0.0145  
## 55 0.0625  
## 56 -0.0162  
## 57 0.0192  
## 58 -0.1002  
## 59 0.0484  
## 60 -0.1111  
## 61 0.0331  
## 62 0.0064  
## 63 0.0076  
## 64 -0.1369  
## 65 -0.0806  
## 66 -0.0235  
## 67 -0.1100  
## 68 -0.0403  
## 69 -0.0351  
## 70 -0.0757  
## 71 -0.0101  
## 72 0.0533  
## 73 0.0138  
## 74 -0.0409  
## 75 0.0375  
## 76 0.0375  
## 77 0.0741  
## 78 -0.0181  
## 79 0.0184  
## 80 0.0599  
## 81 0.0066  
## 82 -0.0141  
## 83 0.0338  
## 84 -0.0232  
## 85 0.0504  
## 86 -0.0662

## 87 0.0243  
## 88 -0.1042  
## 89 0.0419  
## 90 -0.1159  
## 91 -0.0269  
## 92 0.0806  
## 93 -0.0666  
## 94 0.1168  
## 95 0.0749  
## 96 -0.0887  
## 97 -0.0209  
## 98 0.0429  
## 99 -0.0052  
## 100 0.0523  
## 101 -0.0874  
## 102 0.1092  
## 103 0.0385  
## 104 0.0234  
## 105 -0.0893  
## 106 -0.0995  
## 107 0.0285  
## 108 0.0663  
## 109 -0.0265  
## 110 -0.1377  
## 111 0.0368  
## 112 -0.1066  
## 113 0.0713  
## 114 0.0043  
## 115 -0.2107  
## 116 -0.0118  
## 117 0.0781  
## 118 0.0529  
## 119 0.1277  
## 120 0.0301  
## 121 -0.0405  
## 122 -0.1129  
## 123 -0.0767  
## 124 0.0406  
## 125 -0.0145  
## 126 0.0064  
## 127 -0.0182  
## 128 0.0005  
## 129 -0.0116  
## 130 -0.1266  
## 131 0.0491  
## 132 0.0438  
## 133 -0.0126  
## 134 -0.0730

## 135 0.0178  
## 136 -0.0342  
## 137 0.0630  
## 138 -0.0225  
## 139 -0.0443  
## 140 -0.0300  
## 141 -0.0074  
## 142 -0.0863  
## 143 0.0181  
## 144 -0.0326  
## 145 0.0040  
## 146 0.0943  
## 147 -0.0949  
## 148 0.1486  
## 149 0.0334  
## 150 -0.0631  
## 151 0.0074  
## 152 0.1379  
## 153 0.0260  
## 154 -0.0106  
## 155 0.0743  
## 156 -0.0551  
## 157 0.0900  
## 158 0.0248  
## 159 -0.0161  
## 160 -0.0739  
## 161 -0.1443  
## 162 0.0895  
## 163 0.0446  
## 164 0.0051  
## 165 -0.0780  
## 166 0.0382  
## 167 -0.0523  
## 168 0.0868  
## 169 -0.0118  
## 170 0.0303  
## 171 0.0426  
## 172 -0.0278  
## 173 -0.0131  
## 174 0.0881  
## 175 0.0727  
## 176 -0.0645  
## 177 0.0179  
## 178 -0.0125  
## 179 0.0030  
## 180 0.0629  
## 181 0.0101  
## 182 0.0034

## 183 -0.0580  
## 184 0.0484  
## 185 0.0412  
## 186 -0.1819  
## 187 -0.0163  
## 188 -0.1324  
## 189 0.0732  
## 190 -0.1072  
## 191 0.1106  
## 192 -0.0157  
## 193 0.0313  
## 194 -0.0399  
## 195 -0.0346  
## 196 -0.0318  
## 197 -0.0225  
## 198 0.0987  
## 199 -0.0581  
## 200 0.0711  
## 201 -0.2053  
## 202 0.0736  
## 203 -0.0577  
## 204 0.0260  
## 205 0.0548  
## 206 -0.0025  
## 207 0.0471  
## 208 0.0405  
## 209 -0.0101  
## 210 0.0145  
## 211 -0.0084  
## 212 -0.0864  
## 213 -0.0488  
## 214 -0.0115  
## 215 0.0512  
## 216 -0.0241  
## 217 -0.0646  
## 218 -0.0020  
## 219 0.0709  
## 220 -0.0102  
## 221 -0.0463  
## 222 0.0034  
## 223 0.0762  
## 224 0.0096  
## 225 0.0880  
## 226 0.0176  
## 227 0.0339  
## 228 -0.0956  
## 229 -0.0204  
## 230 -0.1215

```

## 231 0.1129
## 232 -0.0436
## 233 -0.0552
## 234 -0.0031
## 235 0.0092
## 236 0.0811
## 237 -0.1023
## 238 0.0154
## 239 -0.2010
## 240 0.0321
## 241 0.0519
## 242 0.0141
## 243 0.0419
## 244 0.0502
## 245 0.0415
## 246 0.0053
## 247 0.0107
## 248 0.0121

```

```

hatvalues.rma.uni(rma(effect_size,
                      var,
                      data = bflpedat,
                      method = "REML"), type="diagonal")

```

```

##          1          2          3          4          5          6
## 0.0039976817 0.0046381325 0.0037188458 0.0047636536 0.0043704690 0.0047023241
##          7          8          9         10         11         12
## 0.0046728803 0.0042068236 0.0044029440 0.0033427716 0.0041935256 0.0041812410
##          13         14         15         16         17         18
## 0.0049733863 0.0044817284 0.0047810800 0.0043725122 0.0046805798 0.0044526043
##          19         20         21         22         23         24
## 0.0037124862 0.0045347945 0.0039431337 0.0044563850 0.0041676686 0.0043839321
##          25         26         27         28         29         30
## 0.0044466799 0.0035366034 0.0045289860 0.0041115428 0.0046298785 0.0047237703
##          31         32         33         34         35         36
## 0.0031587938 0.0043643654 0.0046959757 0.0041650091 0.0046457608 0.0040285689
##          37         38         39         40         41         42
## 0.0039367293 0.0043690815 0.0030316757 0.0038271338 0.0042999148 0.0046547290
##          43         44         45         46         47         48
## 0.0047435427 0.0047017836 0.0045754563 0.0045519271 0.0032524347 0.0015457364
##          49         50         51         52         53         54
## 0.0029920816 0.0046806655 0.0044759014 0.0039606155 0.0048661296 0.0044437237
##          55         56         57         58         59         60
## 0.0033342547 0.0042859959 0.0039139711 0.0035973026 0.0044340856 0.0043181491
##          61         62         63         64         65         66
## 0.0041769990 0.0048392786 0.0043007241 0.0034456613 0.0033112046 0.0045073648
##          67         68         69         70         71         72
## 0.0046052495 0.0038696413 0.0047053444 0.0046246434 0.0038206505 0.0038540374

```

##	73	74	75	76	77	78
##	0.0047618020	0.0042144523	0.0045437996	0.0040758682	0.0045462093	0.0047810274
##	79	80	81	82	83	84
##	0.0046698431	0.0046034135	0.0044119232	0.0043201994	0.0040943414	0.0045961420
##	85	86	87	88	89	90
##	0.0033337776	0.0039710281	0.0046363294	0.0038477736	0.0044334995	0.0035923404
##	91	92	93	94	95	96
##	0.0045978767	0.0046338216	0.0039525263	0.0046384375	0.0032392082	0.0045756248
##	97	98	99	100	101	102
##	0.0034747971	0.0045781776	0.0042632021	0.0047630067	0.0043485190	0.0038773080
##	103	104	105	106	107	108
##	0.0041474862	0.0045361678	0.0040224880	0.0041417958	0.0027743622	0.0045336442
##	109	110	111	112	113	114
##	0.0037810161	0.0040508174	0.0042281727	0.0043356510	0.0045844852	0.0048589931
##	115	116	117	118	119	120
##	0.0033152600	0.0044922924	0.0033081738	0.0026437862	0.0045635448	0.0045379064
##	121	122	123	124	125	126
##	0.0031778666	0.0040956061	0.0044620591	0.0043700210	0.0045789922	0.0045415189
##	127	128	129	130	131	132
##	0.0038886468	0.0046598926	0.0034726514	0.0044984693	0.0043681936	0.0045929353
##	133	134	135	136	137	138
##	0.0034958195	0.0033275368	0.0042317712	0.0042815612	0.0042030941	0.0043180105
##	139	140	141	142	143	144
##	0.0035454545	0.0031567227	0.0042567664	0.0037278775	0.0045828475	0.0038285129
##	145	146	147	148	149	150
##	0.0044877697	0.0046250959	0.0034282150	0.0048724645	0.0031256789	0.0045257043
##	151	152	153	154	155	156
##	0.0017929731	0.0042762316	0.0043110308	0.0042998071	0.0041257624	0.0044224812
##	157	158	159	160	161	162
##	0.0042836164	0.0047258286	0.0026128507	0.0033527961	0.0041934474	0.0037208502
##	163	164	165	166	167	168
##	0.0044755436	0.0037946453	0.0042020180	0.0044811250	0.0040586161	0.0045710289
##	169	170	171	172	173	174
##	0.0048096814	0.0036230448	0.0045946271	0.0036056632	0.0034547668	0.0030304153
##	175	176	177	178	179	180
##	0.0017977758	0.0031730209	0.0032016902	0.0040503748	0.0047999390	0.0040791380
##	181	182	183	184	185	186
##	0.0046015139	0.0042230284	0.0035107949	0.0042977987	0.0045234433	0.0035207043
##	187	188	189	190	191	192
##	0.0043992518	0.0041636573	0.0039677109	0.0042977081	0.0043883660	0.0026613124
##	193	194	195	196	197	198
##	0.0028341144	0.0045201523	0.0039690736	0.0014625899	0.0044745137	0.0043401811
##	199	200	201	202	203	204
##	0.0047221666	0.0037819696	0.0038444823	0.0045389246	0.0017646443	0.0042568928
##	205	206	207	208	209	210
##	0.0047305045	0.0048093836	0.0044583555	0.0047131888	0.0024490141	0.0002960950
##	211	212	213	214	215	216
##	0.0043868639	0.0027963330	0.0023131570	0.0039820388	0.0042271398	0.0019296753

```

##          217          218          219          220          221          222
## 0.0037972472 0.0039079676 0.0032159900 0.0036276484 0.0038693578 0.0047467095
##          223          224          225          226          227          228
## 0.0046277387 0.0047225785 0.0046770262 0.0043595545 0.0043739634 0.0042571623
##          229          230          231          232          233          234
## 0.0025707032 0.0041932830 0.00444492851 0.0040077178 0.0042716675 0.0025465094
##          235          236          237          238          239          240
## 0.0037311610 0.0044280090 0.0042667307 0.0029750117 0.0041015711 0.0038419352
##          241          242          243          244          245          246
## 0.0040805503 0.0048628922 0.0039876359 0.0028621223 0.0009770115 0.0042659969
##          247          248
## 0.0030417899 0.0042965249

```

```

# Leave-one-out analyses
leavelout(rma(effect_size,
              var,
              data = bflpedat,
              method = "REML"))

```

```

##
## estimate se zval pval ci.lb ci.ub Q Qp tau2
## 1 -0.4574 0.0114 -39.9524 0.0000 -0.4798 -0.4349 1257.3521 0.0000 0.0249
## 2 -0.4581 0.0115 -39.9174 0.0000 -0.4806 -0.4356 1261.7647 0.0000 0.0250
## 3 -0.4577 0.0115 -39.9372 0.0000 -0.4802 -0.4353 1260.7458 0.0000 0.0250
## 4 -0.4573 0.0114 -39.9367 0.0000 -0.4797 -0.4348 1248.8334 0.0000 0.0249
## 5 -0.4586 0.0115 -40.0218 0.0000 -0.4811 -0.4362 1259.1472 0.0000 0.0249
## 6 -0.4579 0.0115 -39.8993 0.0000 -0.4804 -0.4354 1261.0269 0.0000 0.0250
## 7 -0.4581 0.0115 -39.9132 0.0000 -0.4806 -0.4356 1261.7490 0.0000 0.0250
## 8 -0.4586 0.0115 -40.0249 0.0000 -0.4811 -0.4362 1259.5680 0.0000 0.0249
## 9 -0.4576 0.0115 -39.9128 0.0000 -0.4801 -0.4352 1259.1251 0.0000 0.0250
## 10 -0.4584 0.0115 -40.0088 0.0000 -0.4808 -0.4359 1261.2775 0.0000 0.0249
## 11 -0.4604 0.0112 -41.1951 0.0000 -0.4823 -0.4385 1207.3691 0.0000 0.0234
## 12 -0.4575 0.0115 -39.9343 0.0000 -0.4799 -0.4350 1257.9161 0.0000 0.0249
## 13 -0.4584 0.0115 -39.9352 0.0000 -0.4809 -0.4359 1260.4548 0.0000 0.0250
## 14 -0.4574 0.0115 -39.9339 0.0000 -0.4798 -0.4349 1255.0120 0.0000 0.0249
## 15 -0.4586 0.0115 -39.9979 0.0000 -0.4811 -0.4361 1257.8482 0.0000 0.0249
## 16 -0.4586 0.0115 -40.0059 0.0000 -0.4810 -0.4361 1259.7448 0.0000 0.0249
## 17 -0.4590 0.0114 -40.1230 0.0000 -0.4814 -0.4365 1250.9250 0.0000 0.0248
## 18 -0.4582 0.0115 -39.9297 0.0000 -0.4807 -0.4357 1261.7464 0.0000 0.0250
## 19 -0.4579 0.0115 -39.9375 0.0000 -0.4803 -0.4354 1261.2742 0.0000 0.0250
## 20 -0.4587 0.0115 -40.0445 0.0000 -0.4812 -0.4363 1257.2788 0.0000 0.0249
## 21 -0.4583 0.0115 -39.9607 0.0000 -0.4807 -0.4358 1261.6398 0.0000 0.0250
## 22 -0.4586 0.0115 -40.0050 0.0000 -0.4810 -0.4361 1259.4865 0.0000 0.0249
## 23 -0.4579 0.0115 -39.9206 0.0000 -0.4804 -0.4354 1261.3445 0.0000 0.0250
## 24 -0.4588 0.0114 -40.0781 0.0000 -0.4812 -0.4364 1256.7076 0.0000 0.0249
## 25 -0.4581 0.0115 -39.9217 0.0000 -0.4806 -0.4356 1261.7507 0.0000 0.0250
## 26 -0.4575 0.0115 -39.9549 0.0000 -0.4800 -0.4351 1259.5262 0.0000 0.0249
## 27 -0.4580 0.0115 -39.9101 0.0000 -0.4805 -0.4355 1261.5207 0.0000 0.0250

```

## 28	-0.4587	0.0115	-40.0467	0.0000	-0.4811	-0.4362	1259.1020	0.0000	0.0249
## 29	-0.4576	0.0115	-39.9030	0.0000	-0.4801	-0.4352	1258.4224	0.0000	0.0250
## 30	-0.4586	0.0115	-39.9949	0.0000	-0.4811	-0.4361	1258.5318	0.0000	0.0250
## 31	-0.4577	0.0115	-39.9620	0.0000	-0.4801	-0.4352	1260.6059	0.0000	0.0249
## 32	-0.4579	0.0115	-39.9111	0.0000	-0.4803	-0.4354	1260.9988	0.0000	0.0250
## 33	-0.4590	0.0114	-40.1182	0.0000	-0.4814	-0.4365	1250.9528	0.0000	0.0248
## 34	-0.4583	0.0115	-39.9597	0.0000	-0.4808	-0.4358	1261.4905	0.0000	0.0250
## 35	-0.4593	0.0114	-40.2798	0.0000	-0.4816	-0.4369	1241.1132	0.0000	0.0246
## 36	-0.4590	0.0114	-40.1863	0.0000	-0.4814	-0.4367	1254.4008	0.0000	0.0247
## 37	-0.4572	0.0114	-39.9794	0.0000	-0.4796	-0.4348	1255.5271	0.0000	0.0248
## 38	-0.4580	0.0115	-39.9164	0.0000	-0.4805	-0.4355	1261.5598	0.0000	0.0250
## 39	-0.4579	0.0115	-39.9685	0.0000	-0.4804	-0.4355	1261.6162	0.0000	0.0250
## 40	-0.4575	0.0115	-39.9469	0.0000	-0.4799	-0.4350	1258.8451	0.0000	0.0249
## 41	-0.4583	0.0115	-39.9600	0.0000	-0.4808	-0.4359	1261.3421	0.0000	0.0250
## 42	-0.4584	0.0115	-39.9603	0.0000	-0.4809	-0.4359	1260.6187	0.0000	0.0250
## 43	-0.4587	0.0115	-40.0202	0.0000	-0.4811	-0.4362	1256.8016	0.0000	0.0249
## 44	-0.4586	0.0115	-40.0086	0.0000	-0.4811	-0.4362	1257.9132	0.0000	0.0249
## 45	-0.4587	0.0115	-40.0338	0.0000	-0.4812	-0.4362	1257.5271	0.0000	0.0249
## 46	-0.4575	0.0115	-39.9170	0.0000	-0.4800	-0.4350	1256.5806	0.0000	0.0249
## 47	-0.4581	0.0115	-39.9717	0.0000	-0.4806	-0.4356	1261.7650	0.0000	0.0250
## 48	-0.4587	0.0114	-40.1682	0.0000	-0.4811	-0.4363	1258.7155	0.0000	0.0248
## 49	-0.4582	0.0115	-39.9917	0.0000	-0.4806	-0.4357	1261.7199	0.0000	0.0250
## 50	-0.4582	0.0115	-39.9292	0.0000	-0.4807	-0.4358	1261.6024	0.0000	0.0250
## 51	-0.4581	0.0115	-39.9193	0.0000	-0.4806	-0.4356	1261.7368	0.0000	0.0250
## 52	-0.4572	0.0114	-39.9808	0.0000	-0.4796	-0.4348	1255.2592	0.0000	0.0248
## 53	-0.4581	0.0115	-39.9007	0.0000	-0.4806	-0.4356	1261.5977	0.0000	0.0250
## 54	-0.4579	0.0115	-39.9081	0.0000	-0.4804	-0.4354	1260.9884	0.0000	0.0250
## 55	-0.4588	0.0114	-40.1037	0.0000	-0.4812	-0.4363	1259.0671	0.0000	0.0249
## 56	-0.4579	0.0115	-39.9141	0.0000	-0.4803	-0.4354	1261.0056	0.0000	0.0250
## 57	-0.4583	0.0115	-39.9628	0.0000	-0.4807	-0.4358	1261.6278	0.0000	0.0250
## 58	-0.4569	0.0114	-40.0596	0.0000	-0.4793	-0.4345	1252.3751	0.0000	0.0247
## 59	-0.4586	0.0115	-40.0097	0.0000	-0.4811	-0.4361	1259.3806	0.0000	0.0249
## 60	-0.4568	0.0114	-40.0646	0.0000	-0.4791	-0.4344	1244.6586	0.0000	0.0246
## 61	-0.4584	0.0115	-39.9798	0.0000	-0.4809	-0.4359	1261.0144	0.0000	0.0250
## 62	-0.4581	0.0115	-39.9066	0.0000	-0.4806	-0.4356	1261.7416	0.0000	0.0250
## 63	-0.4581	0.0115	-39.9301	0.0000	-0.4806	-0.4356	1261.7643	0.0000	0.0250
## 64	-0.4565	0.0114	-40.1950	0.0000	-0.4787	-0.4342	1246.0987	0.0000	0.0244
## 65	-0.4571	0.0114	-40.0208	0.0000	-0.4795	-0.4347	1255.9341	0.0000	0.0248
## 66	-0.4578	0.0115	-39.9047	0.0000	-0.4803	-0.4353	1260.1513	0.0000	0.0250
## 67	-0.4568	0.0114	-40.0497	0.0000	-0.4791	-0.4344	1238.6974	0.0000	0.0246
## 68	-0.4576	0.0115	-39.9377	0.0000	-0.4800	-0.4351	1259.5787	0.0000	0.0249
## 69	-0.4576	0.0115	-39.9005	0.0000	-0.4801	-0.4352	1257.8170	0.0000	0.0250
## 70	-0.4572	0.0114	-39.9575	0.0000	-0.4796	-0.4347	1249.3861	0.0000	0.0248
## 71	-0.4579	0.0115	-39.9351	0.0000	-0.4804	-0.4354	1261.4736	0.0000	0.0250
## 72	-0.4586	0.0115	-40.0494	0.0000	-0.4811	-0.4362	1259.6402	0.0000	0.0249
## 73	-0.4582	0.0115	-39.9193	0.0000	-0.4807	-0.4357	1261.7165	0.0000	0.0250
## 74	-0.4576	0.0115	-39.9242	0.0000	-0.4800	-0.4351	1258.9413	0.0000	0.0249
## 75	-0.4585	0.0115	-39.9752	0.0000	-0.4809	-0.4360	1260.3717	0.0000	0.0250

## 76	-0.4585	0.0115	-39.9948	0.0000	-0.4809	-0.4360	1260.7738	0.0000	0.0250
## 77	-0.4589	0.0114	-40.0984	0.0000	-0.4813	-0.4365	1254.3616	0.0000	0.0248
## 78	-0.4578	0.0115	-39.8937	0.0000	-0.4803	-0.4353	1260.0104	0.0000	0.0250
## 79	-0.4582	0.0115	-39.9303	0.0000	-0.4807	-0.4358	1261.5911	0.0000	0.0250
## 80	-0.4587	0.0115	-40.0406	0.0000	-0.4812	-0.4363	1256.9489	0.0000	0.0249
## 81	-0.4581	0.0115	-39.9244	0.0000	-0.4806	-0.4356	1261.7597	0.0000	0.0250
## 82	-0.4579	0.0115	-39.9133	0.0000	-0.4804	-0.4354	1261.1111	0.0000	0.0250
## 83	-0.4584	0.0115	-39.9850	0.0000	-0.4809	-0.4360	1261.0007	0.0000	0.0250
## 84	-0.4578	0.0115	-39.9011	0.0000	-0.4803	-0.4353	1259.9851	0.0000	0.0250
## 85	-0.4586	0.0114	-40.0622	0.0000	-0.4810	-0.4362	1260.1215	0.0000	0.0249
## 86	-0.4573	0.0114	-39.9656	0.0000	-0.4797	-0.4349	1256.3918	0.0000	0.0248
## 87	-0.4583	0.0115	-39.9422	0.0000	-0.4808	-0.4358	1261.3205	0.0000	0.0250
## 88	-0.4569	0.0114	-40.0616	0.0000	-0.4792	-0.4345	1250.5615	0.0000	0.0246
## 89	-0.4585	0.0115	-39.9913	0.0000	-0.4810	-0.4360	1260.0975	0.0000	0.0250
## 90	-0.4567	0.0114	-40.1096	0.0000	-0.4790	-0.4344	1249.6006	0.0000	0.0246
## 91	-0.4577	0.0115	-39.9016	0.0000	-0.4802	-0.4352	1259.5494	0.0000	0.0250
## 92	-0.4590	0.0114	-40.1231	0.0000	-0.4814	-0.4365	1251.7466	0.0000	0.0248
## 93	-0.4573	0.0114	-39.9671	0.0000	-0.4797	-0.4349	1256.3911	0.0000	0.0248
## 94	-0.4594	0.0114	-40.3222	0.0000	-0.4817	-0.4370	1238.4428	0.0000	0.0246
## 95	-0.4589	0.0114	-40.1573	0.0000	-0.4813	-0.4365	1257.7683	0.0000	0.0248
## 96	-0.4570	0.0114	-39.9892	0.0000	-0.4794	-0.4346	1246.5563	0.0000	0.0248
## 97	-0.4578	0.0115	-39.9467	0.0000	-0.4803	-0.4353	1261.1259	0.0000	0.0250
## 98	-0.4585	0.0115	-39.9879	0.0000	-0.4810	-0.4361	1259.7163	0.0000	0.0250
## 99	-0.4580	0.0115	-39.9198	0.0000	-0.4805	-0.4355	1261.5389	0.0000	0.0250
## 100	-0.4586	0.0115	-40.0082	0.0000	-0.4811	-0.4362	1257.3653	0.0000	0.0249
## 101	-0.4570	0.0114	-39.9950	0.0000	-0.4794	-0.4346	1250.2339	0.0000	0.0248
## 102	-0.4593	0.0114	-40.3074	0.0000	-0.4816	-0.4369	1250.7983	0.0000	0.0246
## 103	-0.4585	0.0115	-39.9943	0.0000	-0.4809	-0.4360	1260.6647	0.0000	0.0250
## 104	-0.4583	0.0115	-39.9446	0.0000	-0.4808	-0.4358	1261.4111	0.0000	0.0250
## 105	-0.4570	0.0114	-40.0128	0.0000	-0.4794	-0.4346	1252.3975	0.0000	0.0247
## 106	-0.4569	0.0114	-40.0360	0.0000	-0.4793	-0.4345	1249.4927	0.0000	0.0247
## 107	-0.4584	0.0115	-40.0280	0.0000	-0.4808	-0.4359	1261.3484	0.0000	0.0249
## 108	-0.4588	0.0115	-40.0672	0.0000	-0.4812	-0.4364	1256.1363	0.0000	0.0249
## 109	-0.4577	0.0115	-39.9347	0.0000	-0.4802	-0.4353	1260.7057	0.0000	0.0250
## 110	-0.4565	0.0114	-40.1742	0.0000	-0.4787	-0.4342	1241.2148	0.0000	0.0244
## 111	-0.4585	0.0115	-39.9864	0.0000	-0.4809	-0.4360	1260.7452	0.0000	0.0250
## 112	-0.4568	0.0114	-40.0494	0.0000	-0.4792	-0.4345	1245.6218	0.0000	0.0246
## 113	-0.4589	0.0114	-40.0849	0.0000	-0.4813	-0.4364	1254.6888	0.0000	0.0249
## 114	-0.4581	0.0115	-39.9035	0.0000	-0.4806	-0.4356	1261.6923	0.0000	0.0250
## 115	-0.4557	0.0112	-40.5829	0.0000	-0.4777	-0.4337	1229.0576	0.0000	0.0237
## 116	-0.4579	0.0115	-39.9070	0.0000	-0.4804	-0.4354	1261.1159	0.0000	0.0250
## 117	-0.4589	0.0114	-40.1685	0.0000	-0.4813	-0.4365	1257.3307	0.0000	0.0248
## 118	-0.4586	0.0114	-40.0993	0.0000	-0.4811	-0.4362	1259.9617	0.0000	0.0249
## 119	-0.4595	0.0114	-40.3991	0.0000	-0.4818	-0.4372	1235.9887	0.0000	0.0245
## 120	-0.4584	0.0115	-39.9580	0.0000	-0.4809	-0.4359	1261.0096	0.0000	0.0250
## 121	-0.4576	0.0114	-39.9659	0.0000	-0.4800	-0.4351	1260.0824	0.0000	0.0249
## 122	-0.4568	0.0114	-40.0796	0.0000	-0.4791	-0.4344	1246.8971	0.0000	0.0246
## 123	-0.4572	0.0114	-39.9661	0.0000	-0.4796	-0.4347	1251.4504	0.0000	0.0248

## 124	-0.4585	0.0115	-39.9904	0.0000	-0.4810	-0.4360	1260.3101	0.0000	0.0250
## 125	-0.4579	0.0115	-39.9026	0.0000	-0.4804	-0.4354	1260.8342	0.0000	0.0250
## 126	-0.4581	0.0115	-39.9188	0.0000	-0.4806	-0.4356	1261.7557	0.0000	0.0250
## 127	-0.4578	0.0115	-39.9300	0.0000	-0.4803	-0.4354	1261.1167	0.0000	0.0250
## 128	-0.4580	0.0115	-39.9079	0.0000	-0.4805	-0.4355	1261.6239	0.0000	0.0250
## 129	-0.4579	0.0115	-39.9487	0.0000	-0.4804	-0.4354	1261.4816	0.0000	0.0250
## 130	-0.4566	0.0114	-40.1123	0.0000	-0.4789	-0.4343	1235.9697	0.0000	0.0245
## 131	-0.4586	0.0115	-40.0149	0.0000	-0.4811	-0.4361	1259.4227	0.0000	0.0249
## 132	-0.4585	0.0115	-39.9899	0.0000	-0.4810	-0.4361	1259.5642	0.0000	0.0250
## 133	-0.4579	0.0115	-39.9474	0.0000	-0.4804	-0.4354	1261.4463	0.0000	0.0250
## 134	-0.4572	0.0114	-40.0041	0.0000	-0.4796	-0.4348	1256.8643	0.0000	0.0248
## 135	-0.4582	0.0115	-39.9473	0.0000	-0.4807	-0.4358	1261.6509	0.0000	0.0250
## 136	-0.4576	0.0115	-39.9172	0.0000	-0.4801	-0.4352	1259.5220	0.0000	0.0250
## 137	-0.4588	0.0114	-40.0686	0.0000	-0.4812	-0.4363	1258.0104	0.0000	0.0249
## 138	-0.4578	0.0115	-39.9123	0.0000	-0.4803	-0.4353	1260.5356	0.0000	0.0250
## 139	-0.4575	0.0115	-39.9541	0.0000	-0.4800	-0.4351	1259.5636	0.0000	0.0249
## 140	-0.4577	0.0115	-39.9610	0.0000	-0.4801	-0.4352	1260.7538	0.0000	0.0249
## 141	-0.4580	0.0115	-39.9187	0.0000	-0.4804	-0.4355	1261.4596	0.0000	0.0250
## 142	-0.4571	0.0114	-40.0173	0.0000	-0.4794	-0.4347	1254.1811	0.0000	0.0247
## 143	-0.4582	0.0115	-39.9333	0.0000	-0.4807	-0.4358	1261.6168	0.0000	0.0250
## 144	-0.4577	0.0115	-39.9348	0.0000	-0.4801	-0.4352	1260.2517	0.0000	0.0250
## 145	-0.4581	0.0115	-39.9184	0.0000	-0.4806	-0.4356	1261.7308	0.0000	0.0250
## 146	-0.4591	0.0114	-40.1910	0.0000	-0.4815	-0.4367	1247.5963	0.0000	0.0247
## 147	-0.4570	0.0114	-40.0515	0.0000	-0.4793	-0.4346	1253.6997	0.0000	0.0247
## 148	-0.4597	0.0113	-40.5453	0.0000	-0.4819	-0.4375	1200.6701	0.0000	0.0243
## 149	-0.4584	0.0115	-40.0244	0.0000	-0.4809	-0.4360	1261.1593	0.0000	0.0249
## 150	-0.4573	0.0115	-39.9382	0.0000	-0.4798	-0.4349	1253.8261	0.0000	0.0249
## 151	-0.4581	0.0114	-40.0334	0.0000	-0.4806	-0.4357	1261.7519	0.0000	0.0249
## 152	-0.4596	0.0114	-40.4871	0.0000	-0.4818	-0.4374	1238.5961	0.0000	0.0243
## 153	-0.4583	0.0115	-39.9589	0.0000	-0.4808	-0.4359	1261.3536	0.0000	0.0250
## 154	-0.4579	0.0115	-39.9154	0.0000	-0.4804	-0.4354	1261.3078	0.0000	0.0250
## 155	-0.4589	0.0114	-40.1171	0.0000	-0.4813	-0.4365	1256.5178	0.0000	0.0248
## 156	-0.4574	0.0115	-39.9307	0.0000	-0.4799	-0.4350	1256.1540	0.0000	0.0249
## 157	-0.4591	0.0114	-40.1832	0.0000	-0.4815	-0.4367	1252.7590	0.0000	0.0247
## 158	-0.4583	0.0115	-39.9394	0.0000	-0.4808	-0.4358	1261.2335	0.0000	0.0250
## 159	-0.4579	0.0115	-39.9825	0.0000	-0.4803	-0.4354	1261.4274	0.0000	0.0249
## 160	-0.4572	0.0114	-40.0048	0.0000	-0.4796	-0.4348	1256.7233	0.0000	0.0248
## 161	-0.4564	0.0114	-40.1962	0.0000	-0.4787	-0.4341	1237.2252	0.0000	0.0244
## 162	-0.4591	0.0114	-40.2050	0.0000	-0.4814	-0.4367	1255.1225	0.0000	0.0247
## 163	-0.4585	0.0115	-39.9970	0.0000	-0.4810	-0.4361	1259.7363	0.0000	0.0250
## 164	-0.4581	0.0115	-39.9479	0.0000	-0.4806	-0.4356	1261.7599	0.0000	0.0250
## 165	-0.4571	0.0114	-39.9792	0.0000	-0.4796	-0.4347	1253.4100	0.0000	0.0248
## 166	-0.4585	0.0115	-39.9794	0.0000	-0.4809	-0.4360	1260.3983	0.0000	0.0250
## 167	-0.4574	0.0115	-39.9418	0.0000	-0.4799	-0.4350	1257.9725	0.0000	0.0249
## 168	-0.4590	0.0114	-40.1550	0.0000	-0.4814	-0.4366	1250.8266	0.0000	0.0248
## 169	-0.4579	0.0115	-39.8941	0.0000	-0.4804	-0.4354	1260.6519	0.0000	0.0250
## 170	-0.4584	0.0115	-39.9965	0.0000	-0.4808	-0.4359	1261.2722	0.0000	0.0250
## 171	-0.4585	0.0115	-39.9866	0.0000	-0.4810	-0.4360	1259.7055	0.0000	0.0250

## 172	-0.4577	0.0115	-39.9421	0.0000	-0.4802	-0.4353	1260.7144	0.0000	0.0250
## 173	-0.4579	0.0115	-39.9489	0.0000	-0.4804	-0.4354	1261.4374	0.0000	0.0250
## 174	-0.4590	0.0114	-40.2274	0.0000	-0.4814	-0.4367	1256.1627	0.0000	0.0247
## 175	-0.4589	0.0114	-40.2092	0.0000	-0.4812	-0.4365	1257.4990	0.0000	0.0247
## 176	-0.4573	0.0114	-39.9950	0.0000	-0.4797	-0.4349	1257.9778	0.0000	0.0248
## 177	-0.4582	0.0115	-39.9902	0.0000	-0.4807	-0.4358	1261.6535	0.0000	0.0250
## 178	-0.4579	0.0115	-39.9248	0.0000	-0.4804	-0.4354	1261.3222	0.0000	0.0250
## 179	-0.4581	0.0115	-39.9046	0.0000	-0.4806	-0.4356	1261.6676	0.0000	0.0250
## 180	-0.4588	0.0114	-40.0734	0.0000	-0.4812	-0.4363	1258.2915	0.0000	0.0249
## 181	-0.4582	0.0115	-39.9209	0.0000	-0.4806	-0.4357	1261.7623	0.0000	0.0250
## 182	-0.4581	0.0115	-39.9286	0.0000	-0.4806	-0.4356	1261.7370	0.0000	0.0250
## 183	-0.4574	0.0114	-39.9713	0.0000	-0.4798	-0.4349	1258.3103	0.0000	0.0249
## 184	-0.4586	0.0115	-40.0154	0.0000	-0.4811	-0.4361	1259.6321	0.0000	0.0249
## 185	-0.4585	0.0115	-39.9857	0.0000	-0.4810	-0.4360	1260.0229	0.0000	0.0250
## 186	-0.4560	0.0113	-40.4069	0.0000	-0.4781	-0.4339	1235.1741	0.0000	0.0240
## 187	-0.4579	0.0115	-39.9095	0.0000	-0.4803	-0.4354	1260.9080	0.0000	0.0250
## 188	-0.4565	0.0114	-40.1482	0.0000	-0.4788	-0.4342	1241.1185	0.0000	0.0245
## 189	-0.4589	0.0114	-40.1190	0.0000	-0.4813	-0.4365	1257.1059	0.0000	0.0248
## 190	-0.4568	0.0114	-40.0530	0.0000	-0.4792	-0.4345	1245.9786	0.0000	0.0246
## 191	-0.4593	0.0114	-40.2942	0.0000	-0.4816	-0.4370	1246.1562	0.0000	0.0246
## 192	-0.4579	0.0115	-39.9806	0.0000	-0.4803	-0.4354	1261.4372	0.0000	0.0249
## 193	-0.4584	0.0115	-40.0317	0.0000	-0.4808	-0.4360	1261.2433	0.0000	0.0249
## 194	-0.4576	0.0115	-39.9110	0.0000	-0.4801	-0.4351	1258.0887	0.0000	0.0250
## 195	-0.4576	0.0115	-39.9300	0.0000	-0.4801	-0.4352	1259.9634	0.0000	0.0250
## 196	-0.4577	0.0114	-40.0309	0.0000	-0.4801	-0.4353	1260.6570	0.0000	0.0249
## 197	-0.4578	0.0115	-39.9060	0.0000	-0.4803	-0.4353	1260.3064	0.0000	0.0250
## 198	-0.4592	0.0114	-40.2268	0.0000	-0.4815	-0.4368	1250.1655	0.0000	0.0247
## 199	-0.4574	0.0115	-39.9227	0.0000	-0.4798	-0.4349	1252.5599	0.0000	0.0249
## 200	-0.4588	0.0114	-40.1182	0.0000	-0.4813	-0.4364	1257.7263	0.0000	0.0248
## 201	-0.4557	0.0112	-40.5286	0.0000	-0.4778	-0.4337	1224.6105	0.0000	0.0238
## 202	-0.4589	0.0114	-40.0966	0.0000	-0.4813	-0.4364	1254.5367	0.0000	0.0248
## 203	-0.4574	0.0114	-40.0416	0.0000	-0.4798	-0.4350	1258.7124	0.0000	0.0248
## 204	-0.4583	0.0115	-39.9612	0.0000	-0.4808	-0.4359	1261.3646	0.0000	0.0250
## 205	-0.4587	0.0115	-40.0175	0.0000	-0.4811	-0.4362	1257.1088	0.0000	0.0249
## 206	-0.4580	0.0115	-39.8994	0.0000	-0.4805	-0.4355	1261.4166	0.0000	0.0250
## 207	-0.4586	0.0115	-40.0050	0.0000	-0.4810	-0.4361	1259.4798	0.0000	0.0249
## 208	-0.4585	0.0115	-39.9759	0.0000	-0.4810	-0.4360	1259.6432	0.0000	0.0250
## 209	-0.4579	0.0115	-39.9914	0.0000	-0.4804	-0.4355	1261.5975	0.0000	0.0249
## 210	-0.4582	0.0114	-40.1055	0.0000	-0.4806	-0.4358	1261.0609	0.0000	0.0249
## 211	-0.4579	0.0115	-39.9128	0.0000	-0.4804	-0.4355	1261.3671	0.0000	0.0250
## 212	-0.4571	0.0114	-40.0550	0.0000	-0.4794	-0.4347	1255.7146	0.0000	0.0247
## 213	-0.4575	0.0114	-40.0085	0.0000	-0.4799	-0.4351	1259.6538	0.0000	0.0249
## 214	-0.4579	0.0115	-39.9279	0.0000	-0.4804	-0.4354	1261.3817	0.0000	0.0250
## 215	-0.4586	0.0115	-40.0271	0.0000	-0.4811	-0.4362	1259.4350	0.0000	0.0249
## 216	-0.4578	0.0114	-40.0100	0.0000	-0.4802	-0.4353	1261.1435	0.0000	0.0249
## 217	-0.4573	0.0114	-39.9699	0.0000	-0.4797	-0.4349	1257.0681	0.0000	0.0248
## 218	-0.4580	0.0115	-39.9366	0.0000	-0.4805	-0.4355	1261.6730	0.0000	0.0250
## 219	-0.4588	0.0114	-40.1414	0.0000	-0.4812	-0.4364	1258.2428	0.0000	0.0248

## 220	-0.4579	0.0115	-39.9430	0.0000	-0.4804	-0.4355	1261.5022	0.0000	0.0250
## 221	-0.4575	0.0115	-39.9429	0.0000	-0.4800	-0.4351	1259.0119	0.0000	0.0249
## 222	-0.4581	0.0115	-39.9072	0.0000	-0.4806	-0.4356	1261.6901	0.0000	0.0250
## 223	-0.4589	0.0114	-40.1039	0.0000	-0.4813	-0.4365	1253.0259	0.0000	0.0248
## 224	-0.4581	0.0115	-39.9152	0.0000	-0.4806	-0.4357	1261.7646	0.0000	0.0250
## 225	-0.4590	0.0114	-40.1563	0.0000	-0.4814	-0.4366	1248.7414	0.0000	0.0248
## 226	-0.4582	0.0115	-39.9417	0.0000	-0.4807	-0.4358	1261.6503	0.0000	0.0250
## 227	-0.4584	0.0115	-39.9734	0.0000	-0.4809	-0.4359	1260.8559	0.0000	0.0250
## 228	-0.4569	0.0114	-40.0201	0.0000	-0.4793	-0.4346	1249.3058	0.0000	0.0247
## 229	-0.4578	0.0114	-39.9836	0.0000	-0.4802	-0.4354	1261.2804	0.0000	0.0249
## 230	-0.4567	0.0114	-40.1055	0.0000	-0.4790	-0.4343	1243.6476	0.0000	0.0245
## 231	-0.4593	0.0114	-40.3054	0.0000	-0.4817	-0.4370	1244.4762	0.0000	0.0246
## 232	-0.4575	0.0115	-39.9348	0.0000	-0.4800	-0.4351	1259.0632	0.0000	0.0249
## 233	-0.4574	0.0115	-39.9368	0.0000	-0.4799	-0.4350	1256.8979	0.0000	0.0249
## 234	-0.4580	0.0115	-39.9917	0.0000	-0.4804	-0.4356	1261.7159	0.0000	0.0249
## 235	-0.4581	0.0115	-39.9555	0.0000	-0.4806	-0.4357	1261.7613	0.0000	0.0250
## 236	-0.4590	0.0114	-40.1345	0.0000	-0.4814	-0.4365	1253.7130	0.0000	0.0248
## 237	-0.4569	0.0114	-40.0391	0.0000	-0.4792	-0.4345	1247.6046	0.0000	0.0247
## 238	-0.4582	0.0115	-39.9954	0.0000	-0.4807	-0.4358	1261.6926	0.0000	0.0250
## 239	-0.4558	0.0113	-40.4928	0.0000	-0.4778	-0.4337	1220.5969	0.0000	0.0238
## 240	-0.4584	0.0115	-39.9915	0.0000	-0.4809	-0.4359	1261.1608	0.0000	0.0250
## 241	-0.4586	0.0115	-40.0354	0.0000	-0.4811	-0.4362	1259.5594	0.0000	0.0249
## 242	-0.4582	0.0115	-39.9156	0.0000	-0.4807	-0.4357	1261.7023	0.0000	0.0250
## 243	-0.4585	0.0115	-40.0097	0.0000	-0.4810	-0.4361	1260.5128	0.0000	0.0249
## 244	-0.4586	0.0114	-40.0814	0.0000	-0.4810	-0.4362	1260.1816	0.0000	0.0249
## 245	-0.4585	0.0114	-40.1350	0.0000	-0.4809	-0.4361	1259.7402	0.0000	0.0248
## 246	-0.4581	0.0115	-39.9288	0.0000	-0.4806	-0.4356	1261.7538	0.0000	0.0250
## 247	-0.4582	0.0115	-39.9859	0.0000	-0.4806	-0.4357	1261.7469	0.0000	0.0250
## 248	-0.4582	0.0115	-39.9360	0.0000	-0.4807	-0.4357	1261.7466	0.0000	0.0250
##	I2	H2							
## 1	82.4535	5.6991							
## 2	82.4796	5.7076							
## 3	82.5112	5.7180							
## 4	82.3621	5.6696							
## 5	82.4611	5.7016							
## 6	82.4656	5.7031							
## 7	82.4742	5.7059							
## 8	82.4719	5.7051							
## 9	82.4782	5.7072							
## 10	82.5131	5.7186							
## 11	81.5395	5.4170							
## 12	82.4652	5.7030							
## 13	82.3438	5.6637							
## 14	82.4277	5.6908							
## 15	82.4077	5.6843							
## 16	82.4706	5.7047							
## 17	82.3490	5.6654							
## 18	82.5001	5.7143							

## 19 82.5206 5.7210  
## 20 82.4282 5.6909  
## 21 82.5193 5.7206  
## 22 82.4628 5.7022  
## 23 82.5143 5.7190  
## 24 82.4230 5.6892  
## 25 82.5021 5.7150  
## 26 82.4862 5.7098  
## 27 82.4936 5.7122  
## 28 82.4641 5.7026  
## 29 82.4555 5.6998  
## 30 82.4252 5.6900  
## 31 82.5048 5.7159  
## 32 82.5019 5.7149  
## 33 82.3491 5.6654  
## 34 82.5100 5.7176  
## 35 82.2393 5.6304  
## 36 82.3721 5.6728  
## 37 82.4194 5.6881  
## 38 82.5071 5.7166  
## 39 82.5252 5.7225  
## 40 82.4770 5.7068  
## 41 82.5014 5.7148  
## 42 82.4597 5.7011  
## 43 82.4044 5.6832  
## 44 82.4219 5.6889  
## 45 82.4293 5.6913  
## 46 82.4433 5.6958  
## 47 82.5280 5.7234  
## 48 82.4441 5.6961  
## 49 82.5248 5.7224  
## 50 82.4686 5.7041  
## 51 82.4997 5.7142  
## 52 82.4157 5.6869  
## 53 82.4205 5.6884  
## 54 82.4967 5.7132  
## 55 82.4562 5.7000  
## 56 82.5058 5.7162  
## 57 82.5195 5.7207  
## 58 82.3380 5.6619  
## 59 82.4623 5.7020  
## 60 82.2790 5.6430  
## 61 82.4998 5.7142  
## 62 82.4311 5.6919  
## 63 82.5116 5.7181  
## 64 82.1868 5.6138  
## 65 82.4024 5.6826  
## 66 82.4833 5.7089

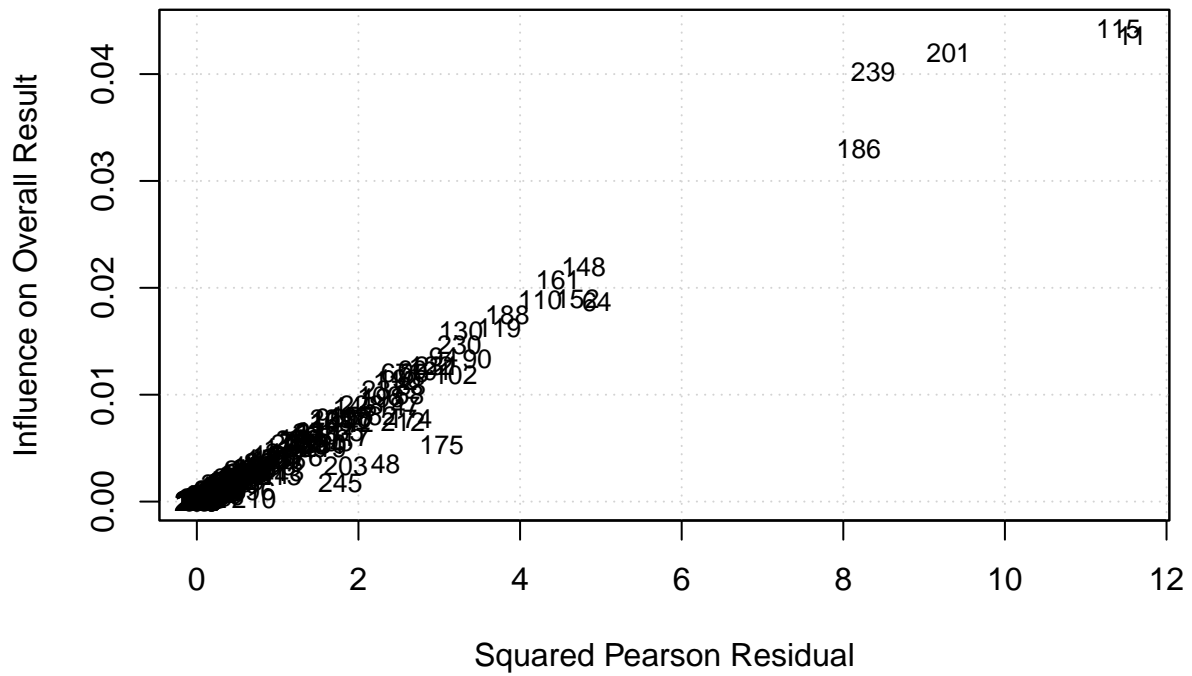
## 67 82.2565 5.6359  
## 68 82.4906 5.7112  
## 69 82.4407 5.6950  
## 70 82.3688 5.6718  
## 71 82.5230 5.7218  
## 72 82.4748 5.7061  
## 73 82.4531 5.6990  
## 74 82.4802 5.7078  
## 75 82.4694 5.7043  
## 76 82.4973 5.7134  
## 77 82.3899 5.6786  
## 78 82.4415 5.6953  
## 79 82.4704 5.7046  
## 80 82.4204 5.6884  
## 81 82.5047 5.7158  
## 82 82.5054 5.7161  
## 83 82.5016 5.7148  
## 84 82.4733 5.7056  
## 85 82.4829 5.7087  
## 86 82.4361 5.6935  
## 87 82.4717 5.7051  
## 88 82.3207 5.6563  
## 89 82.4733 5.7056  
## 90 82.2782 5.6428  
## 91 82.4692 5.7042  
## 92 82.3582 5.6684  
## 93 82.4355 5.6933  
## 94 82.2079 5.6205  
## 95 82.4220 5.6889  
## 96 82.3365 5.6614  
## 97 82.5179 5.7201  
## 98 82.4572 5.7004  
## 99 82.5127 5.7184  
## 100 82.4066 5.6840  
## 101 82.3614 5.6694  
## 102 82.2891 5.6462  
## 103 82.4937 5.7122  
## 104 82.4856 5.7096  
## 105 82.3689 5.6718  
## 106 82.3304 5.6594  
## 107 82.5121 5.7182  
## 108 82.4131 5.6861  
## 109 82.5105 5.7177  
## 110 82.1751 5.6101  
## 111 82.4930 5.7120  
## 112 82.2955 5.6483  
## 113 82.3935 5.6797  
## 114 82.4235 5.6894

## 115 81.7660 5.4843  
## 116 82.4939 5.7123  
## 117 82.4123 5.6858  
## 118 82.4732 5.7056  
## 119 82.1605 5.6055  
## 120 82.4791 5.7075  
## 121 82.4934 5.7121  
## 122 82.2816 5.6438  
## 123 82.3849 5.6769  
## 124 82.4799 5.7077  
## 125 82.4827 5.7086  
## 126 82.4928 5.7119  
## 127 82.5167 5.7198  
## 128 82.4765 5.7066  
## 129 82.5246 5.7223  
## 130 82.2006 5.6182  
## 131 82.4656 5.7031  
## 132 82.4538 5.6993  
## 133 82.5239 5.7221  
## 134 82.4240 5.6896  
## 135 82.5111 5.7179  
## 136 82.4871 5.7101  
## 137 82.4441 5.6961  
## 138 82.4988 5.7139  
## 139 82.4871 5.7101  
## 140 82.5080 5.7169  
## 141 82.5122 5.7183  
## 142 82.3834 5.6765  
## 143 82.4837 5.7090  
## 144 82.5024 5.7151  
## 145 82.4986 5.7138  
## 146 82.3103 5.6530  
## 147 82.3570 5.6680  
## 148 81.9588 5.5429  
## 149 82.5090 5.7172  
## 150 82.4142 5.6864  
## 151 82.5135 5.7187  
## 152 82.1232 5.5938  
## 153 82.5011 5.7147  
## 154 82.5086 5.7171  
## 155 82.4159 5.6870  
## 156 82.4416 5.6953  
## 157 82.3573 5.6681  
## 158 82.4547 5.6995  
## 159 82.5175 5.7200  
## 160 82.4216 5.6888  
## 161 82.1386 5.5987  
## 162 82.3726 5.6730

## 163 82.4654 5.7030  
## 164 82.5263 5.7229  
## 165 82.3974 5.6810  
## 166 82.4748 5.7061  
## 167 82.4650 5.7029  
## 168 82.3458 5.6644  
## 169 82.4372 5.6939  
## 170 82.5130 5.7185  
## 171 82.4555 5.6998  
## 172 82.5104 5.7177  
## 173 82.5237 5.7221  
## 174 82.3770 5.6744  
## 175 82.4129 5.6860  
## 176 82.4453 5.6965  
## 177 82.5235 5.7220  
## 178 82.5172 5.7199  
## 179 82.4446 5.6962  
## 180 82.4485 5.6975  
## 181 82.4846 5.7093  
## 182 82.5158 5.7194  
## 183 82.4604 5.7014  
## 184 82.4712 5.7049  
## 185 82.4662 5.7033  
## 186 81.9475 5.5394  
## 187 82.4988 5.7139  
## 188 82.1959 5.6167  
## 189 82.4233 5.6893  
## 190 82.2951 5.6482  
## 191 82.2643 5.6383  
## 192 82.5182 5.7202  
## 193 82.5095 5.7174  
## 194 82.4609 5.7016  
## 195 82.4974 5.7134  
## 196 82.4879 5.7103  
## 197 82.4874 5.7102  
## 198 82.3200 5.6561  
## 199 82.3954 5.6803  
## 200 82.4321 5.6922  
## 201 81.7979 5.4939  
## 202 82.3922 5.6793  
## 203 82.4477 5.6972  
## 204 82.5038 5.7155  
## 205 82.4094 5.6849  
## 206 82.4410 5.6951  
## 207 82.4626 5.7021  
## 208 82.4389 5.6944  
## 209 82.5193 5.7206  
## 210 82.4875 5.7102

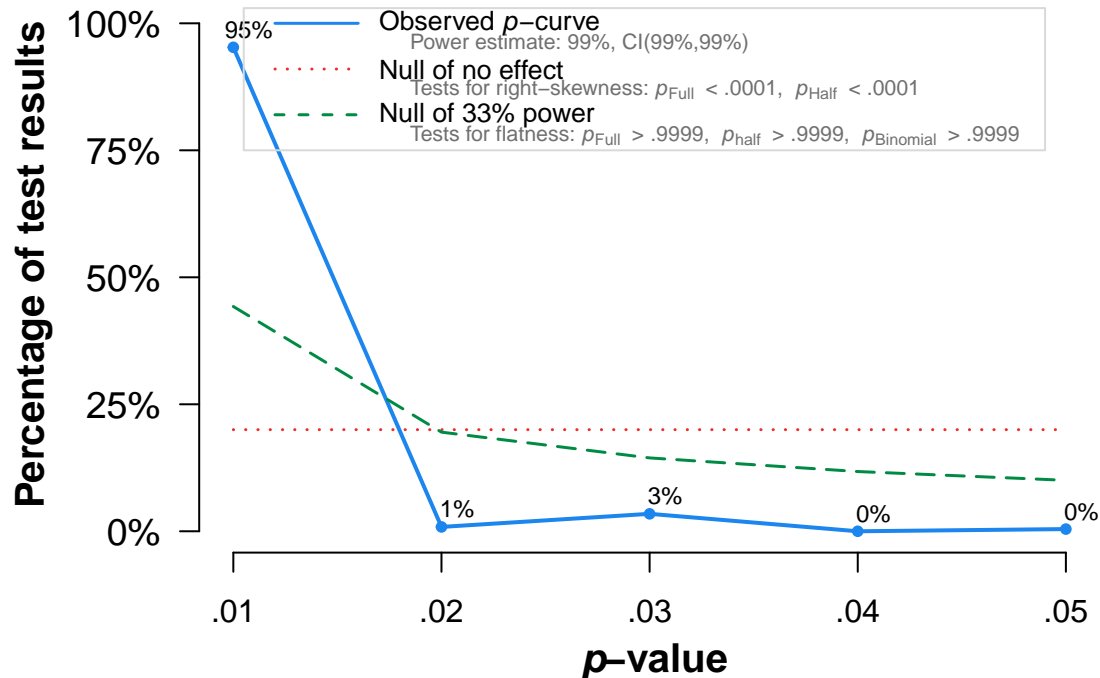
```
## 211 82.5043 5.7157
## 212 82.3825 5.6762
## 213 82.4723 5.7053
## 214 82.5194 5.7206
## 215 82.4692 5.7043
## 216 82.5029 5.7152
## 217 82.4430 5.6957
## 218 82.5243 5.7222
## 219 82.4337 5.6927
## 220 82.5247 5.7224
## 221 82.4805 5.7079
## 222 82.4591 5.7009
## 223 82.3729 5.6731
## 224 82.4638 5.7025
## 225 82.3256 5.6579
## 226 82.5044 5.7157
## 227 82.4889 5.7107
## 228 82.3395 5.6623
## 229 82.5137 5.7188
## 230 82.2428 5.6315
## 231 82.2490 5.6335
## 232 82.4825 5.7086
## 233 82.4511 5.6983
## 234 82.5227 5.7217
## 235 82.5263 5.7229
## 236 82.3791 5.6751
## 237 82.3151 5.6545
## 238 82.5237 5.7220
## 239 81.8193 5.5003
## 240 82.5089 5.7172
## 241 82.4730 5.7055
## 242 82.4196 5.6881
## 243 82.4931 5.7120
## 244 82.4807 5.7080
## 245 82.4706 5.7047
## 246 82.5137 5.7188
## 247 82.5261 5.7228
## 248 82.5106 5.7178
```

```
# Baujat plot
baujat(rma(effect_size,
           var,
           data = bflpedat,
           method = "REML"))
```



#### 4.7 P-Curve and Z-Curve

```
# P-curve estimation
pcurve(metagen(TE = effect_size,
  seTE = se,
  data = bflpedat,
  studlab = paste(country),
  comb.fixed = FALSE,
  comb.random = TRUE,
  method.tau = "REML"))
```



Note: The observed p-curve includes 233 statistically significant ( $p < .05$ ) results, of which 229 are  $p < .025$ . There were 15 additional results entered but excluded from p-curve because they were  $p > .05$ .

```
## P-curve analysis
## -----
## - Total number of provided studies: k = 248
## - Total number of p<0.05 studies included into the analysis: k = 233 (93.95%)
## - Total number of studies with p<0.025: k = 229 (92.34%)
##
## Results
## -----
##                pBinomial   zFull pFull   zHalf pHalf
## Right-skewness test         0 -75.090    0 -72.873    0
## Flatness test                1  61.522    1  65.682    1
## Note: p-values of 0 or 1 correspond to p<0.001 and p>0.999, respectively.
## Power Estimate: 99% (99%-99%)
##
## Evidential value
## -----
## - Evidential value present: yes
## - Evidential value absent/inadequate: no
```

```
# Create z- scores
bflpedat$zval <- bflpedat$effect_size/bflpedat$se

# Z-curve estimation
zedcurve <- zcurve(bflpedat$zval, method = "EM", bootstrap = 1000)
zedcurve
```

```
## Call:
```

```

## zcurve(z = bflpedat$zval, method = "EM", bootstrap = 1000)
##
## Estimates:
##      ERR      EDR
## 0.9601997 0.9330032
# Summarize the z-curve elements
summary.zcurve(zedcurve)

## Call:
## zcurve(z = bflpedat$zval, method = "EM", bootstrap = 1000)
##
## model: EM via EM
##
##      Estimate  l.CI  u.CI
## ERR      0.960 0.892 1.000
## EDR      0.933 0.618 1.000
##
## Model converged in 23 + 920 iterations
## Fitted using 127 z-values. 248 supplied, 233 significant (ODR = 0.94, 95% CI [0.90, 0.96]).
## Q = -172.67, 95% CI[-192.78, -150.18]
# z-curve estimates
ERR(zedcurve) ## Expected replication rate

## Estimate      l.CI      u.CI
##      0.960      0.892      1.000
EDR(zedcurve) ## Expected discovery rate

## Estimate      l.CI      u.CI
##      0.933      0.618      1.000
ODR(zedcurve) ## Observed discovery rate

## Estimate      l.CI      u.CI
##      0.940      0.900      0.965
Soric(zedcurve) ## Soric false discovery rate

## Estimate      l.CI      u.CI
##      0.004      0.000      0.033
file_drawer_ratio(zedcurve)

## Estimate      l.CI      u.CI
##      0.072      0.000      0.619
expected_n(zedcurve)

## Estimate      l.CI      u.CI
##      250      233      377

```

```
missing_n(zedcurve)
```

```
## Estimate      1.CI      u.CI  
##           2       -15       129
```

```
significant_n(zedcurve)
```

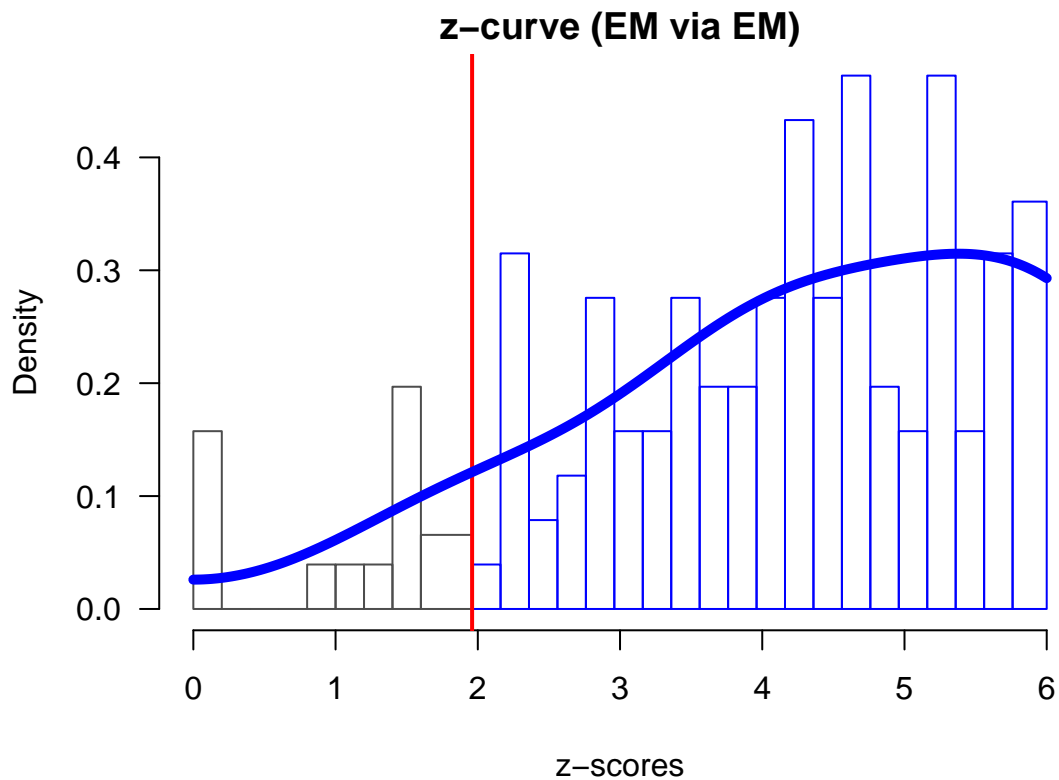
```
## N  
## 233
```

```
included_n(zedcurve)
```

```
## N  
## 127
```

```
# Plot the z-curve
```

```
plot.zcurve(zedcurve)
```



#### 4.8 Publication bias via sensitivity analyses

```
## Worst-case sensitivity analyses  
svalue(yi = bflpedat$effect_size,  
       vi = bflpedat$var,  
       clustervar = bflpedat$country,  
       q = 0, ## Threshold to shift to  
       model = "robust",
```

```
favor.positive = FALSE, ## Not favoring positive effects
return.worst.meta = TRUE)
```

```
## $stats
##      sval.est      sval.ci k.affirmative k.nonaffirmative signs.recorded
## 1 Not possible Not possible          233              15          TRUE
##
## $meta.worst
## RVE: User Specified Weights with Small-Sample Corrections
##
## Model: yi ~ 1
##
## Number of studies = 12
## Number of outcomes = 15 (min = 1 , mean = 1.25 , median = 1 , max = 3 )
##      Estimate StdErr t-value  dfs P(|t|>) 95% CI.L 95% CI.U Sig
## 1 X.Intercept.    0.128 0.0368   3.47 8.96 0.00707  0.0445  0.211 ***
## ---
## Signif. codes: < .01 *** < .05 ** < .10 *
## ---
## Note: If df < 4, do not trust the results
```

## 5 R session info

```
sessionInfo()
```

```
## R version 4.1.1 (2021-08-10)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS Catalina 10.15.7
##
## Matrix products: default
## BLAS:   /Library/Frameworks/R.framework/Versions/4.1/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.1/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] grid      stats      graphics  grDevices  utils      datasets  methods
## [8] base
##
## other attached packages:
## [1] PublicationBias_2.2.0 zcurve_1.0.9          caret_6.0-88
## [4] lattice_0.20-44      metaforest_0.1.3     data.table_1.14.0
## [7] ranger_0.13.1        meta_4.19-0          ggpubr_0.4.0
## [10] psych_2.1.6          forestplot_2.0.0     checkmate_2.0.0
## [13] magrittr_2.0.1       dmetar_0.0.9000      forcats_0.5.1
## [16] stringr_1.4.0        dplyr_1.0.7          purrr_0.3.4
## [19] readr_2.0.0          tidyr_1.1.3          tibble_3.1.3
## [22] ggplot2_3.3.5        tidyverse_1.3.1      clubSandwich_0.5.3
## [25] robumeta_2.0         metaSEM_1.2.5.1      OpenMx_2.19.6
## [28] metafor_3.0-2        Matrix_1.3-4
##
## loaded via a namespace (and not attached):
## [1] readxl_1.3.1          backports_1.2.1      plyr_1.8.6
## [4] splines_4.1.1         digest_0.6.27        foreach_1.5.1
## [7] htmltools_0.5.1.1    fansi_0.5.0          cluster_2.1.2
## [10] tzdb_0.1.2           openxlsx_4.2.4       recipes_0.1.16
## [13] modelr_0.1.8          gower_0.2.2          RcppParallel_5.1.4
## [16] sandwich_3.0-1        colorspace_2.0-2     rvest_1.0.1
## [19] ggrepel_0.9.1        rbibutils_2.2.3      haven_2.4.2
## [22] xfun_0.24            crayon_1.4.1         jsonlite_1.7.2
## [25] lme4_1.1-27.1        survival_3.2-11      zoo_1.8-9
## [28] iterators_1.0.13     glue_1.4.2           gtable_0.3.0
## [31] ipred_0.9-11         car_3.0-11           kernlab_0.9-29
## [34] prabclus_2.3-2       DEoptimR_1.0-9       abind_1.4-5
## [37] scales_1.1.1         mvtnorm_1.1-2        DBI_1.1.1
## [40] rstatix_0.7.0        Rcpp_1.0.7           magic_1.5-9
## [43] tmvnsim_1.0-2        foreign_0.8-81       mclust_5.4.7
```

```

## [46] stats4_4.1.1      lava_1.6.9        prodlim_2019.11.13
## [49] netmeta_1.5-0     httr_1.4.2        fpc_2.2-9
## [52] lavaan_0.6-9      modeltools_0.2-23 ellipsis_0.3.2
## [55] pkgconfig_2.0.3   flexmix_2.3-17    nnet_7.3-16
## [58] dbplyr_2.1.1      utf8_1.2.2        tidyselect_1.1.1
## [61] rlang_0.4.11      reshape2_1.4.4    munsell_0.5.0
## [64] cellranger_1.1.0  tools_4.1.1       cli_3.0.1
## [67] generics_0.1.0    broom_0.7.9       mathjaxr_1.4-0
## [70] evaluate_0.14     yaml_2.2.1        ModelMetrics_1.2.2.2
## [73] knitr_1.33        fs_1.5.0          zip_2.2.0
## [76] robustbase_0.93-8 pbapply_1.4-3     nlme_3.1-152
## [79] xml2_1.3.2        compiler_4.1.1    rstudioapi_0.13
## [82] curl_4.3.2        ggsignif_0.6.2    MetaUtility_2.1.1
## [85] reprex_2.0.0      pbivnorm_0.6.0    stringi_1.7.3
## [88] highr_0.9         poibin_1.5        nloptr_1.2.2.2
## [91] vctrs_0.3.8       CompQuadForm_1.4.3 pillar_1.6.2
## [94] lifecycle_1.0.0   Rdpack_2.1.2      R6_2.5.0
## [97] MuMIn_1.43.17     gridExtra_2.3     rio_0.5.27
## [100] codetools_0.2-18 boot_1.3-28        MASS_7.3-54
## [103] assertthat_0.2.1 withr_2.4.2        mnormt_2.0.2
## [106] diptest_0.76-0    parallel_4.1.1    hms_1.1.0
## [109] rpart_4.1-15      timeDate_3043.102 class_7.3-19
## [112] minqa_1.2.4       rmarkdown_2.9     carData_3.0-4
## [115] pROC_1.17.0.1     lubridate_1.7.10  ellipse_0.4.2

```