

R syntax to accomany Which Omega is Right?

This .rmd file provides all syntax described in the article “Your coefficient alpha is probably wrong, but which coefficient omega is right? A tutorial on using R to obtain better reliability estimates” by David B. Flora.

This document assumes some prior familiarity with R, particularly how to install and load packages and how to import data files. We recommend using RStudio as an environment for all R sessions.

Begin by installing (if necessary) and loading all packages needed for this tutorial:

```
#install.packages("lavaan")
#install.packages("semTools")
#install.packages("psych")
#install.packages("MBESS")

library(lavaan)

## This is lavaan 0.6-5

## lavaan is BETA software! Please report any bugs.

library(semTools)

##

##
#####
##

## This is semTools 0.5-2.920

## All users of R (or SEM) are invited to submit functions or ideas for
functions.

##
#####
##

library(psych)

##
## Attaching package: 'psych'

## The following object is masked from 'package:semTools':
##
##      skew
```

```
## The following object is masked from 'package:lavaan':
##
##      cor2cov

library(MBESS)

##
## Attaching package: 'MBESS'

## The following object is masked from 'package:psych':
##
##      cor2cov

## The following object is masked from 'package:lavaan':
##
##      cor2cov
```

Obtaining omega-u from a one-factor model for the Openness scale.

Download the `open.csv` data from the web to create a data frame called “open”, or import the `open.csv` data file if you have already saved it to the same directory as this `.rmd` file. These data are from the “bfi” dataset in the `psych` package. Note that in the “open” data file used below, items 2 and 5 are already reverse-coded.

```
open <- read.csv("https://osf.io/53wdz/download")

#OR
#open <- read.csv("open.csv")
```

Specify the one-factor model for the Openness items (i.e., congeneric model):

```
mod1f <- 'open =~ 01 + 02+ 03 + 04 + 05'
```

Estimate the one-factor model:

```
fit1f <- cfa(mod1f, data=open, std.lv=T, missing='direct', estimator='MLR')
```

Obtain the results summary:

```
summary(fit1f, fit.measures=T)

## lavaan 0.6-5 ended normally after 31 iterations
##
##      Estimator                ML
##      Optimization method      NLMINB
##      Number of free parameters      15
##
##      Number of observations      2800
##      Number of missing patterns      7
##
```

```

## Model Test User Model:
##
##           Standard      Robust
## Test Statistic      92.411    77.828
## Degrees of freedom           5         5
## P-value (Chi-square)      0.000    0.000
## Scaling correction factor      1.187
##   for the Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
## Test statistic      1399.423    1066.430
## Degrees of freedom      10         10
## P-value      0.000    0.000
## Scaling correction factor      1.312
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI)      0.937    0.931
## Tucker-Lewis Index (TLI)      0.874    0.862
##
## Robust Comparative Fit Index (CFI)      0.938
## Robust Tucker-Lewis Index (TLI)      0.875
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0)      -22581.302    -22581.302
## Scaling correction factor      1.171
##   for the MLR correction
## Loglikelihood unrestricted model (H1)      -22535.096    -22535.096
## Scaling correction factor      1.175
##   for the MLR correction
##
## Akaike (AIC)      45192.603    45192.603
## Bayesian (BIC)      45281.664    45281.664
## Sample-size adjusted Bayesian (BIC)      45234.004    45234.004
##
## Root Mean Square Error of Approximation:
##
## RMSEA      0.079    0.072
## 90 Percent confidence interval - lower      0.065    0.060
## 90 Percent confidence interval - upper      0.094    0.085
## P-value RMSEA <= 0.05      0.000    0.002
##
## Robust RMSEA      0.079
## 90 Percent confidence interval - lower      0.064
## 90 Percent confidence interval - upper      0.094
##
## Standardized Root Mean Square Residual:
##
## SRMR      0.031    0.031

```

```
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|)
## open =~
## 01 0.622 0.029 21.536 0.000
## 02 0.684 0.042 16.466 0.000
## 03 0.794 0.032 24.572 0.000
## 04 0.361 0.031 11.779 0.000
## 05 0.685 0.036 19.069 0.000
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|)
## .01 4.816 0.021 224.892 0.000
## .02 4.287 0.030 144.955 0.000
## .03 4.436 0.023 191.353 0.000
## .04 4.893 0.023 211.544 0.000
## .05 4.509 0.025 179.095 0.000
## open 0.000
##
## Variances:
## Estimate Std.Err z-value P(>|z|)
## .01 0.888 0.037 23.887 0.000
## .02 1.981 0.068 29.245 0.000
## .03 0.860 0.050 17.051 0.000
## .04 1.361 0.052 26.271 0.000
## .05 1.294 0.059 21.957 0.000
## open 1.000
```

As reported in the article, the output above indicates that the model does not fit the data very well. One reason is that there is a notable residual correlation (.097) between items 02 and 05, which can be seen by running the residuals command on the fitted model:

```
residuals(fit1f, type="cor")
## $type
## [1] "cor.bollen"
##
## $cov
## 01 02 03 04 05
## 01 0.000
## 02 -0.028 0.000
## 03 0.037 -0.024 0.000
## 04 0.015 -0.061 0.002 0.000
```

```
## 05 -0.046  0.097 -0.024  0.026  0.000
##
## $mean
## 01 02 03 04 05
##  0  0  0  0  0
```

Despite that the model fit is not great, for illustrative reasons we will continue to estimate omega based on this one-factor model. (Syntax to account for the large error correlation between 02 and 05 and thus obtain a better omega estimate is given later in this document).

Obtain coefficient omega as an estimate of the reliability of the openness scale total score as a measure of the 'open' factor:

```
reliability(fit1f)

##                open
## alpha  0.5999111
## omega  0.6079033
## omega2 0.6079033
## omega3 0.6078732
## avevar 0.2461983
```

Call the `ci.reliability` function to obtain a bootstrap 95% CI for omega. Note that for this tutorial, the number of bootstrap samples is set to only 100 (the default number of samples is 10,000, which may take a long time to run). Also, due to random sampling variability inherent to the bootstrap procedure, the resulting CI from the code below may not exactly match the CI reported in the actual article:

```
ci.reliability(data=open, type = "omega", interval.type = "perc", B = 100)

## $est
## [1] 0.6079033
##
## $se
## [1] 0.0142746
##
## $ci.lower
## [1] 0.5766502
##
## $ci.upper
## [1] 0.6300214
##
## $conf.level
## [1] 0.95
##
## $type
## [1] "omega"
##
## $interval.type
## [1] "percentile bootstrap"
```

It is also possible to calculate omega-u directly within lavaan as a defined parameter. To do so, it is necessary to assign a name to each factor loading and error variance in the model specification; below, l1 through l5 are the names of the factor loadings, and e1 through e5 are the names of the error variances. Then the formula for the defined omega parameter is given following the " := " operator

```
open1f <- 'open=~l1*O1+l2*O2+l3*O3+l4*O4+l5*O5
          O1 ~~ e1*O1
          O2 ~~ e2*O2
          O3 ~~ e3*O3
          O4 ~~ e4*O4
          O5 ~~ e5*O5
omega := ((l1+l2+l3+l4+l5)^2)
/ ((l1+l2+l3+l4+l5)^2 +
(e1+e2+e3+e4+e5))'
```

Fit the model using the `cfa()` function as above, then the estimate of omega in the results summary matches the omega estimate returned by the `reliability()` function:

```
fit1f <- cfa(open1f, data=open, missing="direct", estimator="MLR", std.lv=T)
summary(fit1f, fit.measures=T)
```

```
## lavaan 0.6-5 ended normally after 31 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      15
##
##      Number of observations          2800
##      Number of missing patterns      7
##
## Model Test User Model:
##
##              Standard      Robust
##      Test Statistic      92.411    77.828
##      Degrees of freedom      5        5
##      P-value (Chi-square)    0.000    0.000
##      Scaling correction factor      1.187
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##      Test statistic      1399.423    1066.430
##      Degrees of freedom      10        10
##      P-value              0.000    0.000
##      Scaling correction factor      1.312
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.937    0.931
##      Tucker-Lewis Index (TLI)        0.874    0.862
```

```

##
## Robust Comparative Fit Index (CFI) 0.938
## Robust Tucker-Lewis Index (TLI) 0.875
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -22581.302 -22581.302
## Scaling correction factor 1.171
## for the MLR correction
## Loglikelihood unrestricted model (H1) -22535.096 -22535.096
## Scaling correction factor 1.175
## for the MLR correction
##
## Akaike (AIC) 45192.603 45192.603
## Bayesian (BIC) 45281.664 45281.664
## Sample-size adjusted Bayesian (BIC) 45234.004 45234.004
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.079 0.072
## 90 Percent confidence interval - lower 0.065 0.060
## 90 Percent confidence interval - upper 0.094 0.085
## P-value RMSEA <= 0.05 0.000 0.002
##
## Robust RMSEA 0.079
## 90 Percent confidence interval - lower 0.064
## 90 Percent confidence interval - upper 0.094
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.031 0.031
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|)
## open =~
## 01 (11) 0.622 0.029 21.536 0.000
## 02 (12) 0.684 0.042 16.466 0.000
## 03 (13) 0.794 0.032 24.572 0.000
## 04 (14) 0.361 0.031 11.779 0.000
## 05 (15) 0.685 0.036 19.069 0.000
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|)
## .01 4.816 0.021 224.892 0.000

```

```
##      .02                4.287    0.030  144.955    0.000
##      .03                4.436    0.023  191.353    0.000
##      .04                4.893    0.023  211.544    0.000
##      .05                4.509    0.025  179.095    0.000
##      open              0.000
##
## Variances:
##              Estimate Std.Err  z-value  P(>|z|)
##      .01      (e1)    0.888    0.037   23.887    0.000
##      .02      (e2)    1.981    0.068   29.245    0.000
##      .03      (e3)    0.860    0.050   17.051    0.000
##      .04      (e4)    1.361    0.052   26.271    0.000
##      .05      (e5)    1.294    0.059   21.957    0.000
##      open              1.000
##
## Defined Parameters:
##              Estimate Std.Err  z-value  P(>|z|)
##      omega          0.608    0.013   47.829    0.000
```

The code below specifies a tau-equivalence model for the Openness items. By assigning the same name (“lam”) to each factor loading coefficient, those factor loadings will be constrained to be equal:

```
modtaueq <- 'open =~ lam*01 + lam*02+ lam*03 + lam*04 + lam*05'
fitte <- cfa(modtaueq, data=open, std.lv=T, missing='direct',
estimator='MLR')
```

Check the results to see that the factor loading estimates are in fact equal:

```
summary(fitte, fit.measures=T)

## lavaan 0.6-5 ended normally after 26 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      15
##      Number of equality constraints    4
##      Row rank of the constraints matrix 4
##
##      Number of observations          2800
##      Number of missing patterns       7
##
## Model Test User Model:
##              Standard      Robust
##      Test Statistic      223.263  184.586
##      Degrees of freedom         9      9
##      P-value (Chi-square)      0.000    0.000
##      Scaling correction factor      1.210
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
```



```

##
## Test statistic 1399.423 1066.430
## Degrees of freedom 10 10
## P-value 0.000 0.000
## Scaling correction factor 1.312
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.846 0.834
## Tucker-Lewis Index (TLI) 0.829 0.815
##
## Robust Comparative Fit Index (CFI) 0.847
## Robust Tucker-Lewis Index (TLI) 0.830
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -22646.727 -22646.727
## Scaling correction factor 0.841
## for the MLR correction
## Loglikelihood unrestricted model (H1) -22535.096 -22535.096
## Scaling correction factor 1.175
## for the MLR correction
##
## Akaike (AIC) 45315.455 45315.455
## Bayesian (BIC) 45380.766 45380.766
## Sample-size adjusted Bayesian (BIC) 45345.815 45345.815
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.092 0.083
## 90 Percent confidence interval - lower 0.082 0.074
## 90 Percent confidence interval - upper 0.103 0.093
## P-value RMSEA <= 0.05 0.000 0.000
##
## Robust RMSEA 0.092
## 90 Percent confidence interval - lower 0.081
## 90 Percent confidence interval - upper 0.104
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.070 0.070
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|)

```

```
## open =~
## 01 (lam) 0.628 0.014 45.567 0.000
## 02 (lam) 0.628 0.014 45.567 0.000
## 03 (lam) 0.628 0.014 45.567 0.000
## 04 (lam) 0.628 0.014 45.567 0.000
## 05 (lam) 0.628 0.014 45.567 0.000
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|)
## .01 4.816 0.021 224.874 0.000
## .02 4.287 0.030 144.955 0.000
## .03 4.436 0.023 191.473 0.000
## .04 4.893 0.023 211.539 0.000
## .05 4.509 0.025 179.105 0.000
## open 0.000
##
## Variances:
## Estimate Std.Err z-value P(>|z|)
## .01 0.882 0.033 26.507 0.000
## .02 2.024 0.053 38.354 0.000
## .03 1.002 0.039 25.782 0.000
## .04 1.283 0.053 24.435 0.000
## .05 1.320 0.046 28.976 0.000
## open 1.000
```

Next, results from the `anova()` function show that the original congeneric model fits better than the tau-equivalence model. Because the sample size is so large, the chi-square difference test is easily significant. But the AIC and BIC indices are also lower for the congeneric model, indicating that the congeneric model is more appropriate for these data than the tau-equivalence model:

```
anova(fit1f, fitte)

## Scaled Chi-Squared Difference Test (method = "satorra.bentler.2001")
##
## lavaan NOTE:
## The "Chisq" column contains standard test statistics, not the
## robust test that should be reported per model. A robust difference
## test is a function of two standard (not robust) statistics.
##
## Df AIC BIC Chisq Chisq diff Df diff Pr(>Chisq)
## fit1f 5 45193 45282 92.411
## fitte 9 45315 45381 223.263 105.76 4 < 2.2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Incorporating the error covariance in the one-factor model for the Openness scale to obtain omega-u from a better-fitting model.

Re-specify the one-factor model for the Openness items, but now include a new line to specify the free error covariance between items O2 and O5 using the `~~` operator:

```
mod1fR <- 'open =~ O1 + O2+ O3 + O4 + O5
          O2 ~~ O5'
```

Estimate the revised model and obtain the results summary:

```
fit1fR <- cfa(mod1fR, data=open, std.lv=T, missing='direct', estimator='MLR')
summary(fit1fR, fit.measures=T)
```

```
## lavaan 0.6-5 ended normally after 37 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      16
##
##      Number of observations         2800
##      Number of missing patterns      7
##
## Model Test User Model:
##
##              Standard      Robust
##      Test Statistic      19.968    16.569
##      Degrees of freedom      4        4
##      P-value (Chi-square)    0.001    0.002
##      Scaling correction factor      1.205
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##      Test statistic      1399.423    1066.430
##      Degrees of freedom      10        10
##      P-value              0.000    0.000
##      Scaling correction factor      1.312
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.989    0.988
##      Tucker-Lewis Index (TLI)        0.971    0.970
##
##      Robust Comparative Fit Index (CFI)      0.989
##      Robust Tucker-Lewis Index (TLI)        0.973
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -22545.080  -22545.080
##      Scaling correction factor          1.168
```

```

##      for the MLR correction
##  Loglikelihood unrestricted model (H1)      -22535.096  -22535.096
##  Scaling correction factor                    1.175
##      for the MLR correction
##
##  Akaike (AIC)                                45122.160  45122.160
##  Bayesian (BIC)                             45217.158  45217.158
##  Sample-size adjusted Bayesian (BIC)        45166.321  45166.321
##
## Root Mean Square Error of Approximation:
##
##  RMSEA                                0.038      0.033
##  90 Percent confidence interval - lower      0.022      0.019
##  90 Percent confidence interval - upper      0.055      0.049
##  P-value RMSEA <= 0.05                    0.871      0.957
##
##  Robust RMSEA                                0.037
##  90 Percent confidence interval - lower      0.020
##  90 Percent confidence interval - upper      0.056
##
## Standardized Root Mean Square Residual:
##
##  SRMR                                0.015      0.015
##
## Parameter Estimates:
##
##  Information                                Observed
##  Observed information based on              Hessian
##  Standard errors                          Robust.huber.white
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)
##  open =~
##    01          0.640    0.029   22.171   0.000
##    02          0.573    0.040   14.248   0.000
##    03          0.846    0.033   25.304   0.000
##    04          0.363    0.031   11.639   0.000
##    05          0.597    0.034   17.606   0.000
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)
##  .02 ~~
##  .05          0.329    0.046    7.155   0.000
##
## Intercepts:
##      Estimate  Std.Err  z-value  P(>|z|)
##  .01          4.816    0.021  224.930   0.000
##  .02          4.287    0.030  144.955   0.000
##  .03          4.436    0.023  191.343   0.000
##  .04          4.893    0.023  211.552   0.000

```

```
##      .05          4.509      0.025  179.133      0.000
##      open          0.000
##
## Variances:
##              Estimate Std.Err  z-value  P(>|z|)
##      .01           0.865    0.038   22.703    0.000
##      .02           2.120    0.062   33.937    0.000
##      .03           0.775    0.054   14.431    0.000
##      .04           1.359    0.052   26.120    0.000
##      .05           1.406    0.055   25.676    0.000
##      open          1.000
```

One use the `anova()` function to see that the revised model fits better then the original congeneric model:

```
anova(fit1f, fit1fR)

## Scaled Chi-Squared Difference Test (method = "satorra.bentler.2001")
##
## lavaan NOTE:
##      The "Chisq" column contains standard test statistics, not the
##      robust test that should be reported per model. A robust difference
##      test is a function of two standard (not robust) statistics.
##
##      Df   AIC   BIC  Chisq Chisq diff Df diff Pr(>Chisq)
## fit1fR  4 45122 45217 19.968
## fit1f   5 45193 45282 92.411      64.899      1 7.883e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Obtain an updated coefficient omega that correctly accounts for the error covariance:

```
reliability(fit1fR)

##              open
## alpha 0.5999111
## omega 0.5593717
## omega2 0.5593717
## omega3 0.5601129
## avevar 0.2294343
```

As above, an updated omega can be calculated directly within the lavaan syntax as a defined parameter. The new syntax below includes a formula for omega that accounts for the error covariance:

```
open1fR <- 'open=~l1*01+l2*02+l3*03+l4*04+l5*05
           01 ~~ e1*01
           02 ~~ e2*02
           03 ~~ e3*03
           04 ~~ e4*04'
```

```

05 ~~ e5*05
02 ~~ r*05
omega := ((l1+l2+l3+l4+l5)^2)
/ ((l1+l2+l3+l4+l5)^2 +
(e1+e2+e3+e4+e5)+2*r)'

```

Fit the model using the `cfa()` function as above, then the estimate of omega in the results summary matches the omega estimate returned by the `reliability()` function:

```

fit1fR <- cfa(open1fR, data=open, missing="direct", estimator="MLR",
std.lv=T)
summary(fit1fR, fit.measures=T)

## lavaan 0.6-5 ended normally after 37 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      16
##
##      Number of observations          2800
##      Number of missing patterns      7
##
## Model Test User Model:
##
##              Standard      Robust
##      Test Statistic      19.968    16.569
##      Degrees of freedom           4         4
##      P-value (Chi-square)      0.001    0.002
##      Scaling correction factor      1.205
##      for the Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##      Test statistic      1399.423    1066.430
##      Degrees of freedom      10         10
##      P-value                0.000    0.000
##      Scaling correction factor      1.312
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.989    0.988
##      Tucker-Lewis Index (TLI)        0.971    0.970
##
##      Robust Comparative Fit Index (CFI)      0.989
##      Robust Tucker-Lewis Index (TLI)        0.973
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -22545.080  -22545.080
##      Scaling correction factor          1.168
##      for the MLR correction

```

```

## Loglikelihood unrestricted model (H1) -22535.096 -22535.096
## Scaling correction factor 1.175
## for the MLR correction
##
## Akaike (AIC) 45122.160 45122.160
## Bayesian (BIC) 45217.158 45217.158
## Sample-size adjusted Bayesian (BIC) 45166.321 45166.321
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.038 0.033
## 90 Percent confidence interval - lower 0.022 0.019
## 90 Percent confidence interval - upper 0.055 0.049
## P-value RMSEA <= 0.05 0.871 0.957
##
## Robust RMSEA 0.037
## 90 Percent confidence interval - lower 0.020
## 90 Percent confidence interval - upper 0.056
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.015 0.015
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|)
## open =~
## 01 (11) 0.640 0.029 22.171 0.000
## 02 (12) 0.573 0.040 14.248 0.000
## 03 (13) 0.846 0.033 25.304 0.000
## 04 (14) 0.363 0.031 11.639 0.000
## 05 (15) 0.597 0.034 17.606 0.000
##
## Covariances:
## Estimate Std.Err z-value P(>|z|)
## .02 ~~
## .05 (r) 0.329 0.046 7.155 0.000
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|)
## .01 4.816 0.021 224.930 0.000
## .02 4.287 0.030 144.955 0.000
## .03 4.436 0.023 191.343 0.000
## .04 4.893 0.023 211.552 0.000
## .05 4.509 0.025 179.133 0.000

```

```
##      open                0.000
##
## Variances:
##              Estimate Std.Err z-value P(>|z|)
##      .01      (e1)    0.865   0.038  22.703  0.000
##      .02      (e2)    2.120   0.062  33.937  0.000
##      .03      (e3)    0.775   0.054  14.431  0.000
##      .04      (e4)    1.359   0.052  26.120  0.000
##      .05      (e5)    1.406   0.055  25.676  0.000
##      open                1.000
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|)
##      omega                0.559   0.015  37.015  0.000
```

Obtaining omega-u-cat from a one-factor model for the Psychoticism scale.

Download the `potic.csv` data from the web to create a data frame called 'potic', or import the `potic.csv` data file if you have already saved it to the same directory as this `.rmd` file.

```
potic <- read.csv("https://osf.io/atqc6/download")
```

#OR

```
#potic <- read.csv("potic.csv")
```

Specify the one-factor model for the Psychoticism items (i.e., congeneric model):

```
mod1f <- 'psycctsm =~ DDP1 + DDP2 + DDP3 + DDP4'
```

Estimate the one-factor model; note that the 'ordered' option is used to indicate that all items are ordered, categorical:

```
fit1f <- cfa(mod1f, data=potic, std.lv=T, ordered = T)
```

Obtain the results summary:

```
summary(fit1f, fit.measures=T)
```

```
## lavaan 0.6-5 ended normally after 10 iterations
```

```
##
```

```
##      Estimator                DWLS
```

```
##      Optimization method        NLMINB
```

```
##      Number of free parameters          20
```

```
##
```

```
##              Used          Total
```

```
##      Number of observations          498          500
```

```
##
```

```
## Model Test User Model:
```

```
##              Standard          Robust
```



```

## Test Statistic          7.117      14.910
## Degrees of freedom      2          2
## P-value (Chi-square)    0.028      0.001
## Scaling correction factor      0.480
## Shift parameter        0.089
##   for the simple second-order correction
##
## Model Test Baseline Model:
##
## Test statistic          1635.647    1316.086
## Degrees of freedom      6          6
## P-value                  0.000      0.000
## Scaling correction factor      1.244
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI)    0.997      0.990
## Tucker-Lewis Index (TLI)      0.991      0.970
##
## Robust Comparative Fit Index (CFI)      NA
## Robust Tucker-Lewis Index (TLI)      NA
##
## Root Mean Square Error of Approximation:
##
## RMSEA                    0.072      0.114
## 90 Percent confidence interval - lower    0.020      0.065
## 90 Percent confidence interval - upper    0.132      0.171
## P-value RMSEA <= 0.05    0.200      0.019
##
## Robust RMSEA      NA
## 90 Percent confidence interval - lower    NA
## 90 Percent confidence interval - upper    NA
##
## Standardized Root Mean Square Residual:
##
## SRMR                    0.033      0.033
##
## Parameter Estimates:
##
## Information              Expected
## Information saturated (h1) model    Unstructured
## Standard errors          Robust.sem
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)
## psyctcsm =~
## DDP1          0.894    0.028   31.538   0.000
## DDP2          0.753    0.028   26.711   0.000
## DDP3          0.698    0.030   23.314   0.000
## DDP4          0.513    0.040   12.738   0.000

```

```
##
## Intercepts:
##           Estimate Std.Err  z-value  P(>|z|)
##   .DDP1          0.000
##   .DDP2          0.000
##   .DDP3          0.000
##   .DDP4          0.000
##   psyctcsm       0.000
##
## Thresholds:
##           Estimate Std.Err  z-value  P(>|z|)
##   DDP1|t1      -0.716    0.062  -11.591    0.000
##   DDP1|t2      -0.040    0.056   -0.716    0.474
##   DDP1|t3       0.296    0.057    5.185    0.000
##   DDP1|t4       1.034    0.069   15.065    0.000
##   DDP2|t1      -0.580    0.060   -9.693    0.000
##   DDP2|t2       0.208    0.057    3.668    0.000
##   DDP2|t3       0.562    0.060    9.431    0.000
##   DDP2|t4       1.096    0.070   15.570    0.000
##   DDP3|t1      -1.203    0.074  -16.298    0.000
##   DDP3|t2      -0.425    0.058   -7.318    0.000
##   DDP3|t3       0.081    0.056    1.432    0.152
##   DDP3|t4       0.913    0.066   13.909    0.000
##   DDP4|t1      -1.605    0.092  -17.382    0.000
##   DDP4|t2      -1.060    0.069  -15.285    0.000
##   DDP4|t3      -0.521    0.059   -8.817    0.000
##   DDP4|t4       0.431    0.058    7.406    0.000
##
## Variances:
##           Estimate Std.Err  z-value  P(>|z|)
##   .DDP1          0.201
##   .DDP2          0.433
##   .DDP3          0.513
##   .DDP4          0.737
##   psyctcsm       1.000
##
## Scales y*:
##           Estimate Std.Err  z-value  P(>|z|)
##   DDP1           1.000
##   DDP2           1.000
##   DDP3           1.000
##   DDP4           1.000
```

Obtain omega-u-cat as an estimate of the reliability of the total score for the Psychotcism items as a measure of the 'psyctcsm' factor:

```
reliability(fit1f)
```

```
## For constructs with categorical indicators, the alpha and the average  
variance extracted are calculated from polychoric (polyserial) correlations,  
not from Pearson correlations.
```

```
##          psyctcsm  
## alpha  0.8007496  
## omega  0.7902953  
## omega2 0.7902953  
## omega3 0.7932682  
## avevar 0.5289638
```

Although the message in red font in the output above indicates that alpha is calculated based on polychoric correlations, rest assured that the omega estimates are also omega-u-cat estimates obtained by fitting the one-factor model to polychoric correlations. The purpose of the message is to prevent confusion about the difference between $\alpha = .77$ for this scale and ordinal $\alpha = .80$, as described in the article.

Call the `ci.reliability` function to obtain a bootstrap 95% CI for omega. Again, the number of bootstrap samples is set to only 100 (the default number of samples is 10,000, which may take a long time to run). Also, due to random sampling variability inherent to the bootstrap procedure, the resulting CI from the code below may not exactly match the CI reported in the actual article:

```
ci.reliability(data=potic, type = "categorical", interval.type = "perc", B =  
100)  
  
## $est  
## [1] 0.7932682  
##  
## $se  
## [1] 0.01490839  
##  
## $ci.lower  
## [1] 0.7618729  
##  
## $ci.upper  
## [1] 0.8196438  
##  
## $conf.level  
## [1] 0.95  
##  
## $type  
## [1] "categorical omega"  
##  
## $interval.type  
## [1] "percentile bootstrap"
```

Obtaining omega-h from a bifactor model for the Psychological Cost Scale

Download the pcs.csv data from the web to create a data frame called 'pcs' or import the pcs.csv data file if you have already saved it to the same directory as this .rmd file

```
pcs <- read.csv("https://osf.io/xd2tu/download")
```

#OR

```
#pcs <- read.csv("pcs.csv")
```

Specify and estimate the bifactor model for the PCS items:

```
modBf <- 'gen =~ TE1 + TE2 + TE3 + TE4 + TE5 + OE1 + OE2 + OE3 + OE4
          + LVA1 + LVA2 + LVA3 + LVA4 + EM1 + EM2 + EM3 + EM4 + EM5 + EM6
s1 =~ TE1 + TE2 + TE3 + TE4 + TE5
s2 =~ OE1 + OE2 + OE3 + OE4
s3 =~ LVA1 + LVA2 + LVA3 + LVA4
s4 =~ EM1 + EM2 + EM3 + EM4 + EM5 + EM6
          '
```

```
fitBf <- cfa(modBf, data=pcs, std.lv=T, estimator='MLR', orthogonal=T)
summary(fitBf, fit.measures=T)
```

```
## lavaan 0.6-5 ended normally after 36 iterations
```

```
##
```

```
##   Estimator                               ML
```

```
##   Optimization method                     NLMINB
```

```
##   Number of free parameters                57
```

```
##
```

```
##                                     Used      Total
```

```
##   Number of observations                154      172
```

```
##
```

```
## Model Test User Model:
```

```
##                                     Standard      Robust
```

```
##   Test Statistic                       211.382    182.509
```

```
##   Degrees of freedom                    133        133
```

```
##   P-value (Chi-square)                  0.000        0.003
```

```
##   Scaling correction factor              1.158
```

```
##     for the Yuan-Bentler correction (Mplus variant)
```

```
##
```

```
## Model Test Baseline Model:
```

```
##
```

```
##   Test statistic                       2799.877    2260.239
```

```
##   Degrees of freedom                    171        171
```

```
##   P-value                              0.000        0.000
```

```
##   Scaling correction factor              1.239
```

```
##
```

```
## User Model versus Baseline Model:
```

```
##
```

```
##   Comparative Fit Index (CFI)           0.970        0.976
```

```
##   Tucker-Lewis Index (TLI)             0.962        0.970
```

```

##
## Robust Comparative Fit Index (CFI) 0.978
## Robust Tucker-Lewis Index (TLI) 0.972
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -3456.819 -3456.819
## Scaling correction factor 1.269
## for the MLR correction
## Loglikelihood unrestricted model (H1) -3351.128 -3351.128
## Scaling correction factor 1.191
## for the MLR correction
##
## Akaike (AIC) 7027.638 7027.638
## Bayesian (BIC) 7200.744 7200.744
## Sample-size adjusted Bayesian (BIC) 7020.331 7020.331
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.062 0.049
## 90 Percent confidence interval - lower 0.046 0.031
## 90 Percent confidence interval - upper 0.077 0.065
## P-value RMSEA <= 0.05 0.108 0.520
##
## Robust RMSEA 0.053
## 90 Percent confidence interval - lower 0.032
## 90 Percent confidence interval - upper 0.071
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.038 0.038
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|)
## gen =~
## TE1 1.041 0.071 14.715 0.000
## TE2 1.045 0.084 12.430 0.000
## TE3 0.848 0.090 9.392 0.000
## TE4 0.984 0.075 13.097 0.000
## TE5 1.004 0.082 12.264 0.000
## OE1 0.787 0.083 9.537 0.000
## OE2 0.819 0.080 10.207 0.000
## OE3 0.738 0.089 8.319 0.000
## OE4 0.742 0.087 8.512 0.000

```

```

##      LVA1          0.937    0.084    11.180    0.000
##      LVA2          0.863    0.071    12.205    0.000
##      LVA3          0.816    0.085     9.649    0.000
##      LVA4          0.865    0.098     8.802    0.000
##      EM1           0.968    0.095    10.215    0.000
##      EM2           0.930    0.078    11.957    0.000
##      EM3           0.959    0.091    10.542    0.000
##      EM4           0.885    0.091     9.679    0.000
##      EM5           1.043    0.086    12.121    0.000
##      EM6           1.108    0.100    11.054    0.000
##      s1 =~
##      TE1           0.351    0.114     3.071    0.002
##      TE2           0.451    0.144     3.142    0.002
##      TE3           0.402    0.170     2.360    0.018
##      TE4           0.162    0.111     1.457    0.145
##      TE5           0.269    0.154     1.747    0.081
##      s2 =~
##      OE1           0.626    0.107     5.860    0.000
##      OE2           0.516    0.096     5.399    0.000
##      OE3           0.673    0.107     6.291    0.000
##      OE4           0.739    0.085     8.738    0.000
##      s3 =~
##      LVA1          0.253    0.107     2.357    0.018
##      LVA2          0.573    0.081     7.081    0.000
##      LVA3          0.422    0.104     4.051    0.000
##      LVA4          0.528    0.104     5.074    0.000
##      s4 =~
##      EM1           0.506    0.152     3.324    0.001
##      EM2           0.346    0.086     4.026    0.000
##      EM3           0.567    0.121     4.682    0.000
##      EM4           0.562    0.098     5.707    0.000
##      EM5           0.479    0.097     4.930    0.000
##      EM6           0.651    0.148     4.403    0.000
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|)
##      gen ~~
##      s1          0.000
##      s2          0.000
##      s3          0.000
##      s4          0.000
##      s1 ~~
##      s2          0.000
##      s3          0.000
##      s4          0.000
##      s2 ~~
##      s3          0.000
##      s4          0.000
##      s3 ~~
##      s4          0.000

```

```
##
## Variances:
##           Estimate Std.Err z-value P(>|z|)
##   .TE1           0.318   0.067   4.777   0.000
##   .TE2           0.434   0.088   4.913   0.000
##   .TE3           0.501   0.104   4.827   0.000
##   .TE4           0.397   0.066   5.987   0.000
##   .TE5           0.421   0.073   5.799   0.000
##   .OE1           0.546   0.106   5.129   0.000
##   .OE2           0.451   0.075   6.003   0.000
##   .OE3           0.490   0.117   4.183   0.000
##   .OE4           0.263   0.071   3.718   0.000
##   .LVA1          0.442   0.064   6.873   0.000
##   .LVA2          0.100   0.062   1.607   0.108
##   .LVA3          0.443   0.075   5.892   0.000
##   .LVA4          0.523   0.088   5.954   0.000
##   .EM1           0.639   0.100   6.388   0.000
##   .EM2           0.365   0.060   6.049   0.000
##   .EM3           0.482   0.090   5.383   0.000
##   .EM4           0.364   0.083   4.395   0.000
##   .EM5           0.318   0.063   5.032   0.000
##   .EM6           0.534   0.117   4.564   0.000
##   gen            1.000
##   s1             1.000
##   s2             1.000
##   s3             1.000
##   s4             1.000
```

Obtain omega-h as for the PCS total score as a measure of a general ‘psychological cost’ construct:

```
reliability(fitBf)
```

```
##           gen           s1           s2           s3           s4
## alpha  0.9638781 0.92504205 0.8992820 0.9052459 0.9405882
## omega  0.9741033 0.56377307 0.7884791 0.6766430 0.7816839
## omega2 0.9094893 0.09237594 0.3666293 0.1880759 0.2054075
## omega3 0.9077636 0.09240479 0.3666634 0.1878380 0.2053012
## avevar      NA           NA           NA           NA           NA
```

Obtaining omega-ho from a higher-order model for the Psychological Cost Scale

This analysis uses the same pcs data frame used above.

Specify the higher-order factor model for the PCS items:

```
homod <- 'TE =~ TE1 + TE2 + TE3 + TE4 + TE5
          OE =~ OE1 + OE2 + OE3 + OE4
          LV =~ LVA1 + LVA2 + LVA3 + LVA4'
```

```
EM =~ EM1 + EM2 + EM3 + EM4 + EM5 + EM6
cost =~ TE + OE + LV + EM'
```

Estimate the model and get the results:

```
fitHo <- cfa(homod, data=pcs, std.lv=T, estimator='MLM')
summary(fitHo, fit.measures=T)
```

```
## lavaan 0.6-5 ended normally after 57 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      42
##
##                               Used      Total
##      Number of observations         154      172
##
## Model Test User Model:
##
##                               Standard      Robust
##      Test Statistic                243.444    185.699
##      Degrees of freedom                148      148
##      P-value (Chi-square)              0.000      0.019
##      Scaling correction factor                    1.311
##      for the Satorra-Bentler correction
##
## Model Test Baseline Model:
##
##      Test statistic                2799.877    2282.637
##      Degrees of freedom                171      171
##      P-value                          0.000      0.000
##      Scaling correction factor                    1.227
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)                0.964      0.982
##      Tucker-Lewis Index (TLI)                   0.958      0.979
##
##      Robust Comparative Fit Index (CFI)                    0.981
##      Robust Tucker-Lewis Index (TLI)                      0.978
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)                -3472.850    -3472.850
##      Loglikelihood unrestricted model (H1)          -3351.128    -3351.128
##
##      Akaike (AIC)                                7029.700      7029.700
##      Bayesian (BIC)                                7157.252      7157.252
##      Sample-size adjusted Bayesian (BIC)           7024.315      7024.315
##
## Root Mean Square Error of Approximation:
```



```

##
## RMSEA                                0.065      0.041
## 90 Percent confidence interval - lower 0.050      0.021
## 90 Percent confidence interval - upper 0.079      0.056
## P-value RMSEA <= 0.05                 0.052      0.832
##
## Robust RMSEA                                0.047
## 90 Percent confidence interval - lower 0.020
## 90 Percent confidence interval - upper 0.066
##
## Standardized Root Mean Square Residual:
##
## SRMR                                0.045      0.045
##
## Parameter Estimates:
##
## Information                                Expected
## Information saturated (h1) model          Structured
## Standard errors                          Robust.sem
##
## Latent Variables:
##           Estimate Std.Err z-value P(>|z|)
## TE =~
## TE1      0.356    0.067   5.289   0.000
## TE2      0.364    0.065   5.554   0.000
## TE3      0.299    0.059   5.056   0.000
## TE4      0.323    0.058   5.605   0.000
## TE5      0.338    0.065   5.198   0.000
## OE =~
## OE1      0.641    0.074   8.642   0.000
## OE2      0.616    0.060  10.334   0.000
## OE3      0.630    0.066   9.523   0.000
## OE4      0.647    0.064  10.078   0.000
## LV =~
## LVA1     0.442    0.056   7.836   0.000
## LVA2     0.457    0.059   7.751   0.000
## LVA3     0.427    0.059   7.286   0.000
## LVA4     0.460    0.056   8.258   0.000
## EM =~
## EM1      0.509    0.069   7.394   0.000
## EM2      0.462    0.066   7.018   0.000
## EM3      0.517    0.074   6.955   0.000
## EM4      0.483    0.067   7.179   0.000
## EM5      0.535    0.069   7.756   0.000
## EM6      0.595    0.080   7.430   0.000
## cost =~
## TE       2.914    0.595   4.894   0.000
## OE       1.223    0.155   7.880   0.000
## LV       1.951    0.281   6.933   0.000
## EM       1.900    0.327   5.809   0.000

```

```
##
## Variances:
##           Estimate Std.Err z-value P(>|z|)
##   .TE1           0.321   0.063   5.081   0.000
##   .TE2           0.473   0.069   6.837   0.000
##   .TE3           0.531   0.100   5.289   0.000
##   .TE4           0.398   0.073   5.494   0.000
##   .TE5           0.418   0.070   5.935   0.000
##   .OE1           0.531   0.099   5.353   0.000
##   .OE2           0.440   0.081   5.428   0.000
##   .OE3           0.495   0.099   5.001   0.000
##   .OE4           0.315   0.055   5.719   0.000
##   .LVA1          0.443   0.081   5.466   0.000
##   .LVA2          0.170   0.035   4.851   0.000
##   .LVA3          0.412   0.064   6.473   0.000
##   .LVA4          0.533   0.090   5.945   0.000
##   .EM1           0.637   0.085   7.508   0.000
##   .EM2           0.367   0.062   5.872   0.000
##   .EM3           0.493   0.075   6.564   0.000
##   .EM4           0.387   0.064   6.017   0.000
##   .EM5           0.315   0.062   5.059   0.000
##   .EM6           0.554   0.085   6.515   0.000
##   .TE            1.000
##   .OE            1.000
##   .LV            1.000
##   .EM            1.000
##   cost           1.000
```

Obtain omega-ho for the PCS total score as a measure of the 'psychological cost' higher-order factor:

```
reliabilityL2(fitHo, 'cost')

##           omegaL1      omegaL2 partialOmegaL1
##   0.9088176      0.9410190      0.9734520
```

Obtain omega estimates for the subscale scores as measures of the lower-order factors:

```
reliability(fitHo)

## Higher-order factors were ignored.

##           TE           OE           LV           EM
## alpha  0.9250420 0.8992820 0.9052459 0.9405882
## omega  0.9260209 0.9000548 0.9077522 0.9415490
## omega2 0.9260209 0.9000548 0.9077522 0.9415490
## omega3 0.9256773 0.9014397 0.9125259 0.9404921
## avevar 0.7155347 0.6925123 0.7111716 0.7299181
```

Obtaining omega-h and omega-h-ss from a bifactor model for the Psychological Cost Scale

Specify and estimate the bifactor model for the PCS items:

```
modBf <- 'gen =~ TE1 + TE2 + TE3 + TE4 + TE5 + OE1 + OE2 + OE3 + OE4
          + LVA1 + LVA2 + LVA3 + LVA4 + EM1 + EM2 + EM3 + EM4 + EM5 + EM6
s1 =~ TE1 + TE2 + TE3 + TE4 + TE5
s2 =~ OE1 + OE2 + OE3 + OE4
s3 =~ LVA1 + LVA2 + LVA3 + LVA4
s4 =~ EM1 + EM2 + EM3 + EM4 + EM5 + EM6
          '

fitBf <- cfa(modBf, data=pcs, std.lv=T, estimator='MLR', orthogonal=T)
summary(fitBf, fit.measures=T)
```

lavaan 0.6-5 ended normally after 36 iterations

##

## Estimator	ML	
## Optimization method	NLMINB	
## Number of free parameters	57	
##		
##	Used	Total
## Number of observations	154	172
##		

Model Test User Model:

##	Standard	Robust
## Test Statistic	211.382	182.509
## Degrees of freedom	133	133
## P-value (Chi-square)	0.000	0.003
## Scaling correction factor		1.158
## for the Yuan-Bentler correction (Mplus variant)		
##		

Model Test Baseline Model:

##		
## Test statistic	2799.877	2260.239
## Degrees of freedom	171	171
## P-value	0.000	0.000
## Scaling correction factor		1.239
##		

User Model versus Baseline Model:

##		
## Comparative Fit Index (CFI)	0.970	0.976
## Tucker-Lewis Index (TLI)	0.962	0.970
##		
## Robust Comparative Fit Index (CFI)		0.978
## Robust Tucker-Lewis Index (TLI)		0.972
##		

Loglikelihood and Information Criteria:

##		
## Loglikelihood user model (H0)	-3456.819	-3456.819

```

## Scaling correction factor 1.269
## for the MLR correction
## Loglikelihood unrestricted model (H1) -3351.128 -3351.128
## Scaling correction factor 1.191
## for the MLR correction
##
## Akaike (AIC) 7027.638 7027.638
## Bayesian (BIC) 7200.744 7200.744
## Sample-size adjusted Bayesian (BIC) 7020.331 7020.331
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.062 0.049
## 90 Percent confidence interval - lower 0.046 0.031
## 90 Percent confidence interval - upper 0.077 0.065
## P-value RMSEA <= 0.05 0.108 0.520
##
## Robust RMSEA 0.053
## 90 Percent confidence interval - lower 0.032
## 90 Percent confidence interval - upper 0.071
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.038 0.038
##
## Parameter Estimates:
##
## Information Observed
## Observed information based on Hessian
## Standard errors Robust.huber.white
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|)
## gen =~
## TE1 1.041 0.071 14.715 0.000
## TE2 1.045 0.084 12.430 0.000
## TE3 0.848 0.090 9.392 0.000
## TE4 0.984 0.075 13.097 0.000
## TE5 1.004 0.082 12.264 0.000
## OE1 0.787 0.083 9.537 0.000
## OE2 0.819 0.080 10.207 0.000
## OE3 0.738 0.089 8.319 0.000
## OE4 0.742 0.087 8.512 0.000
## LVA1 0.937 0.084 11.180 0.000
## LVA2 0.863 0.071 12.205 0.000
## LVA3 0.816 0.085 9.649 0.000
## LVA4 0.865 0.098 8.802 0.000
## EM1 0.968 0.095 10.215 0.000
## EM2 0.930 0.078 11.957 0.000
## EM3 0.959 0.091 10.542 0.000

```

```

##      EM4      0.885      0.091      9.679      0.000
##      EM5      1.043      0.086     12.121      0.000
##      EM6      1.108      0.100     11.054      0.000
##    s1 =~
##      TE1      0.351      0.114      3.071      0.002
##      TE2      0.451      0.144      3.142      0.002
##      TE3      0.402      0.170      2.360      0.018
##      TE4      0.162      0.111      1.457      0.145
##      TE5      0.269      0.154      1.747      0.081
##    s2 =~
##      OE1      0.626      0.107      5.860      0.000
##      OE2      0.516      0.096      5.399      0.000
##      OE3      0.673      0.107      6.291      0.000
##      OE4      0.739      0.085      8.738      0.000
##    s3 =~
##      LVA1      0.253      0.107      2.357      0.018
##      LVA2      0.573      0.081      7.081      0.000
##      LVA3      0.422      0.104      4.051      0.000
##      LVA4      0.528      0.104      5.074      0.000
##    s4 =~
##      EM1      0.506      0.152      3.324      0.001
##      EM2      0.346      0.086      4.026      0.000
##      EM3      0.567      0.121      4.682      0.000
##      EM4      0.562      0.098      5.707      0.000
##      EM5      0.479      0.097      4.930      0.000
##      EM6      0.651      0.148      4.403      0.000
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|)
##    gen ~~
##      s1      0.000
##      s2      0.000
##      s3      0.000
##      s4      0.000
##    s1 ~~
##      s2      0.000
##      s3      0.000
##      s4      0.000
##    s2 ~~
##      s3      0.000
##      s4      0.000
##    s3 ~~
##      s4      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|)
##      .TE1      0.318      0.067      4.777      0.000
##      .TE2      0.434      0.088      4.913      0.000
##      .TE3      0.501      0.104      4.827      0.000
##      .TE4      0.397      0.066      5.987      0.000

```

```
##      .TE5                0.421    0.073    5.799    0.000
##      .OE1                0.546    0.106    5.129    0.000
##      .OE2                0.451    0.075    6.003    0.000
##      .OE3                0.490    0.117    4.183    0.000
##      .OE4                0.263    0.071    3.718    0.000
##      .LVA1               0.442    0.064    6.873    0.000
##      .LVA2               0.100    0.062    1.607    0.108
##      .LVA3               0.443    0.075    5.892    0.000
##      .LVA4               0.523    0.088    5.954    0.000
##      .EM1                0.639    0.100    6.388    0.000
##      .EM2                0.365    0.060    6.049    0.000
##      .EM3                0.482    0.090    5.383    0.000
##      .EM4                0.364    0.083    4.395    0.000
##      .EM5                0.318    0.063    5.032    0.000
##      .EM6                0.534    0.117    4.564    0.000
##      gen                 1.000
##      s1                  1.000
##      s2                  1.000
##      s3                  1.000
##      s4                  1.000
```

Obtain omega-h-ss for the PCS subscale scores as measures of constructs orthogonal to the general factor, and omega-h as for the PCS total score as a measure of a general 'psychological cost' construct:

```
reliability(fitBf)
```

```
##      gen      s1      s2      s3      s4
## alpha 0.9638781 0.92504205 0.8992820 0.9052459 0.9405882
## omega 0.9741033 0.56377307 0.7884791 0.6766430 0.7816839
## omega2 0.9094893 0.09237594 0.3666293 0.1880759 0.2054075
## omega3 0.9077636 0.09240479 0.3666634 0.1878380 0.2053012
## avevar      NA      NA      NA      NA      NA
```

Using the psych package to calculate omega estimates

****Use the 'omega' function to estimate omega-h based on an exploratory bifactor model of the PCS items (this bifactor model is based on the Schmid-Leiman transformation, and so the resulting omega-h is equivalent to omega-ho obtained from an exploratory higher-order model):***

```
omega(pcs, nfactors = 4, plot = F)
```

```
## Loading required namespace: GPArotation
```

```
## Omega
## Call: omega(m = pcs, nfactors = 4, plot = F)
## Alpha:      0.96
## G.6:       0.98
```

```

## Omega Hierarchical:      0.85
## Omega H asymptotic:     0.88
## Omega Total              0.98
##
## Schmid Leiman Factor loadings greater than 0.2
##      g    F1*   F2*   F3*   F4*   h2   u2   p2
## TE1  0.81                0.39      0.81 0.19 0.80
## TE2  0.76                0.40      0.75 0.25 0.77
## TE3  0.70                0.35      0.62 0.38 0.80
## TE4  0.78                0.29      0.72 0.28 0.84
## TE5  0.77                0.34      0.74 0.26 0.80
## OE1  0.61            0.55                0.67 0.33 0.56
## OE2  0.67            0.46                0.66 0.34 0.68
## OE3  0.61            0.54                0.68 0.32 0.54
## OE4  0.63            0.62                0.80 0.20 0.50
## LVA1 0.75                0.28 0.67 0.33 0.85
## LVA2 0.80                0.48 0.87 0.13 0.73
## LVA3 0.71            0.20                0.37 0.71 0.29 0.72
## LVA4 0.70                0.43 0.67 0.33 0.72
## EM1  0.69  0.39                0.66 0.34 0.73
## EM2  0.76  0.38                0.75 0.25 0.77
## EM3  0.72  0.47                0.73 0.27 0.71
## EM4  0.72  0.48                0.74 0.26 0.70
## EM5  0.78  0.43                0.81 0.19 0.76
## EM6  0.73  0.48                0.78 0.22 0.68
##
## With eigenvalues of:
##      g    F1*   F2*   F3*   F4*
## 9.95 1.21 1.29 0.70 0.68
##
## general/max 7.74    max/min = 1.88
## mean percent general = 0.72    with sd = 0.1 and cv of 0.13
## Explained Common Variance of the general factor = 0.72
##
## The degrees of freedom are 101 and the fit is 1
## The number of observations was 172 with Chi Square = 161.11 with prob
< 0.00013
## The root mean square of the residuals is 0.02
## The df corrected root mean square of the residuals is 0.03
## RMSEA index = 0.064 and the 10 % confidence intervals are 0.041 0.076
## BIC = -358.79
##
## Compare this with the adequacy of just a general factor and no group
factors
## The degrees of freedom for just the general factor are 152 and the fit is
4.49
## The number of observations was 172 with Chi Square = 732.57 with prob
< 8.2e-77
## The root mean square of the residuals is 0.1
## The df corrected root mean square of the residuals is 0.11

```

```

##
## RMSEA index = 0.154 and the 10 % confidence intervals are 0.139 0.16
## BIC = -49.85
##
## Measures of factor score adequacy
##
##           g  F1*  F2*  F3*  F4*
## Correlation of scores with factors 0.93 0.79 0.86 0.72 0.77
## Multiple R square of scores with factors 0.86 0.63 0.73 0.52 0.59
## Minimum correlation of factor score estimates 0.73 0.26 0.47 0.04 0.18
##
## Total, General and Subset omega for each subset
##
##           g  F1*  F2*  F3*  F4*
## Omega total for total scores and subscales 0.98 0.94 0.90 0.92 0.9
## Omega general for total scores and subscales 0.85 0.69 0.52 0.76 0.7
## Omega group for total scores and subscales 0.08 0.25 0.38 0.16 0.2

```