

MID - Measurement Model Code

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The platform used to write and generate this .html document is `cat(platform)` on RStudio, `cat(r_info)`. The required packages to run all necessary code chunks are checked and installed using the `pacman` R package. `p_load` from the package is used to load and install all listed packages. The details for a number of packages are described before their use. If they are not, the `.rmd` file will provide the list of packages that are installed.

Simulating data

This section is unique only to the Stage 1 registered reported. We are not using any real data, so to ensure that each step is completed correctly, we use simulated data. We can operate under the assumption that all participants will have no missing values given how exclusion criteria are applied and fMRI data are extracted from the BOLD timeseries.

In this section, a population model is fit reflecting the approach and avoidance (measured constructs) that are [presumed] to be associated with stimuli within/across tasks (Manifest variables). As formalized here, in the MID task the contrasts reflect the manifest variables (i.e., reflective model) as opposed of the other way, whereby the items form the variables predict the construct (i.e., formative model).

Specify Population Model

Start off by specifying the **population** model. In this scenario, the individual runs load onto the specific Contrast and ROI combinations. Then, the ROIs are loaded onto the factors **approach** and **avoidance**. The approach and avoidance are specified as *negatively* correlated and the factor variances are fixed to 1. This is because in the multidimensional circumplex model, when the span of the contrasts are considered, these measures will likely not be orthogonal but instead >90 degree in space.

```
population_model<-'
# By run loadings for bilateral regions
AWin_v_Neut_L_NAcc =~      .7*AWin_v_Neut_L_NAcc_run1      + .7*AWin_v_Neut_L_NAcc_run2
AWin_v_Neut_L_Insula =~    .7*AWin_v_Neut_L_Insula_run1    + .7*AWin_v_Neut_L_Insula_run2
BWin_v_Neut_L_NAcc =~      .7*BWin_v_Neut_L_NAcc_run1      + .7*BWin_v_Neut_L_NAcc_run2
BWin_v_Neut_L_Insula =~    .7*BWin_v_Neut_L_Insula_run1    + .7*BWin_v_Neut_L_Insula_run2
BWin_v_BLose_L_NAcc =~     .7*BWin_v_BLose_L_NAcc_run1     + .7*BWin_v_BLose_L_NAcc_run2
BWin_v_BLose_L_Insula =~   .7*BWin_v_BLose_L_Insula_run1   + .7*BWin_v_BLose_L_Insula_run2
ALose_v_Neut_L_NAcc =~     .7*ALose_v_Neut_L_NAcc_run1     + .7*ALose_v_Neut_L_NAcc_run2
ALose_v_Neut_L_Insula =~   .7*ALose_v_Neut_L_Insula_run1   + .7*ALose_v_Neut_L_Insula_run2
BLose_v_Neut_L_NAcc =~     .7*BLose_v_Neut_L_NAcc_run1     + .7*BLose_v_Neut_L_NAcc_run2
BLose_v_Neut_L_Insula =~   .7*BLose_v_Neut_L_Insula_run1   + .7*BLose_v_Neut_L_Insula_run2
BLose_v_BWin_L_NAcc =~     .7*BLose_v_BWin_L_NAcc_run1     + .7*BLose_v_BWin_L_NAcc_run2
BLose_v_BWin_L_Insula =~   .7*BLose_v_BWin_L_Insula_run1   + .7*BLose_v_BWin_L_Insula_run2

AWin_v_Neut_R_NAcc =~      .7*AWin_v_Neut_R_NAcc_run1      + .7*AWin_v_Neut_R_NAcc_run2
AWin_v_Neut_R_Insula =~    .7*AWin_v_Neut_R_Insula_run1    + .7*AWin_v_Neut_R_Insula_run2
BWin_v_Neut_R_NAcc =~      .7*BWin_v_Neut_R_NAcc_run1      + .7*BWin_v_Neut_R_NAcc_run2
BWin_v_Neut_R_Insula =~    .7*BWin_v_Neut_R_Insula_run1    + .7*BWin_v_Neut_R_Insula_run2
BWin_v_BLose_R_NAcc =~     .7*BWin_v_BLose_R_NAcc_run1     + .7*BWin_v_BLose_R_NAcc_run2
BWin_v_BLose_R_Insula =~   .7*BWin_v_BLose_R_Insula_run1   + .7*BWin_v_BLose_R_Insula_run2
ALose_v_Neut_R_NAcc =~     .7*ALose_v_Neut_R_NAcc_run1     + .7*ALose_v_Neut_R_NAcc_run2
ALose_v_Neut_R_Insula =~   .7*ALose_v_Neut_R_Insula_run1   + .7*ALose_v_Neut_R_Insula_run2
BLose_v_Neut_R_NAcc =~     .7*BLose_v_Neut_R_NAcc_run1     + .7*BLose_v_Neut_R_NAcc_run2
BLose_v_Neut_R_Insula =~   .7*BLose_v_Neut_R_Insula_run1   + .7*BLose_v_Neut_R_Insula_run2
BLose_v_BWin_R_NAcc =~     .7*BLose_v_BWin_R_NAcc_run1     + .7*BLose_v_BWin_R_NAcc_run2
BLose_v_BWin_R_Insula =~   .7*BLose_v_BWin_R_Insula_run1   + .7*BLose_v_BWin_R_Insula_run2

#Factor item loadings
Approach =~      .8*AWin_v_Neut_L_NAcc + .8*AWin_v_Neut_R_NAcc + .45*AWin_v_Neut_R_Insula +
                 .7*BWin_v_Neut_L_NAcc + .7*BWin_v_Neut_R_NAcc + .4*BWin_v_Neut_R_Insula +
```

```

      .8*BWin_v_BLose_L_NAcc + .8*BWin_v_BLose_R_NAcc

Avoid =~ .8*ALose_v_Neut_L_Insula + .8*ALose_v_Neut_R_Insula +
        .75*BLose_v_Neut_L_Insula + .75*BLose_v_Neut_R_Insula +
        .8*BLose_v_BWin_L_Insula + .45*BLose_v_BWin_R_Insula

# Factor Covariances
Approach ~~ -.6*Avoid

# Fixing factor variances
Approach ~~ 1*Approach
Avoid ~~ 1*Avoid

'

```

General samples

Using the population model, `simsem` is used to create simulated data. This generates a **fake** dataset that is used to pilot the planned Multigroup CFA, ESEM, EFA and Local SEM models

In this case, 50 repetitions are simulated per model for an *approximate* N sample for each study. Even though the factor variances are specified in the population model as '1', this model fixes all latent variables using `std.lv = TRUE`.

1. AHRB N = 108
2. MLS N = 159
3. ABCD N = 1000

```

set.seed(25151215)
sim_AHRB <- simsem::sim(nRep = 50, model = "lavaan", n = 108,
  generate = population_model, std.lv = TRUE, lavaanfun = "sem",
  # std.lv ~ ix the variances of all the latent variables
  dataOnly=T, meanstructure = FALSE, seed=123)

sim_MLS <- simsem::sim(nRep = 50, model = "lavaan", n = 159,
  generate = population_model, std.lv = TRUE, lavaanfun = "sem",
  dataOnly=T, meanstructure = FALSE, seed=123)

sim_ABCD <- simsem::sim(nRep = 50, model = "lavaan", n = 1000,
  generate = population_model, std.lv = TRUE, lavaanfun = "sem",
  dataOnly=T, meanstructure = FALSE, seed=123)

```

Average each repetition for each simulated sample. For example, after 50 repetitions of 1000 participants for the population model of ABCD sample, an average estimate is derived using `aaply`. For each study, the `set` variable is created to differentiate which sample the data is associated with (i.e., grouping variable). In this case, AHRB = 3, MLS = 2, ABCD = 1.

```

sim_AHRB_data <- data.frame(aaply(laply(sim_AHRB, as.matrix), c(2,3), mean))
  sim_AHRB_data$set <-3

sim_MLS_data <- data.frame(aaply(laply(sim_MLS, as.matrix), c(2,3), mean))
  sim_MLS_data$set <-2

sim_ABCD_data <- data.frame(aaply(laply(sim_ABCD, as.matrix), c(2,3), mean))
  sim_ABCD_data$set <-1

```

Next, row bind the data sets to form one complete data. This equivalent to stacking the datasets and only retaining a single header row (since all variable names are constant). This creates a 1267 x 49 data matrix (48 brain variables + sample label)

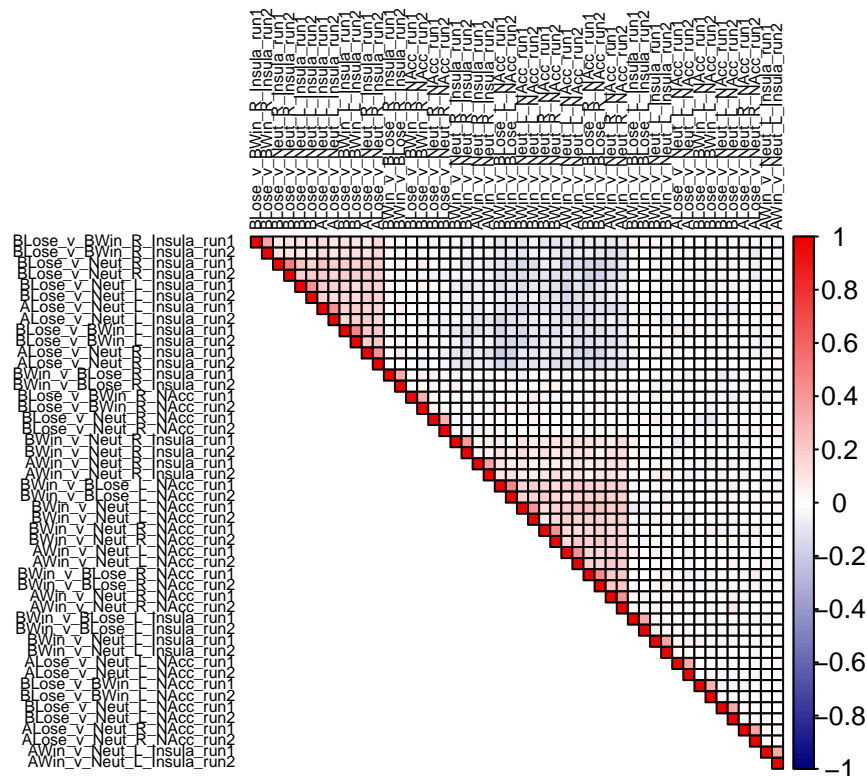
```
brain_set <- rbind(sim_AHRB_data,sim_MLS_data,sim_ABCD_data)
```

Correlation matrix of data

Here, a combination of rcorr and corrplot are used to visualize the associations among the brain variables in the dataset.

```
# Using Hmisc to create a 24x24 matrix for a list (3) that contains: the pearson's r corr, sample size
Brain_corr = rcorr(as.matrix(subset(brain_set,select=-c(set))), # excluding the set of data related to
                    type = "pearson")

# Using corrplot() to create heatmap of the data.
par(mfrow=c(1,1))
corrplot(Brain_corr$r, type = "upper",
          order = 'hclust',
          method = "color",
          tl.cex = 0.5, tl.col = 'black',
          cl.pos = 'r', tl.pos = 'lt', outline = TRUE,
          col=colorRampPalette(c("navyblue","white","red2"))(100),# colours http://www.stat.columbia.edu,
          mar = c(2,.15,.25,.15)#bottom, left, top and right,
          )
```



Running [restricted] Multigroup CFA

Run the CFA multi-group analysis for the three datasets. Multi-group CFA tests the measurement invariance across defined groups to determine whether soft and strict invariance criteria are met and the degree to which the derive estimates for an item in one study can be compared to the same item in another sample. In this case, the focus is on the configural (structure) and metric invariance (loadings). In short, this model evaluates whether factor structure and loadings for the approach and avoidance model are invariant (dont significant differ) across the samples.

Code here is based on measurement invariance models from Maasen et al. 2019, Measurement invariance presentation from Kate Xu and Multi-group CFA tutorial from Hirschfeld & Brachel (2014).

The issue of multi-group is invariance what is discussed in Borsboom (2006). In short, (1) Interpretation of group differences on observed scores DEPENDS on the invariance of measurement models & (2) many make conclusions without doing a single test of measurement invariance.

A tutorial on CFA broadly is available from Lizbeth Benzon and Nilam Ram here and tutorial of measurement invariance (in context of longitudinal data) from Nilam Ram is available here

CFA model

The below specified model will be used. The number of estimate parameters are fewer and may be more appropriate for the theoretical model. This model may result in few convergence issues if the number of participants ends up to be few and the coefficients/estimates are lower.(see Kline 2015 book on **Principles and Practice of Structural Equation Modeling**).

```
MID_model <- '  
  
# Factor loadings  
Approach =~ AWin_v_Neut_L_NAcc_run1 + AWin_v_Neut_R_NAcc_run1 + AWin_v_Neut_R_Insula_run1 +  
            AWin_v_Neut_L_NAcc_run2 + AWin_v_Neut_R_NAcc_run2 + AWin_v_Neut_R_Insula_run2 +  
            BWin_v_Neut_L_NAcc_run1 + BWin_v_Neut_R_NAcc_run1 + BWin_v_Neut_R_Insula_run1 +  
            BWin_v_Neut_L_NAcc_run2 + BWin_v_Neut_R_NAcc_run2 + BWin_v_Neut_R_Insula_run2 +  
            BWin_v_BLose_L_NAcc_run1 + BWin_v_BLose_R_NAcc_run1 +  
            BWin_v_BLose_L_NAcc_run2 + BWin_v_BLose_R_NAcc_run2  
  
Avoid =~    ALoose_v_Neut_L_Insula_run1 + ALoose_v_Neut_R_Insula_run1 +  
            ALoose_v_Neut_L_Insula_run2 + ALoose_v_Neut_R_Insula_run2 +  
            BLoose_v_Neut_L_Insula_run1 + BLoose_v_Neut_R_Insula_run1 +  
            BLoose_v_Neut_L_Insula_run2 + BLoose_v_Neut_R_Insula_run2 +  
            BLoose_v_BWin_L_Insula_run1 + BLoose_v_BWin_R_Insula_run1 +  
            BLoose_v_BWin_L_Insula_run2 + BLoose_v_BWin_R_Insula_run2  
  
,
```

Running CFA: Three Samples

Below is the CFA model that is used to test the proposed restricted model (see Figure 1 in the manuscript). The CFA fitting procedure is consistent with the description here. For each CFA model, the full sample is filtered for each type sample, e.g. AHRB, MLS, ABCD. The `std.lv = TRUE` constrain the latent factor variances to 1. The estimator being used is MLR, a maximum likelihood robust estimator. A CFA model is estimated for the complete data (i.e., all three datasets).

```
all_sample <- cfa(model = MID_model, data = brain_set,  
                  estimator = "MLR", std.lv = TRUE, meanstructure = TRUE)
```

Fitting Configural CFA

Here, the configural multigroup model is fit. As described in D'Urso et al. (2022) measurement invariance pre-print, the configural model tests:

is the structure of the factors is invariant across the samples ('set'). In other words, if we *a priori* propose a two-factor structure (FA 1 = approach and FA 2 = Avoidance), does this two factor structure represent the between-person variability in the items that reflect the factors across each sample?

If the variability in one sample suggests a one, three, or four factor structure, this will be degrade the fit statistics.

A pre-specified CFA model is used to evaluate whether the measures/items that reflect the factor are the same across groups. `group= 'set'` is used to define the grouping variable. All loadings and intercepts are free to vary across groups, and the factor variance is set to '1' via `std.lv = TRUE`

```
configural_cfa <- cfa(model = MID_model, data = brain_set, group = 'set',  
                      estimator = "MLR", std.lv = TRUE, meanstructure = TRUE)
```

Fitting Metric CFA

After fitting the CFA configural (factor structure) invariance, if the model fit is not poor, then the next step is to test the metric invariance. Metric invariance tests:

are the loadings are consistent across the groups. In other words,are the phenomena (i.e., approach and avoidance) reflected by the same pattern across the measures/items?

One cause for concern may be that the phenomenon are not invariant across age groups, in that the items/measures (ROIs for a given contrast) do not load in the same manner onto each factor. This 'soft' measure of invariance can determine whether the items functions differ across the items and so cannot be easily compared.

The model is fit using the same procedure as for configural invariance with one exception: In metric invariance the loadings group equality constraint is added to the model via `group.equal=c("loadings")`. The model fit statistics are used to evaluate whether the fit is poor.

```
metric_cfa <-cfa(model = MID_model, data = brain_set,  
                 group = 'set', group.equal=c("loadings"),  
                 estimator = "MLR", std.lv = TRUE, meanstructure = TRUE)
```

Extracting Fit Statistics

Once the above models are fit, the following information is pulled out and saved into a out data frame:

1. Model name
2. Chi-square statistics
3. Model Degrees of Freedom (df)
4. Model p-value
5. RMSEA
6. CFI
7. SRMR
8. AIC
9. BIC

```
# Below selects specific fit data as described in Maassen et al. 2019 OSF. No comparisons are made to c  
out <- matrix(NA, ncol = 10, nrow = 4)  
colnames(out) <- c("model","chisq","df","pvalue", "rmsea", "cfi","tli", "srmr",  
                  "AIC", "BIC")
```

```

# save fit measures from models
out[1,2:8] <- round(data.matrix(fitmeasures(all_sample,
                                         fit.measures = c("chisq","df","pvalue",
                                                         "rmsea", "cfi","tli", "srmr"))),
                    digits=3)

out[2,2:8] <- round(data.matrix(fitmeasures(configural_cfa,
                                         fit.measures = c("chisq","df","pvalue",
                                                         "rmsea", "cfi","tli", "srmr"))),
                    digits=3)

out[3,2:8] <- round(data.matrix(fitmeasures(metric_cfa,
                                         fit.measures = c("chisq","df","pvalue",
                                                         "rmsea", "cfi","tli", "srmr"))),
                    digits=3)

# AIC models
out[1,9] <- round(AIC(all_sample),3)
out[2,9] <- round(AIC(configural_cfa),3)
out[3,9] <- round(AIC(metric_cfa),3)

# BIC models
out[1,10] <- round(BIC(all_sample),3)
out[2,10] <- round(BIC(configural_cfa),3)
out[3,10] <- round(BIC(metric_cfa),3)

out[1:3,1] <- c("Overall CFA", "Configg MG-CFA", "Metric MG-CFA")

```

Model Parameter Summary

Reporting standardized coefficients.

All Sample CFA model

```

##### Summarizing All Samples CFA model #####
parameters(all_sample, standardize = T)

```

```

## # Loading
##
## Link | Coefficient | SE | 95% CI | z | p
## -----
## Approach =~ AWin_v_Neut_L_NAcc_run1 | 0.44 | 0.03 | [0.38, 0.51] | 14.45 | < .001
## Approach =~ AWin_v_Neut_R_NAcc_run1 | 0.45 | 0.03 | [0.40, 0.51] | 15.78 | < .001
## Approach =~ AWin_v_Neut_R_Insula_run1 | 0.22 | 0.03 | [0.16, 0.29] | 7.08 | < .001
## Approach =~ AWin_v_Neut_L_NAcc_run2 | 0.48 | 0.03 | [0.43, 0.54] | 16.93 | < .001
## Approach =~ AWin_v_Neut_R_NAcc_run2 | 0.48 | 0.03 | [0.43, 0.53] | 17.82 | < .001
## Approach =~ AWin_v_Neut_R_Insula_run2 | 0.24 | 0.03 | [0.17, 0.30] | 7.47 | < .001
## Approach =~ BWin_v_Neut_L_NAcc_run1 | 0.48 | 0.03 | [0.42, 0.53] | 17.17 | < .001
## Approach =~ BWin_v_Neut_R_NAcc_run1 | 0.37 | 0.03 | [0.31, 0.43] | 12.53 | < .001
## Approach =~ BWin_v_Neut_R_Insula_run1 | 0.23 | 0.03 | [0.17, 0.29] | 7.32 | < .001
## Approach =~ BWin_v_Neut_L_NAcc_run2 | 0.40 | 0.03 | [0.35, 0.46] | 13.99 | < .001
## Approach =~ BWin_v_Neut_R_NAcc_run2 | 0.41 | 0.03 | [0.36, 0.47] | 14.73 | < .001

```

```
## Approach =~ BWin_v_Neut_R_Insula_run2 |          0.27 | 0.03 | [0.21, 0.33] |    8.30 | < .001
## Approach =~ BWin_v_BLose_L_NAcc_run1 |          0.45 | 0.03 | [0.39, 0.50] |   15.12 | < .001
## Approach =~ BWin_v_BLose_R_NAcc_run1 |          0.47 | 0.03 | [0.41, 0.53] |   16.47 | < .001
## Approach =~ BWin_v_BLose_L_NAcc_run2 |          0.42 | 0.03 | [0.37, 0.48] |   14.19 | < .001
## Approach =~ BWin_v_BLose_R_NAcc_run2 |          0.47 | 0.03 | [0.42, 0.53] |   16.34 | < .001
## Avoid =~ ALose_v_Neut_L_Insula_run1 |          0.46 | 0.03 | [0.40, 0.52] |   15.36 | < .001
## Avoid =~ ALose_v_Neut_R_Insula_run1 |          0.48 | 0.03 | [0.42, 0.53] |   16.58 | < .001
## Avoid =~ ALose_v_Neut_L_Insula_run2 |          0.42 | 0.03 | [0.36, 0.48] |   14.25 | < .001
## Avoid =~ ALose_v_Neut_R_Insula_run2 |          0.47 | 0.03 | [0.41, 0.52] |   16.74 | < .001
## Avoid =~ BLose_v_Neut_L_Insula_run1 |          0.44 | 0.03 | [0.38, 0.50] |   14.51 | < .001
## Avoid =~ BLose_v_Neut_R_Insula_run1 |          0.50 | 0.03 | [0.44, 0.56] |   16.42 | < .001
## Avoid =~ BLose_v_Neut_L_Insula_run2 |          0.41 | 0.03 | [0.35, 0.47] |   13.36 | < .001
## Avoid =~ BLose_v_Neut_R_Insula_run2 |          0.48 | 0.03 | [0.43, 0.54] |   16.06 | < .001
## Avoid =~ BLose_v_BWin_L_Insula_run1 |          0.45 | 0.03 | [0.40, 0.51] |   14.94 | < .001
## Avoid =~ BLose_v_BWin_R_Insula_run1 |          0.27 | 0.03 | [0.21, 0.34] |    8.55 | < .001
## Avoid =~ BLose_v_BWin_L_Insula_run2 |          0.45 | 0.03 | [0.39, 0.51] |   14.91 | < .001
## Avoid =~ BLose_v_BWin_R_Insula_run2 |          0.27 | 0.03 | [0.21, 0.34] |    8.53 | < .001
```

```
##
```

```
## # Correlation
```

```
##
```

```
## Link          | Coefficient | SE |          95% CI |      z |      p
```

```
## -----
```

```
## Approach ~~ Avoid |          -0.55 | 0.03 | [-0.61, -0.49] | -18.77 | < .001
```

Configural CFA model

```
##### Summarizing Configural MG-CFA model #####
```

```
parameters(configural_cfa, standardize = T)
```

```
## # Loading
```

```
##
```

```
## Link          | Coefficient | SE |          95% CI |      z |      p | Group
```

```
## -----
```

```
## Approach =~ AWin_v_Neut_L_NAcc_run1 |          0.51 | 0.13 | [ 0.26, 0.76] |    4.03 | < .001 | 1.00
```

```
## Approach =~ AWin_v_Neut_R_NAcc_run1 |          0.51 | 0.10 | [ 0.31, 0.71] |    5.07 | < .001 | 1.00
```

```
## Approach =~ AWin_v_Neut_R_Insula_run1 |          0.19 | 0.13 | [-0.07, 0.45] |    1.43 | 0.154 | 1.00
```

```
## Approach =~ AWin_v_Neut_L_NAcc_run2 |          0.47 | 0.13 | [ 0.21, 0.72] |    3.63 | < .001 | 1.00
```

```
## Approach =~ AWin_v_Neut_R_NAcc_run2 |          0.38 | 0.12 | [ 0.15, 0.61] |    3.23 | 0.001 | 1.00
```

```
## Approach =~ AWin_v_Neut_R_Insula_run2 |          0.18 | 0.14 | [-0.09, 0.45] |    1.29 | 0.198 | 1.00
```

```
## Approach =~ BWin_v_Neut_L_NAcc_run1 |          0.33 | 0.11 | [ 0.12, 0.54] |    3.12 | 0.002 | 1.00
```

```
## Approach =~ BWin_v_Neut_R_NAcc_run1 |          0.30 | 0.15 | [ 0.01, 0.59] |    2.00 | 0.046 | 1.00
```

```
## Approach =~ BWin_v_Neut_R_Insula_run1 |          0.10 | 0.14 | [-0.16, 0.37] |    0.76 | 0.445 | 1.00
```

```
## Approach =~ BWin_v_Neut_L_NAcc_run2 |          0.34 | 0.10 | [ 0.14, 0.55] |    3.28 | 0.001 | 1.00
```

```
## Approach =~ BWin_v_Neut_R_NAcc_run2 |          0.45 | 0.11 | [ 0.25, 0.66] |    4.30 | < .001 | 1.00
```

```
## Approach =~ BWin_v_Neut_R_Insula_run2 |          0.26 | 0.11 | [ 0.04, 0.48] |    2.27 | 0.023 | 1.00
```

```
## Approach =~ BWin_v_BLose_L_NAcc_run1 |          0.34 | 0.14 | [ 0.07, 0.61] |    2.45 | 0.014 | 1.00
```

```
## Approach =~ BWin_v_BLose_R_NAcc_run1 |          0.48 | 0.10 | [ 0.29, 0.67] |    4.94 | < .001 | 1.00
```

```
## Approach =~ BWin_v_BLose_L_NAcc_run2 |          0.44 | 0.15 | [ 0.14, 0.73] |    2.88 | 0.004 | 1.00
```

```
## Approach =~ BWin_v_BLose_R_NAcc_run2 |          0.56 | 0.09 | [ 0.38, 0.75] |    6.06 | < .001 | 1.00
```

```
## Avoid =~ ALose_v_Neut_L_Insula_run1 |          0.52 | 0.10 | [ 0.32, 0.72] |    5.16 | < .001 | 1.00
```

```
## Avoid =~ ALose_v_Neut_R_Insula_run1 |          0.53 | 0.09 | [ 0.34, 0.71] |    5.62 | < .001 | 1.00
```

```
## Avoid =~ ALose_v_Neut_L_Insula_run2 |          0.55 | 0.11 | [ 0.33, 0.77] |    4.86 | < .001 | 1.00
```

```
## Avoid =~ ALose_v_Neut_R_Insula_run2 |          0.54 | 0.10 | [ 0.34, 0.73] |    5.33 | < .001 | 1.00
```

```
## Avoid =~ BLose_v_Neut_L_Insula_run1 |          0.50 | 0.10 | [ 0.30, 0.69] |    4.95 | < .001 | 1.00
```


## Avoid == Blose_v_Neut_R_Insula_run1		0.41		0.10		[0.21, 0.61]		3.99		< .001		1.00
## Avoid == Blose_v_Neut_L_Insula_run2		0.23		0.09		[0.05, 0.41]		2.54		0.011		1.00
## Avoid == Blose_v_Neut_R_Insula_run2		0.46		0.11		[0.24, 0.68]		4.11		< .001		1.00
## Avoid == Blose_v_BWin_L_Insula_run1		0.34		0.10		[0.15, 0.53]		3.52		< .001		1.00
## Avoid == Blose_v_BWin_R_Insula_run1		0.37		0.09		[0.19, 0.56]		3.98		< .001		1.00
## Avoid == Blose_v_BWin_L_Insula_run2		0.41		0.12		[0.17, 0.64]		3.38		< .001		1.00
## Avoid == Blose_v_BWin_R_Insula_run2		0.37		0.10		[0.17, 0.57]		3.61		< .001		1.00
## Approach == AWin_v_Neut_L_NAcc_run1		0.42		0.10		[0.22, 0.61]		4.23		< .001		2.00
## Approach == AWin_v_Neut_R_NAcc_run1		0.54		0.07		[0.40, 0.69]		7.43		< .001		2.00
## Approach == AWin_v_Neut_R_Insula_run1		0.15		0.08		[-0.01, 0.31]		1.89		0.059		2.00
## Approach == AWin_v_Neut_L_NAcc_run2		0.51		0.08		[0.36, 0.67]		6.55		< .001		2.00
## Approach == AWin_v_Neut_R_NAcc_run2		0.41		0.09		[0.23, 0.59]		4.52		< .001		2.00
## Approach == AWin_v_Neut_R_Insula_run2		2.85e-03		0.09		[-0.18, 0.18]		0.03		0.975		2.00
## Approach == BWin_v_Neut_L_NAcc_run1		0.40		0.11		[0.19, 0.61]		3.73		< .001		2.00
## Approach == BWin_v_Neut_R_NAcc_run1		0.32		0.10		[0.11, 0.52]		3.04		0.002		2.00
## Approach == BWin_v_Neut_R_Insula_run1		0.23		0.10		[0.03, 0.44]		2.23		0.026		2.00
## Approach == BWin_v_Neut_L_NAcc_run2		0.31		0.10		[0.12, 0.49]		3.23		0.001		2.00
## Approach == BWin_v_Neut_R_NAcc_run2		0.33		0.09		[0.16, 0.50]		3.80		< .001		2.00
## Approach == BWin_v_Neut_R_Insula_run2		0.27		0.10		[0.08, 0.47]		2.74		0.006		2.00
## Approach == BWin_v_Blose_L_NAcc_run1		0.37		0.10		[0.17, 0.56]		3.71		< .001		2.00
## Approach == BWin_v_Blose_R_NAcc_run1		0.51		0.08		[0.35, 0.67]		6.29		< .001		2.00
## Approach == BWin_v_Blose_L_NAcc_run2		0.44		0.09		[0.27, 0.61]		5.10		< .001		2.00
## Approach == BWin_v_Blose_R_NAcc_run2		0.36		0.09		[0.19, 0.54]		3.99		< .001		2.00
## Avoid == Alose_v_Neut_L_Insula_run1		0.51		0.08		[0.34, 0.67]		6.14		< .001		2.00
## Avoid == Alose_v_Neut_R_Insula_run1		0.37		0.09		[0.20, 0.53]		4.24		< .001		2.00
## Avoid == Alose_v_Neut_L_Insula_run2		0.39		0.09		[0.22, 0.56]		4.38		< .001		2.00
## Avoid == Alose_v_Neut_R_Insula_run2		0.51		0.07		[0.38, 0.64]		7.65		< .001		2.00
## Avoid == Blose_v_Neut_L_Insula_run1		0.35		0.08		[0.19, 0.51]		4.24		< .001		2.00
## Avoid == Blose_v_Neut_R_Insula_run1		0.50		0.07		[0.35, 0.64]		6.64		< .001		2.00
## Avoid == Blose_v_Neut_L_Insula_run2		0.44		0.07		[0.30, 0.58]		6.24		< .001		2.00
## Avoid == Blose_v_Neut_R_Insula_run2		0.47		0.09		[0.31, 0.64]		5.55		< .001		2.00
## Avoid == Blose_v_BWin_L_Insula_run1		0.61		0.08		[0.46, 0.76]		7.97		< .001		2.00
## Avoid == Blose_v_BWin_R_Insula_run1		0.48		0.07		[0.33, 0.62]		6.40		< .001		2.00
## Avoid == Blose_v_BWin_L_Insula_run2		0.63		0.06		[0.50, 0.75]		9.85		< .001		2.00
## Avoid == Blose_v_BWin_R_Insula_run2		0.44		0.08		[0.29, 0.59]		5.74		< .001		2.00
## Approach == AWin_v_Neut_L_NAcc_run1		0.45		0.03		[0.38, 0.51]		13.11		< .001		3.00
## Approach == AWin_v_Neut_R_NAcc_run1		0.43		0.03		[0.37, 0.50]		13.35		< .001		3.00
## Approach == AWin_v_Neut_R_Insula_run1		0.24		0.04		[0.17, 0.31]		6.69		< .001		3.00
## Approach == AWin_v_Neut_L_NAcc_run2		0.48		0.03		[0.42, 0.55]		14.87		< .001		3.00
## Approach == AWin_v_Neut_R_NAcc_run2		0.49		0.03		[0.44, 0.55]		17.00		< .001		3.00
## Approach == AWin_v_Neut_R_Insula_run2		0.27		0.03		[0.20, 0.34]		7.91		< .001		3.00
## Approach == BWin_v_Neut_L_NAcc_run1		0.51		0.03		[0.45, 0.56]		16.98		< .001		3.00
## Approach == BWin_v_Neut_R_NAcc_run1		0.39		0.03		[0.32, 0.45]		12.10		< .001		3.00
## Approach == BWin_v_Neut_R_Insula_run1		0.24		0.03		[0.17, 0.31]		6.99		< .001		3.00
## Approach == BWin_v_Neut_L_NAcc_run2		0.42		0.03		[0.36, 0.48]		13.20		< .001		3.00
## Approach == BWin_v_Neut_R_NAcc_run2		0.42		0.03		[0.36, 0.48]		13.55		< .001		3.00
## Approach == BWin_v_Neut_R_Insula_run2		0.27		0.04		[0.20, 0.35]		7.45		< .001		3.00
## Approach == BWin_v_Blose_L_NAcc_run1		0.47		0.03		[0.41, 0.53]		15.00		< .001		3.00
## Approach == BWin_v_Blose_R_NAcc_run1		0.46		0.03		[0.40, 0.53]		14.42		< .001		3.00
## Approach == BWin_v_Blose_L_NAcc_run2		0.42		0.03		[0.36, 0.48]		12.84		< .001		3.00
## Approach == BWin_v_Blose_R_NAcc_run2		0.48		0.03		[0.42, 0.55]		14.99		< .001		3.00
## Avoid == Alose_v_Neut_L_Insula_run1		0.44		0.03		[0.38, 0.51]		12.82		< .001		3.00
## Avoid == Alose_v_Neut_R_Insula_run1		0.50		0.03		[0.43, 0.56]		15.48		< .001		3.00
## Avoid == Alose_v_Neut_L_Insula_run2		0.41		0.03		[0.34, 0.47]		12.44		< .001		3.00

```

## Avoid =~ Alose_v_Neut_R_Insula_run2 | 0.46 | 0.03 | [ 0.39, 0.52] | 14.14 | < .001 | 3.00
## Avoid =~ Blose_v_Neut_L_Insula_run1 | 0.44 | 0.03 | [ 0.38, 0.51] | 12.90 | < .001 | 3.00
## Avoid =~ Blose_v_Neut_R_Insula_run1 | 0.51 | 0.03 | [ 0.44, 0.57] | 14.46 | < .001 | 3.00
## Avoid =~ Blose_v_Neut_L_Insula_run2 | 0.43 | 0.04 | [ 0.36, 0.50] | 11.76 | < .001 | 3.00
## Avoid =~ Blose_v_Neut_R_Insula_run2 | 0.48 | 0.03 | [ 0.42, 0.55] | 14.05 | < .001 | 3.00
## Avoid =~ Blose_v_BWin_L_Insula_run1 | 0.44 | 0.03 | [ 0.37, 0.51] | 12.62 | < .001 | 3.00
## Avoid =~ Blose_v_BWin_R_Insula_run1 | 0.23 | 0.04 | [ 0.16, 0.30] | 6.18 | < .001 | 3.00
## Avoid =~ Blose_v_BWin_L_Insula_run2 | 0.42 | 0.04 | [ 0.35, 0.49] | 11.89 | < .001 | 3.00
## Avoid =~ Blose_v_BWin_R_Insula_run2 | 0.23 | 0.04 | [ 0.16, 0.31] | 6.25 | < .001 | 3.00
##
## # Correlation
##
## Link | Coefficient | SE | 95% CI | z | p | Group
## -----
## Approach ~~ Avoid | -0.42 | 0.12 | [-0.65, -0.19] | -3.63 | < .001 | 1.00
## Approach ~~ Avoid | -0.54 | 0.09 | [-0.71, -0.36] | -6.01 | < .001 | 2.00
## Approach ~~ Avoid | -0.57 | 0.03 | [-0.63, -0.51] | -17.66 | < .001 | 3.00

```

Metric CFA model

```

##### Summarizing Metric Multi-group CFA model #####
parameters(metric_cfa, standardize = T)

```

```

## # Loading
##
## Link | Coefficient | SE | 95% CI | z | p |
## -----
## Approach =~ AWin_v_Neut_L_NAcc_run1 (.p1.) | 0.43 | 0.05 | [0.33, 0.53] | 8.39 | < .001 |
## Approach =~ AWin_v_Neut_R_NAcc_run1 (.p2.) | 0.43 | 0.06 | [0.32, 0.54] | 7.56 | < .001 |
## Approach =~ AWin_v_Neut_R_Insula_run1 (.p3.) | 0.20 | 0.04 | [0.13, 0.27] | 5.63 | < .001 |
## Approach =~ AWin_v_Neut_L_NAcc_run2 (.p4.) | 0.46 | 0.05 | [0.36, 0.56] | 9.09 | < .001 |
## Approach =~ AWin_v_Neut_R_NAcc_run2 (.p5.) | 0.46 | 0.05 | [0.36, 0.55] | 9.65 | < .001 |
## Approach =~ AWin_v_Neut_R_Insula_run2 (.p6.) | 0.23 | 0.04 | [0.14, 0.31] | 5.41 | < .001 |
## Approach =~ BWin_v_Neut_L_NAcc_run1 (.p7.) | 0.43 | 0.05 | [0.34, 0.53] | 8.94 | < .001 |
## Approach =~ BWin_v_Neut_R_NAcc_run1 (.p8.) | 0.36 | 0.05 | [0.26, 0.45] | 7.54 | < .001 |
## Approach =~ BWin_v_Neut_R_Insula_run1 (.p9.) | 0.22 | 0.04 | [0.15, 0.30] | 5.99 | < .001 |
## Approach =~ BWin_v_Neut_L_NAcc_run2 (.p10.) | 0.39 | 0.05 | [0.30, 0.48] | 8.24 | < .001 |
## Approach =~ BWin_v_Neut_R_NAcc_run2 (.p11.) | 0.38 | 0.05 | [0.28, 0.47] | 7.77 | < .001 |
## Approach =~ BWin_v_Neut_R_Insula_run2 (.p12.) | 0.29 | 0.04 | [0.20, 0.38] | 6.53 | < .001 |
## Approach =~ BWin_v_BLose_L_NAcc_run1 (.p13.) | 0.40 | 0.05 | [0.31, 0.49] | 8.48 | < .001 |
## Approach =~ BWin_v_BLose_R_NAcc_run1 (.p14.) | 0.45 | 0.05 | [0.36, 0.55] | 9.18 | < .001 |
## Approach =~ BWin_v_BLose_L_NAcc_run2 (.p15.) | 0.36 | 0.05 | [0.27, 0.45] | 7.80 | < .001 |
## Approach =~ BWin_v_BLose_R_NAcc_run2 (.p16.) | 0.50 | 0.06 | [0.38, 0.61] | 8.40 | < .001 |
## Avoid =~ Alose_v_Neut_L_Insula_run1 (.p17.) | 0.47 | 0.05 | [0.37, 0.57] | 9.59 | < .001 |
## Avoid =~ Alose_v_Neut_R_Insula_run1 (.p18.) | 0.48 | 0.05 | [0.38, 0.58] | 9.33 | < .001 |
## Avoid =~ Alose_v_Neut_L_Insula_run2 (.p19.) | 0.41 | 0.05 | [0.32, 0.51] | 8.48 | < .001 |
## Avoid =~ Alose_v_Neut_R_Insula_run2 (.p20.) | 0.44 | 0.05 | [0.34, 0.54] | 8.57 | < .001 |
## Avoid =~ Blose_v_Neut_L_Insula_run1 (.p21.) | 0.46 | 0.05 | [0.36, 0.55] | 9.63 | < .001 |
## Avoid =~ Blose_v_Neut_R_Insula_run1 (.p22.) | 0.48 | 0.04 | [0.39, 0.57] | 10.63 | < .001 |
## Avoid =~ Blose_v_Neut_L_Insula_run2 (.p23.) | 0.40 | 0.04 | [0.32, 0.49] | 9.31 | < .001 |
## Avoid =~ Blose_v_Neut_R_Insula_run2 (.p24.) | 0.51 | 0.05 | [0.41, 0.61] | 10.19 | < .001 |
## Avoid =~ Blose_v_BWin_L_Insula_run1 (.p25.) | 0.48 | 0.04 | [0.39, 0.56] | 11.05 | < .001 |
## Avoid =~ Blose_v_BWin_R_Insula_run1 (.p26.) | 0.30 | 0.04 | [0.21, 0.39] | 6.78 | < .001 |
## Avoid =~ Blose_v_BWin_L_Insula_run2 (.p27.) | 0.47 | 0.05 | [0.37, 0.56] | 9.97 | < .001 |

```

## Avoid == Blose_v_BWin_R_Insula_run2 (.p28.)		0.28		0.04		[0.19, 0.36]		6.44		< .001	
## Approach == AWin_v_Neut_L_NAcc_run1 (.p1.)		0.42		0.04		[0.33, 0.50]		9.65		< .001	
## Approach == AWin_v_Neut_R_NAcc_run1 (.p2.)		0.44		0.05		[0.34, 0.53]		9.03		< .001	
## Approach == AWin_v_Neut_R_Insula_run1 (.p3.)		0.22		0.03		[0.15, 0.29]		6.46		< .001	
## Approach == AWin_v_Neut_L_NAcc_run2 (.p4.)		0.44		0.04		[0.35, 0.52]		10.49		< .001	
## Approach == AWin_v_Neut_R_NAcc_run2 (.p5.)		0.44		0.04		[0.37, 0.52]		10.97		< .001	
## Approach == AWin_v_Neut_R_Insula_run2 (.p6.)		0.22		0.03		[0.16, 0.28]		7.38		< .001	
## Approach == BWin_v_Neut_L_NAcc_run1 (.p7.)		0.42		0.04		[0.34, 0.51]		10.00		< .001	
## Approach == BWin_v_Neut_R_NAcc_run1 (.p8.)		0.32		0.04		[0.25, 0.40]		8.44		< .001	
## Approach == BWin_v_Neut_R_Insula_run1 (.p9.)		0.20		0.03		[0.14, 0.27]		6.20		< .001	
## Approach == BWin_v_Neut_L_NAcc_run2 (.p10.)		0.38		0.04		[0.30, 0.45]		9.46		< .001	
## Approach == BWin_v_Neut_R_NAcc_run2 (.p11.)		0.36		0.04		[0.28, 0.44]		9.06		< .001	
## Approach == BWin_v_Neut_R_Insula_run2 (.p12.)		0.24		0.04		[0.17, 0.32]		6.51		< .001	
## Approach == BWin_v_Blose_L_NAcc_run1 (.p13.)		0.41		0.04		[0.33, 0.48]		10.32		< .001	
## Approach == BWin_v_Blose_R_NAcc_run1 (.p14.)		0.45		0.04		[0.36, 0.54]		10.14		< .001	
## Approach == BWin_v_Blose_L_NAcc_run2 (.p15.)		0.37		0.04		[0.29, 0.45]		9.01		< .001	
## Approach == BWin_v_Blose_R_NAcc_run2 (.p16.)		0.41		0.04		[0.33, 0.49]		10.34		< .001	
## Avoid == Alose_v_Neut_L_Insula_run1 (.p17.)		0.48		0.05		[0.39, 0.57]		10.43		< .001	
## Avoid == Alose_v_Neut_R_Insula_run1 (.p18.)		0.50		0.04		[0.43, 0.58]		13.51		< .001	
## Avoid == Alose_v_Neut_L_Insula_run2 (.p19.)		0.44		0.04		[0.36, 0.53]		10.74		< .001	
## Avoid == Alose_v_Neut_R_Insula_run2 (.p20.)		0.53		0.04		[0.45, 0.61]		12.91		< .001	
## Avoid == Blose_v_Neut_L_Insula_run1 (.p21.)		0.45		0.04		[0.37, 0.53]		11.26		< .001	
## Avoid == Blose_v_Neut_R_Insula_run1 (.p22.)		0.56		0.04		[0.48, 0.64]		13.66		< .001	
## Avoid == Blose_v_Neut_L_Insula_run2 (.p23.)		0.46		0.04		[0.38, 0.55]		10.64		< .001	
## Avoid == Blose_v_Neut_R_Insula_run2 (.p24.)		0.52		0.04		[0.43, 0.60]		12.04		< .001	
## Avoid == Blose_v_BWin_L_Insula_run1 (.p25.)		0.51		0.05		[0.42, 0.60]		10.92		< .001	
## Avoid == Blose_v_BWin_R_Insula_run1 (.p26.)		0.33		0.05		[0.23, 0.43]		6.59		< .001	
## Avoid == Blose_v_BWin_L_Insula_run2 (.p27.)		0.50		0.05		[0.40, 0.59]		10.01		< .001	
## Avoid == Blose_v_BWin_R_Insula_run2 (.p28.)		0.28		0.04		[0.20, 0.36]		6.93		< .001	
## Approach == AWin_v_Neut_L_NAcc_run1 (.p1.)		0.45		0.03		[0.39, 0.51]		14.42		< .001	
## Approach == AWin_v_Neut_R_NAcc_run1 (.p2.)		0.46		0.03		[0.40, 0.51]		15.97		< .001	
## Approach == AWin_v_Neut_R_Insula_run1 (.p3.)		0.23		0.03		[0.16, 0.29]		7.03		< .001	
## Approach == AWin_v_Neut_L_NAcc_run2 (.p4.)		0.49		0.03		[0.43, 0.55]		16.52		< .001	
## Approach == AWin_v_Neut_R_NAcc_run2 (.p5.)		0.48		0.03		[0.43, 0.54]		17.30		< .001	
## Approach == AWin_v_Neut_R_Insula_run2 (.p6.)		0.24		0.03		[0.17, 0.30]		7.16		< .001	
## Approach == BWin_v_Neut_L_NAcc_run1 (.p7.)		0.50		0.03		[0.44, 0.55]		17.20		< .001	
## Approach == BWin_v_Neut_R_NAcc_run1 (.p8.)		0.38		0.03		[0.32, 0.44]		12.64		< .001	
## Approach == BWin_v_Neut_R_Insula_run1 (.p9.)		0.24		0.03		[0.17, 0.30]		7.28		< .001	
## Approach == BWin_v_Neut_L_NAcc_run2 (.p10.)		0.41		0.03		[0.35, 0.47]		13.61		< .001	
## Approach == BWin_v_Neut_R_NAcc_run2 (.p11.)		0.42		0.03		[0.37, 0.48]		14.80		< .001	
## Approach == BWin_v_Neut_R_Insula_run2 (.p12.)		0.27		0.03		[0.21, 0.34]		8.27		< .001	
## Approach == BWin_v_Blose_L_NAcc_run1 (.p13.)		0.46		0.03		[0.40, 0.52]		15.10		< .001	
## Approach == BWin_v_Blose_R_NAcc_run1 (.p14.)		0.47		0.03		[0.42, 0.53]		16.39		< .001	
## Approach == BWin_v_Blose_L_NAcc_run2 (.p15.)		0.43		0.03		[0.37, 0.49]		14.26		< .001	
## Approach == BWin_v_Blose_R_NAcc_run2 (.p16.)		0.49		0.03		[0.43, 0.54]		16.43		< .001	
## Avoid == Alose_v_Neut_L_Insula_run1 (.p17.)		0.46		0.03		[0.40, 0.52]		15.12		< .001	
## Avoid == Alose_v_Neut_R_Insula_run1 (.p18.)		0.47		0.03		[0.41, 0.53]		15.57		< .001	
## Avoid == Alose_v_Neut_L_Insula_run2 (.p19.)		0.42		0.03		[0.36, 0.48]		14.03		< .001	
## Avoid == Alose_v_Neut_R_Insula_run2 (.p20.)		0.46		0.03		[0.40, 0.51]		16.08		< .001	
## Avoid == Blose_v_Neut_L_Insula_run1 (.p21.)		0.43		0.03		[0.37, 0.49]		13.74		< .001	
## Avoid == Blose_v_Neut_R_Insula_run1 (.p22.)		0.49		0.03		[0.42, 0.55]		15.06		< .001	
## Avoid == Blose_v_Neut_L_Insula_run2 (.p23.)		0.40		0.03		[0.34, 0.47]		12.48		< .001	
## Avoid == Blose_v_Neut_R_Insula_run2 (.p24.)		0.47		0.03		[0.41, 0.53]		15.11		< .001	
## Avoid == Blose_v_BWin_L_Insula_run1 (.p25.)		0.44		0.03		[0.38, 0.51]		14.04		< .001	

```
## Avoid =~ Blose_v_BWin_R_Insula_run1 (.p26.) | 0.27 | 0.03 | [0.21, 0.33] | 8.91 | < .001 |
## Avoid =~ Blose_v_BWin_L_Insula_run2 (.p27.) | 0.44 | 0.03 | [0.38, 0.50] | 14.73 | < .001 |
## Avoid =~ Blose_v_BWin_R_Insula_run2 (.p28.) | 0.27 | 0.03 | [0.21, 0.33] | 8.55 | < .001 |
##
## # Correlation
##
## Link | Coefficient | SE | 95% CI | z | p | Group
## -----
## Approach ~~ Avoid | -0.40 | 0.11 | [-0.62, -0.18] | -3.61 | < .001 | 1.00
## Approach ~~ Avoid | -0.53 | 0.09 | [-0.71, -0.36] | -6.03 | < .001 | 2.00
## Approach ~~ Avoid | -0.57 | 0.03 | [-0.64, -0.51] | -17.75 | < .001 | 3.00
```

Comparing models w/ BIC/AIC (anova)

The below compares whether the complete data (across all three samples) in the `all_cfa` model is significantly improved by the configural invariance model. A significant value indicates that the configural model is significantly better than the full sample cfa.

```
anova(all_sample, configural_cfa)

## 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)
## all_sample    349 -21672 -21235 2200.6
## configural_cfa 1047 -21482 -20170 2961.4      777.17    698    0.01964 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Next, anova is used to compare the model improve in AIC/BIC by between the configural and metric invariance. A significantly result in the anova would indicate a significant improvement of the metric model over the configural model.

```
anova(configural_cfa, metric_cfa)

## 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)
## configural_cfa 1047 -21482 -20170 2961.4
## metric_cfa    1099 -21523 -20479 3024.3      53.862    52    0.403
```

Plotting multi-group config. CFA

Use `semPaths` to plot the configural invariance CFA multigroup model

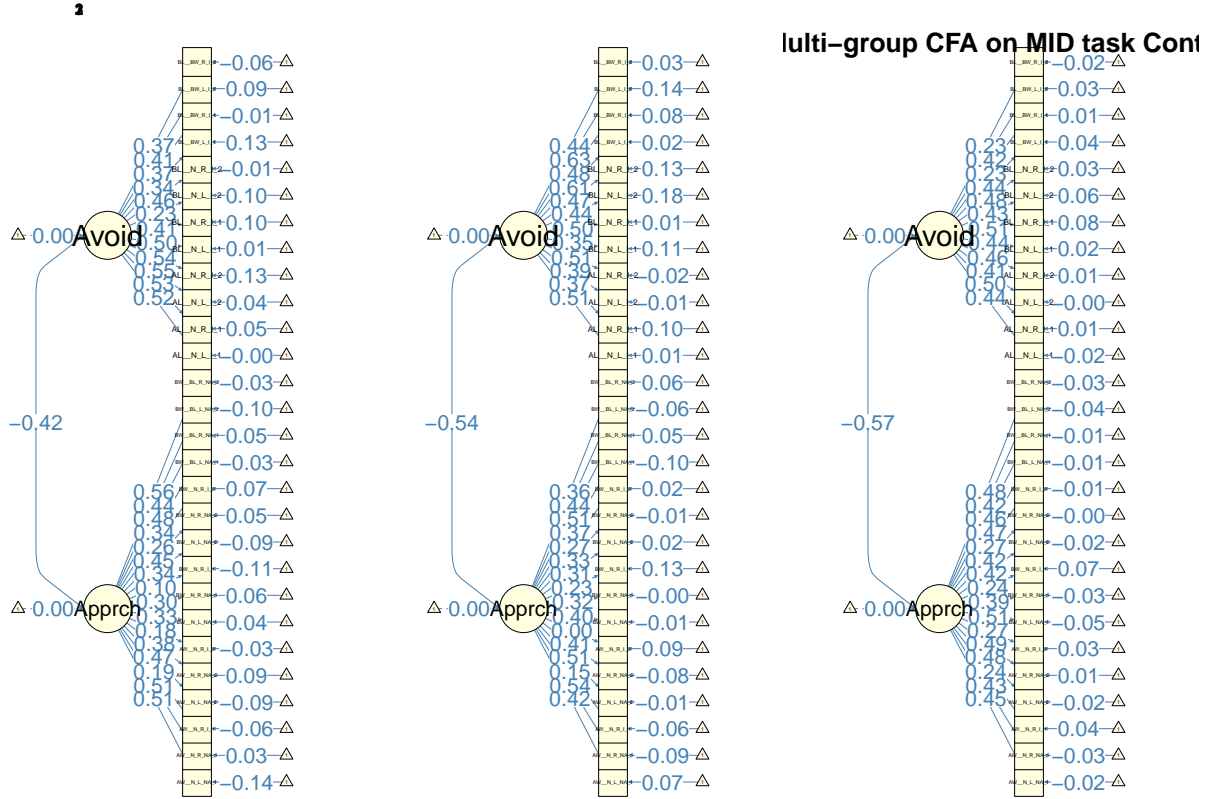
```
# this plotting is not function with runs loading onto ROIs

layout(t(1:3))
```

```

semPaths(configural_cfa,
  color = "lightyellow",
  theme="colorblind",
  whatLabels = "std",
  style = "lisrel",
  sizeLat = 10,
  sizeLat2 = 10,
  sizeMan = 6,
  edge.color = "steelblue",
  edge.label.cex = 2,
  label.cex = 2,
  rotation = 2,
  layout = "tree2",
  intercepts = TRUE,
  residuals = FALSE,
  #residScale = 10,
  curve = 2,
  title = T,
  title.color = "black",
  cardinal = "lat cov",
  curvePivot = T,
  nCharNodes = 6,
  #nodeLabels = label,
  mar = c(2,5,2,6))
# Title
title("Multi-group CFA on MID task Contrasts")

```



Running [semi-restricted] ESEM Model

As described in the manuscript, the restricted CFA may incorrectly account for some measurement error in the items. This may degrade the fit statistics. See Marsh et al. (2014) for an in-depth discussion.

In this case, Exploratory Structural Equation Modeling (ESEM) is used to fit a CFA pre-specified model that allows for non-zero loadings. The technique and application of ESEM is available through the `psych` and `esemcomp` packages. Here, the `esemcomp` package is used to fit a model using the steps described by Mateus Silvestrin here and by Guàrdia-Olmos et al.. The github code for `esemcomp` is available here. Below can be used to download the `esemComp` package – which worked with R version 4.2.1 on x86_64-apple-darwin17.0 during September 2022.

```
devtools::install_github("MateusPsi/esemComp", build_vignettes = TRUE)
```

Selected items for ESEM

First, select the items that are consistent with those in the CFA model

```
# ordering so can specify numerically
esem_data = brain_set[,c("AWin_v_Neut_L_NAcc_run1" , "AWin_v_Neut_L_NAcc_run2" ,
  "BWin_v_Neut_L_NAcc_run1" , "BWin_v_Neut_L_NAcc_run2" ,
  "BWin_v_BLose_L_NAcc_run1" , "BWin_v_BLose_L_NAcc_run2",
  "AWin_v_Neut_R_NAcc_run1" , "AWin_v_Neut_R_NAcc_run2",
  "BWin_v_Neut_R_NAcc_run1" , "BWin_v_Neut_R_NAcc_run2",
  "BWin_v_BLose_R_NAcc_run1" , "BWin_v_BLose_R_NAcc_run2",
  # insula values approach
  "AWin_v_Neut_R_Insula_run1" , "AWin_v_Neut_R_Insula_run2",
```

```

"BWin_v_Neut_R_Insula_run1", "BWin_v_Neut_R_Insula_run2",
# avoidance
"Alose_v_Neut_L_Insula_run1", "Alose_v_Neut_L_Insula_run2",
"Blose_v_Neut_L_Insula_run1", "Blose_v_Neut_L_Insula_run2",
"Blose_v_BWin_L_Insula_run1", "Blose_v_BWin_L_Insula_run2",
"Alose_v_Neut_R_Insula_run1", "Alose_v_Neut_R_Insula_run2",
"Blose_v_Neut_R_Insula_run1", "Blose_v_Neut_R_Insula_run2",
"Blose_v_BWin_R_Insula_run1", "Blose_v_BWin_R_Insula_run2",
"set")]
```

Specify EFA Model

As described in March et al. (2014), create a target rotation for items onto factors. In this case two factors are specified by the CFA model, so factor 1 and factor 2 are specified in `make_target`.

```

target_rot <- make_target(28, mainloadings = list(f1 = 1:16, f2 = 17:28))
esem.efa <- esem_efa(data = esem_data[,1:28], nfactors = 2,
  target = target_rot, fm = "ml")
```

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

```
esem.efa$loadings
```

```

##
## Loadings:
##
##              ML1      ML2
## AWin_v_Neut_L_NAcc_run1    0.421
## AWin_v_Neut_L_NAcc_run2    0.459
## BWin_v_Neut_L_NAcc_run1    0.455
## BWin_v_Neut_L_NAcc_run2    0.414
## BWin_v_BLose_L_NAcc_run1    0.398
## BWin_v_BLose_L_NAcc_run2    0.364
## AWin_v_Neut_R_NAcc_run1    0.516
## AWin_v_Neut_R_NAcc_run2    0.564  0.112
## BWin_v_Neut_R_NAcc_run1    0.391
## BWin_v_Neut_R_NAcc_run2    0.419
## BWin_v_BLose_R_NAcc_run1    0.450
## BWin_v_BLose_R_NAcc_run2    0.461
## AWin_v_Neut_R_Insula_run1    0.204
## AWin_v_Neut_R_Insula_run2    0.226
## BWin_v_Neut_R_Insula_run1    0.245
## BWin_v_Neut_R_Insula_run2    0.255
## Alose_v_Neut_L_Insula_run1          0.470
## Alose_v_Neut_L_Insula_run2          0.435
## Blose_v_Neut_L_Insula_run1          0.450
## Blose_v_Neut_L_Insula_run2          0.431
## Blose_v_BWin_L_Insula_run1          0.462
## Blose_v_BWin_L_Insula_run2          0.451
## Alose_v_Neut_R_Insula_run1          0.466
## Alose_v_Neut_R_Insula_run2          0.427
## Blose_v_Neut_R_Insula_run1          0.502
## Blose_v_Neut_R_Insula_run2          0.489
## Blose_v_BWin_R_Insula_run1          0.228
## Blose_v_BWin_R_Insula_run2          0.215
##
```

```
##           ML1    ML2
## SS loadings  2.616 2.243
## Proportion Var 0.093 0.080
## Cumulative Var 0.093 0.174
```

Using item that loads highest on factor 1 and lowest on factor 2 and vice versa, and define as anchor using `find_referents`

```
# per the example from Mateus Silverstrin, need to define anchor for each factor (value to loads higher)
anchor <- find_referents(efa_object = esem.efa, factor_names = c("f1", "f2"))
```

Once the esem efa and anchors are defined, use `syntax_composer` to specied the esem model. This will produce a lavaan specified model that references starting values that will be used in the cfa model

```
# Pull starting parameters
esem_mid_model <- syntax_composer(efa_object = esem.efa, referents = anchor)
```

Run ESEM model

Specified Model

The starting values are printed below to provide reference for how starting values differ from a strict CFA model. Notice, how some values that were original not fit onto the Approach factor (f1), such as big lose contrasts, they are now specified with loading values that are between .05 to -.05.

```
cat(esem_mid_model)

## f1 =~ start(0.421)*AWin_v_Neut_L_NAcc_run1+
## start(0.459)*AWin_v_Neut_L_NAcc_run2+
## start(0.455)*BWin_v_Neut_L_NAcc_run1+
## start(0.414)*BWin_v_Neut_L_NAcc_run2+
## start(0.398)*BWin_v_BLose_L_NAcc_run1+
## start(0.364)*BWin_v_BLose_L_NAcc_run2+
## start(0.516)*AWin_v_Neut_R_NAcc_run1+
## start(0.564)*AWin_v_Neut_R_NAcc_run2+
## start(0.391)*BWin_v_Neut_R_NAcc_run1+
## start(0.419)*BWin_v_Neut_R_NAcc_run2+
## start(0.45)*BWin_v_BLose_R_NAcc_run1+
## start(0.461)*BWin_v_BLose_R_NAcc_run2+
## start(0.204)*AWin_v_Neut_R_Insula_run1+
## start(0.226)*AWin_v_Neut_R_Insula_run2+
## start(0.245)*BWin_v_Neut_R_Insula_run1+
## start(0.255)*BWin_v_Neut_R_Insula_run2+
## start(0.012)*ALose_v_Neut_L_Insula_run1+
## start(0.016)*ALose_v_Neut_L_Insula_run2+
## start(0.012)*BLose_v_Neut_L_Insula_run1+
## start(0.025)*BLose_v_Neut_L_Insula_run2+
## start(0.01)*BLose_v_BWin_L_Insula_run1+
## start(0)*BLose_v_BWin_L_Insula_run2+
## start(-0.019)*ALose_v_Neut_R_Insula_run1+
## start(-0.054)*ALose_v_Neut_R_Insula_run2+
## 0.004*BLose_v_Neut_R_Insula_run1+
## start(0.009)*BLose_v_Neut_R_Insula_run2+
## start(-0.058)*BLose_v_BWin_R_Insula_run1+
## start(-0.073)*BLose_v_BWin_R_Insula_run2
##
## f2 =~ start(-0.034)*AWin_v_Neut_L_NAcc_run1+
```



```
## start(-0.034)*AWin_v_Neut_L_NAcc_run2+
## start(-0.037)*BWin_v_Neut_L_NAcc_run1+
## start(0.019)*BWin_v_Neut_L_NAcc_run2+
## start(-0.063)*BWin_v_BLose_L_NAcc_run1+
## start(-0.084)*BWin_v_BLose_L_NAcc_run2+
## start(0.079)*AWin_v_Neut_R_NAcc_run1+
## 0.112*AWin_v_Neut_R_NAcc_run2+
## start(0.03)*BWin_v_Neut_R_NAcc_run1+
## start(0.011)*BWin_v_Neut_R_NAcc_run2+
## start(-0.029)*BWin_v_BLose_R_NAcc_run1+
## start(-0.02)*BWin_v_BLose_R_NAcc_run2+
## start(-0.03)*AWin_v_Neut_R_Insula_run1+
## start(-0.012)*AWin_v_Neut_R_Insula_run2+
## start(0.021)*BWin_v_Neut_R_Insula_run1+
## start(-0.026)*BWin_v_Neut_R_Insula_run2+
## start(0.47)*ALose_v_Neut_L_Insula_run1+
## start(0.435)*ALose_v_Neut_L_Insula_run2+
## start(0.45)*BLose_v_Neut_L_Insula_run1+
## start(0.431)*BLose_v_Neut_L_Insula_run2+
## start(0.462)*BLose_v_BWin_L_Insula_run1+
## start(0.451)*BLose_v_BWin_L_Insula_run2+
## start(0.466)*ALose_v_Neut_R_Insula_run1+
## start(0.427)*ALose_v_Neut_R_Insula_run2+
## start(0.502)*BLose_v_Neut_R_Insula_run1+
## start(0.489)*BLose_v_Neut_R_Insula_run2+
## start(0.228)*BLose_v_BWin_R_Insula_run1+
## start(0.215)*BLose_v_BWin_R_Insula_run2
```

Running full ESEM model

After the EFA loadings are extracted using a target rotation, starting values are now available. These are now used to specify a less restrictive CFA model

```
esem_mid_fit<- cfa(esem_mid_model, esim_data[,1:28], std.lv=TRUE, meanstructure = TRUE,
  estimator = "MLR")
```

Pull and add fit statistics to the out dataframe and print results to see decreases in AIC/BIC

```
# adding values to the CFA model fit indices
out[4,2:8] <- round(data.matrix(fitmeasures(esem_mid_fit,
  fit.measures = c("chisq","df","pvalue",
    "rmsea", "cfi","tli", "srmr"))),
  digits=3)
out[4,9] <- round(AIC(esem_mid_fit),3)
out[4,10] <- round(BIC(esem_mid_fit),3)
out[4,1] <- c("Overall ESEM")

out <- as.data.frame(out)

out %>%
  knitr::kable(
    col.names = c("Model", "Chi-sq", "DF", "p value", "RMSEA", "CFI", "TLI","SRMR", "AIC", "BIC"),
    caption = "Fit statistics from MG-CFA and ESEM models",
    booktabs = TRUE
  )
```

Table 1: Fit statistics from MG-CFA and ESEM models

Model	Chi-sq	DF	p value	RMSEA	CFI	TLI	SRMR	AIC	BIC
Overall CFA	2200.622	349	0	0.065	0.657	0.628	0.053	-21672.085	-21234.81
Configg MG-CFA	2961.409	1047	0	0.066	0.65	0.621	0.06	-21482.229	-20170.405
Metric MG-CFA	3024.264	1099	0	0.064	0.648	0.637	0.062	-21523.374	-20479.059
Overall ESEM	2174.511	323	0	0.067	0.657	0.598	0.052	-21646.197	-21075.167

Aim: 2. Running EFA [Unrestricted] model

Here, a data-driven exploratory factor analysis is performed as implemented using the (<https://www.rdocumentation.org/packages/stats/versions/3.6.2/topics/factanal>) [<https://www.rdocumentation.org/packages/stats/versions/3.6.2/topics/factanal>] in the stats package. The same variables as in the CFA and ESEM dataset are used. A tutorial from Nilam Ram on EFA is also available [here](#)

By Sample EFA

Used the ([factanal](https://www.rdocumentation.org/packages/stats/versions/3.6.2/topics/factanal)) [<https://www.rdocumentation.org/packages/stats/versions/3.6.2/topics/factanal>] to run EFA model. Specifying the number of factors and using the **promax** (non-orthogonal) rotation.

```
abcd_efa_df <- esem_data %>% filter(set == 1)
mls_efa_df <- esem_data %>% filter(set == 2)
ahrb_efa_df <- esem_data %>% filter(set == 3)
```

Sample: *ABCD*

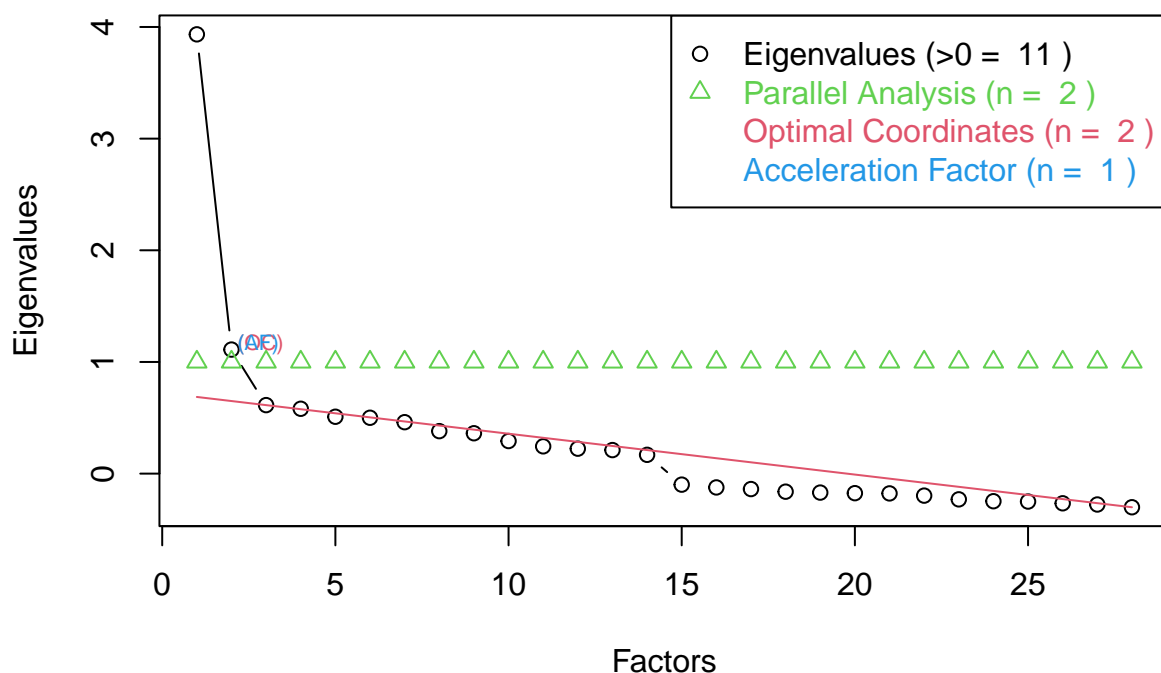
A number of methods can be used to estimate the recommended data driven factors in the data. There are nuanced differences in calculations between packages and methods. Therefore, a number of packages are used to acquire consistent evidence to acquire the most parsimonious model.

Using nFactors package see the recommended factors for the EFA model using a number of models.

```
fa_abcd <- subset(abcd_efa_df[,1:28])

plot(nScree(x=fa_abcd,model="factors"))
```

Non Graphical Solutions to Scree Test



To avoid biasing of packages in different calculations of recommendation factors that depend on strong correlations between bilateral regions by using parallel analysis. Parallel analysis is also implemented using the paran package.

```
paran(x = fa_abcd,
      iterations = 1000, quietly = FALSE, centile = 95,
      status = FALSE, all = TRUE, cfa = TRUE, graph = TRUE, color = TRUE,
      col = c("black", "red", "blue"), lty = c(1, 2, 3), lwd = 1, legend = TRUE,
      seed = 100)
```

```
##
## Using eigendecomposition of correlation matrix.
##
## Results of Horn's Parallel Analysis for factor retention
## 1000 iterations, using the 95 centile estimate
##
## -----
## Factor      Adjusted      Unadjusted      Estimated
##             Eigenvalue    Eigenvalue      Bias
## -----
## No components passed.
## -----
## 1           3.542178      3.934015        0.391836
## 2           0.775202      1.111071        0.335869
## 3           0.313670      0.613506        0.299835
## 4           0.313864      0.580679        0.266815
## 5           0.271159      0.509467        0.238307
```

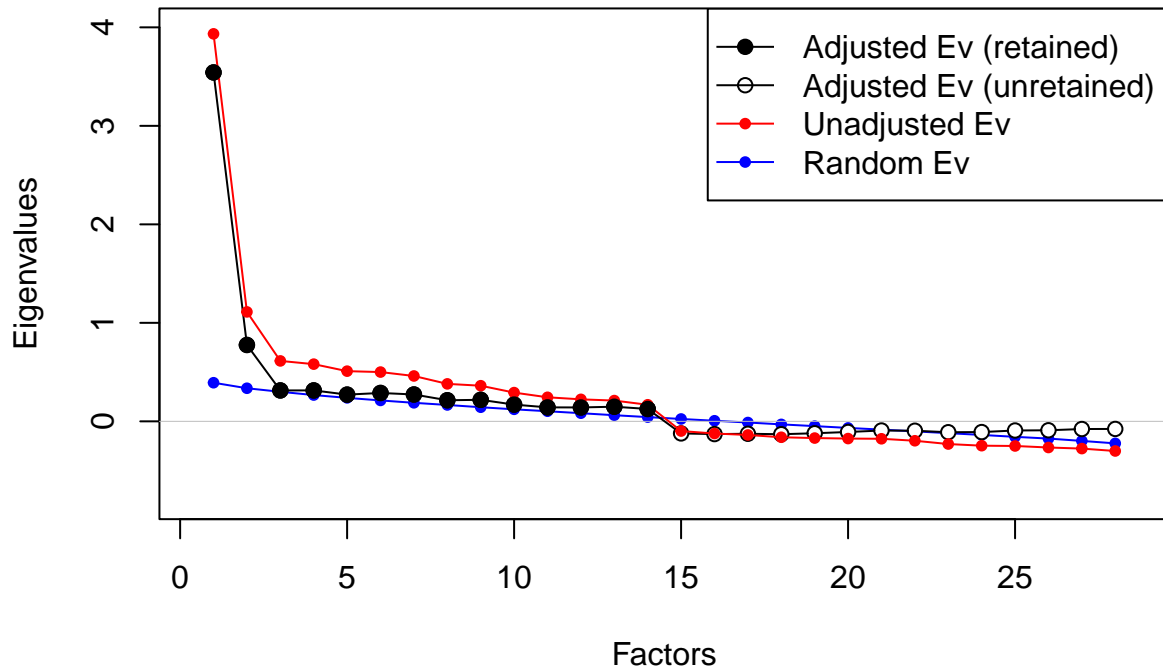
## 6	0.287997	0.500248	0.212250
## 7	0.272560	0.460549	0.187988
## 8	0.214122	0.380351	0.166228
## 9	0.218417	0.361595	0.143178
## 10	0.169955	0.291872	0.121916
## 11	0.141768	0.244225	0.102457
## 12	0.140901	0.223414	0.082512
## 13	0.148606	0.210788	0.062181
## 14	0.125450	0.168313	0.042863
## 15	-0.122634	-0.09868	0.023946
## 16	-0.129530	-0.12376	0.005761
## 17	-0.126362	-0.13831	-0.01194
## 18	-0.130790	-0.16166	-0.03087
## 19	-0.121125	-0.16974	-0.04861
## 20	-0.108212	-0.17507	-0.06685
## 21	-0.093136	-0.17720	-0.08406
## 22	-0.095347	-0.19714	-0.10179
## 23	-0.110512	-0.23053	-0.12002
## 24	-0.109017	-0.24716	-0.13814
## 25	-0.092890	-0.25014	-0.15725
## 26	-0.090723	-0.26576	-0.17504
## 27	-0.077950	-0.27668	-0.19873
## 28	-0.077263	-0.30109	-0.22383

##

Adjusted eigenvalues > 0 indicate dimensions to retain.

(14 factors retained)

Parallel Analysis



Comparing the above with the BIC comparison of an EFA model to determine the best fitting model based on fit statistics. Factor Analysis is submitted across a range of factors, e.g., 1-5, and the BIC is extracted from the model to determine the optimal number of factors

```
abcd_rec_factors <- matrix(NA, ncol = 2, nrow = 20)
colnames(abcd_rec_factors) <- c("Nfactors", "BIC")

for (f in 1:20) {
  test_fac <- fa(r = fa_abcd, #raw data
                nfactors = f, fm = 'minres',
                rotate = "oblimin")
  abcd_rec_factors[f,1] <- f
  abcd_rec_factors[f,2] <- test_fac$BIC
}
```

```
## Warning in GPFoblq(L, Tmat = Tmat, normalize = normalize, eps = eps, maxit =
## maxit, : convergence not obtained in GPFoblq. 1000 iterations used.
```

```
## Warning in GPFoblq(L, Tmat = Tmat, normalize = normalize, eps = eps, maxit =
## maxit, : convergence not obtained in GPFoblq. 1000 iterations used.
```

```
## Warning in GPFoblq(L, Tmat = Tmat, normalize = normalize, eps = eps, maxit =
## maxit, : convergence not obtained in GPFoblq. 1000 iterations used.
```

```
## Warning in GPFoblq(L, Tmat = Tmat, normalize = normalize, eps = eps, maxit =
## maxit, : convergence not obtained in GPFoblq. 1000 iterations used.
```

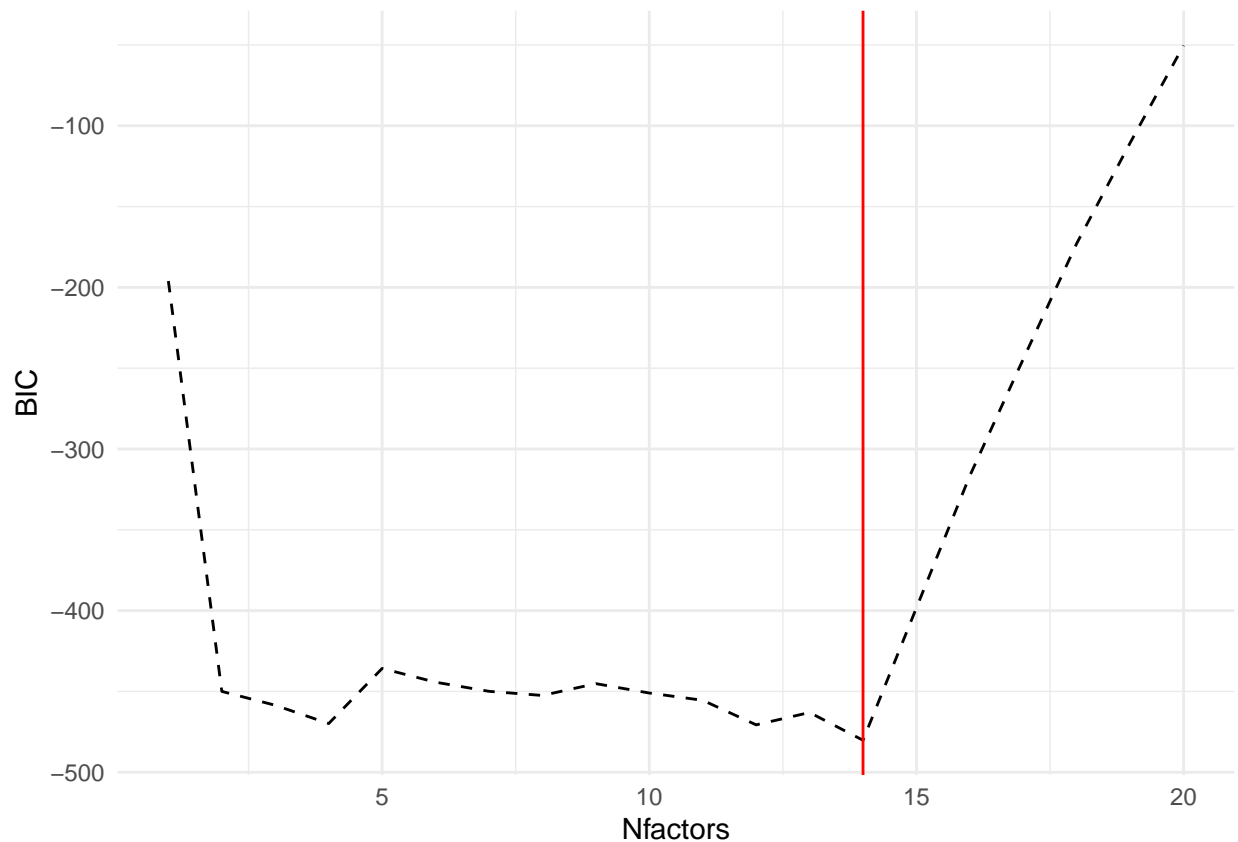
```

abcd_bic_fact = as.data.frame(abcd_rec_factors)

abcd_lowest_bic <- which.min(abcd_bic_fact$BIC)

abcd_bic_fact %>%
  ggplot(aes(x = Nfactors, y = BIC)) +
  geom_line(colour = 'black', linetype = 'dashed') +
  geom_vline(xintercept = abcd_bic_fact$Nfactors[abcd_lowest_bic], colour = 'red') +
  theme_minimal()

```



Running EFA

```

abcd_efa <- factanal(x = fa_abcd, #raw data
  factors = 2, fm = 'minres', rotation = "promax" # oblique rotation allow for non-orthogona
)

heatmaply(round(abcd_efa$loadings[,1:2],2) %>% print(sort = T),
  scale_fill_gradient_fun = ggplot2::scale_fill_gradient2(
    low = "blue",
    high = "darkred",
    space = "Lab",
    midpoint = 0,
    limits = c(-1, 1)
  ),
  dendrogram = "none",
  xlab = "", ylab = "",
  main = "",

```

```

    margins = c(60,100,40,20),
    grid_color = "white",
    grid_width = 0.00001,
    titleX = FALSE,
    hide_colorbar = FALSE,
    branches_lwd = 0.1,
    label_names = c("Brain:", "Feature:", "Value"),
    fontsize_row = 9, fontsize_col = 9,
    labCol = colnames(abcd_efa$loadings[,1:2]),
    labRow = rownames(abcd_efa$loadings[,1:2]),
    heatmap_layers = theme(axis.line=element_blank()),
)

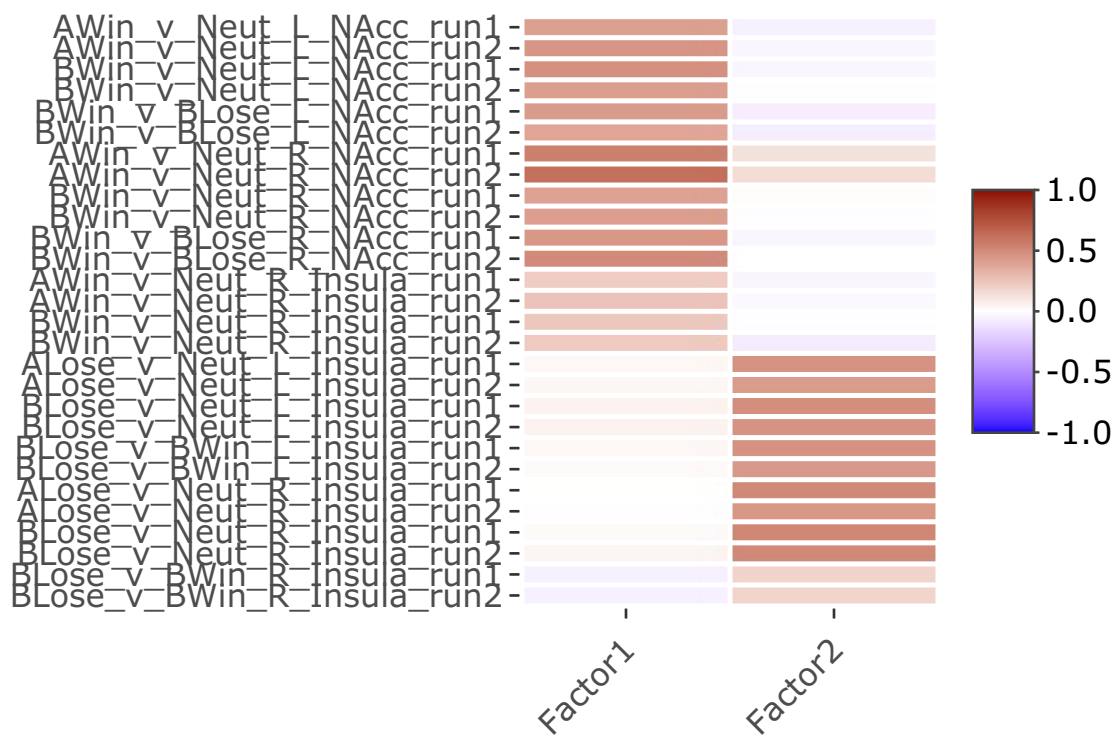
```

```

##                               Factor1 Factor2
## AWin_v_Neut_L_NAcc_run1      0.40   -0.06
## AWin_v_Neut_L_NAcc_run2      0.45   -0.04
## BWin_v_Neut_L_NAcc_run1      0.47   -0.04
## BWin_v_Neut_L_NAcc_run2      0.41   -0.01
## BWin_v_BLose_L_NAcc_run1      0.42   -0.07
## BWin_v_BLose_L_NAcc_run2      0.37   -0.07
## AWin_v_Neut_R_NAcc_run1      0.54    0.12
## AWin_v_Neut_R_NAcc_run2      0.61    0.14
## BWin_v_Neut_R_NAcc_run1      0.39    0.01
## BWin_v_Neut_R_NAcc_run2      0.41   -0.01
## BWin_v_BLose_R_NAcc_run1      0.44   -0.04
## BWin_v_BLose_R_NAcc_run2      0.49    0.00
## AWin_v_Neut_R_Insula_run1     0.21   -0.04
## AWin_v_Neut_R_Insula_run2     0.25   -0.03
## BWin_v_Neut_R_Insula_run1     0.23   -0.01
## BWin_v_Neut_R_Insula_run2     0.22   -0.08
## ALose_v_Neut_L_Insula_run1     0.03    0.46
## ALose_v_Neut_L_Insula_run2     0.03    0.42
## BLose_v_Neut_L_Insula_run1     0.05    0.48
## BLose_v_Neut_L_Insula_run2     0.05    0.46
## BLose_v_BWin_L_Insula_run1     0.03    0.46
## BLose_v_BWin_L_Insula_run2     0.02    0.43
## ALose_v_Neut_R_Insula_run1     0.00    0.50
## ALose_v_Neut_R_Insula_run2    -0.01    0.44
## BLose_v_Neut_R_Insula_run1     0.02    0.51
## BLose_v_Neut_R_Insula_run2     0.04    0.50
## BLose_v_BWin_R_Insula_run1    -0.06    0.18
## BLose_v_BWin_R_Insula_run2    -0.06    0.18

## Warning: `gather()` was deprecated in tidyr 1.2.0.
## i Please use `gather()` instead.
## i The deprecated feature was likely used in the plotly package.
##   Please report the issue at <https://github.com/plotly/plotly.R/issues>.

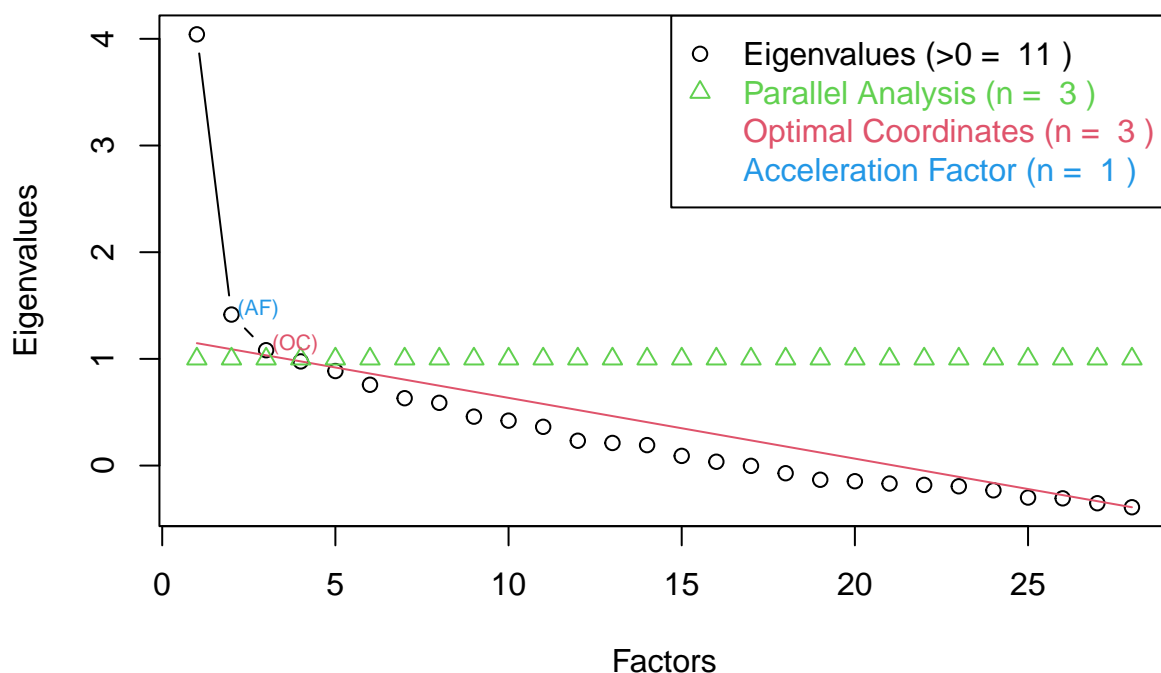
```



Sample: *MLS*

```
fa_mls <- subset(mls_efa_df[,1:28])
plot(nScree(x=fa_mls,model="factors"))
```


Non Graphical Solutions to Scree Test

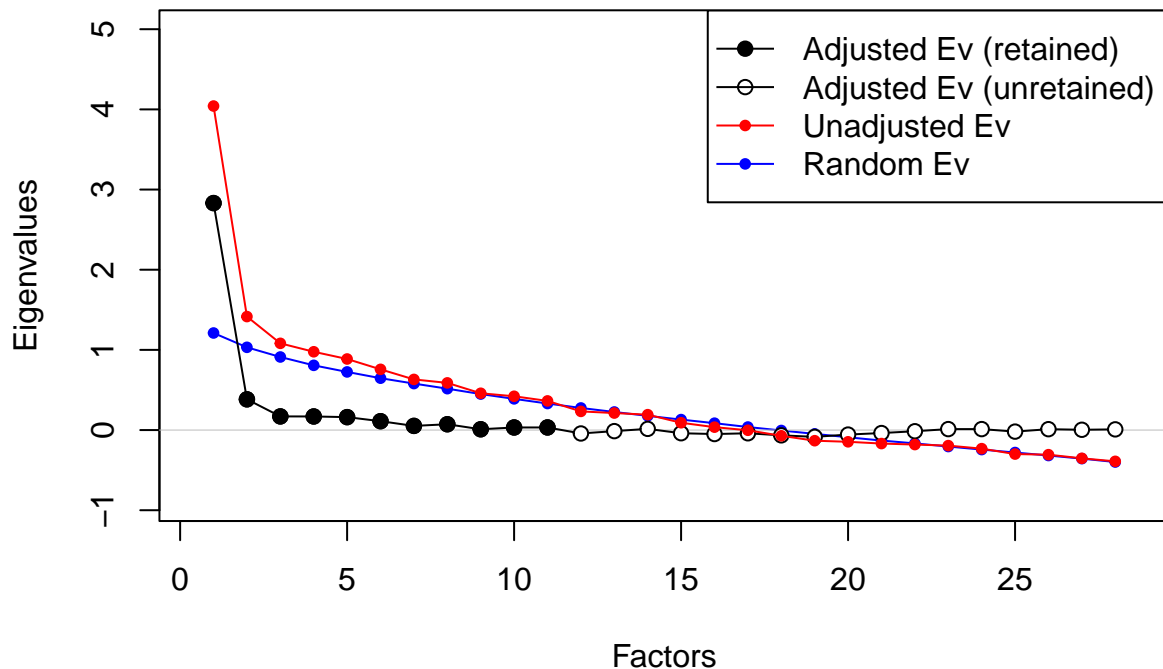


```
paran(x = fa_mls,
      iterations = 1000, quietly = FALSE, centile = 95,
      status = FALSE, all = TRUE, cfa = TRUE, graph = TRUE, color = TRUE,
      col = c("black", "red", "blue"), lty = c(1, 2, 3), lwd = 1, legend = TRUE,
      seed = 100)
```

```
##
## Using eigendecomposition of correlation matrix.
##
## Results of Horn's Parallel Analysis for factor retention
## 1000 iterations, using the 95 centile estimate
##
## -----
## Factor      Adjusted      Unadjusted      Estimated
##             Eigenvalue    Eigenvalue      Bias
## -----
## 1           2.831197      4.041832        1.210634
## 2           0.383345      1.415416        1.032070
## 3           0.170152      1.081438        0.911286
## 4           0.169402      0.976468        0.807066
## 5           0.161286      0.886233        0.724946
## 6           0.110471      0.757309        0.646837
## 7           0.051888      0.631413        0.579524
## 8           0.072342      0.588307        0.515965
## 9           0.009946      0.458769        0.448823
## 10          0.032104      0.421235        0.389130
```

```
## 11      0.032538    0.363316    0.330777
## 12     -0.042810    0.232814    0.275625
## 13     -0.013344    0.211961    0.225305
## 14      0.014754    0.191760    0.177006
## 15     -0.039773    0.090890    0.130663
## 16     -0.049639    0.036074    0.085713
## 17     -0.040487   -0.00265    0.037831
## 18     -0.066096   -0.07167   -0.00558
## 19     -0.084986   -0.13281   -0.04782
## 20     -0.056403   -0.14665   -0.09025
## 21     -0.037420   -0.16846   -0.13104
## 22     -0.014872   -0.18121   -0.16634
## 23      0.011672   -0.19418   -0.20586
## 24      0.011884   -0.23185   -0.24374
## 25     -0.019505   -0.29958   -0.28007
## 26      0.010809   -0.30709   -0.31790
## 27      0.003662   -0.35286   -0.35652
## 28      0.008372   -0.39046   -0.39884
## -----
##
## Adjusted eigenvalues > 0 indicate dimensions to retain.
## (11 factors      retained)
```

Parallel Analysis



```
mls_rec_factors <- matrix(NA, ncol = 2, nrow = 10)
colnames(mls_rec_factors) <- c("Nfactors", "BIC")

for (f in 1:10) {
```

```

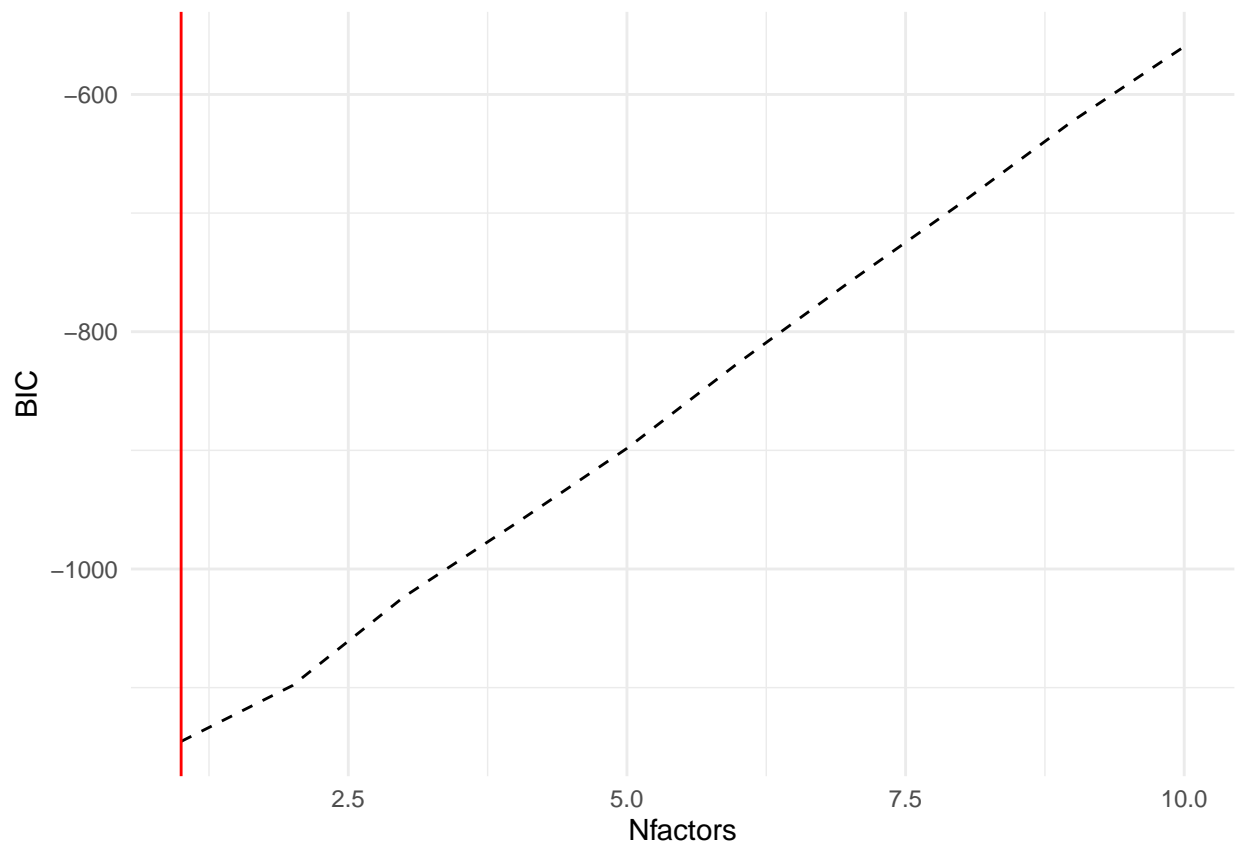
test_fac <- fa(r = fa_mls, #raw data
              nfactors = f, fm = 'minres',
              rotate = "promax")
mls_rec_factors[f,1] <- f
mls_rec_factors[f,2] <- test_fac$BIC
}

mls_bic_fact = as.data.frame(mls_rec_factors)

mls_lowest_bic <- which.min(mls_bic_fact$BIC)

mls_bic_fact %>%
  ggplot(aes(x = Nfactors, y = BIC)) +
  geom_line(colour = 'black', linetype = 'dashed') +
  geom_vline(xintercept = mls_bic_fact$Nfactors[mls_lowest_bic], colour = 'red') +
  theme_minimal()

```



MLS factor analysis

```

mls_efa <- factanal(x = fa_mls, #raw data
                  factors = 3, rotation = "promax" # oblique rotation allow for non-orthogonal structure
                  )

heatmaply(round(mls_efa$loadings[,1:3],2) %>% print(sort = T),
          scale_fill_gradient_fun = ggplot2::scale_fill_gradient2(
            low = "blue",
            high = "darkred",

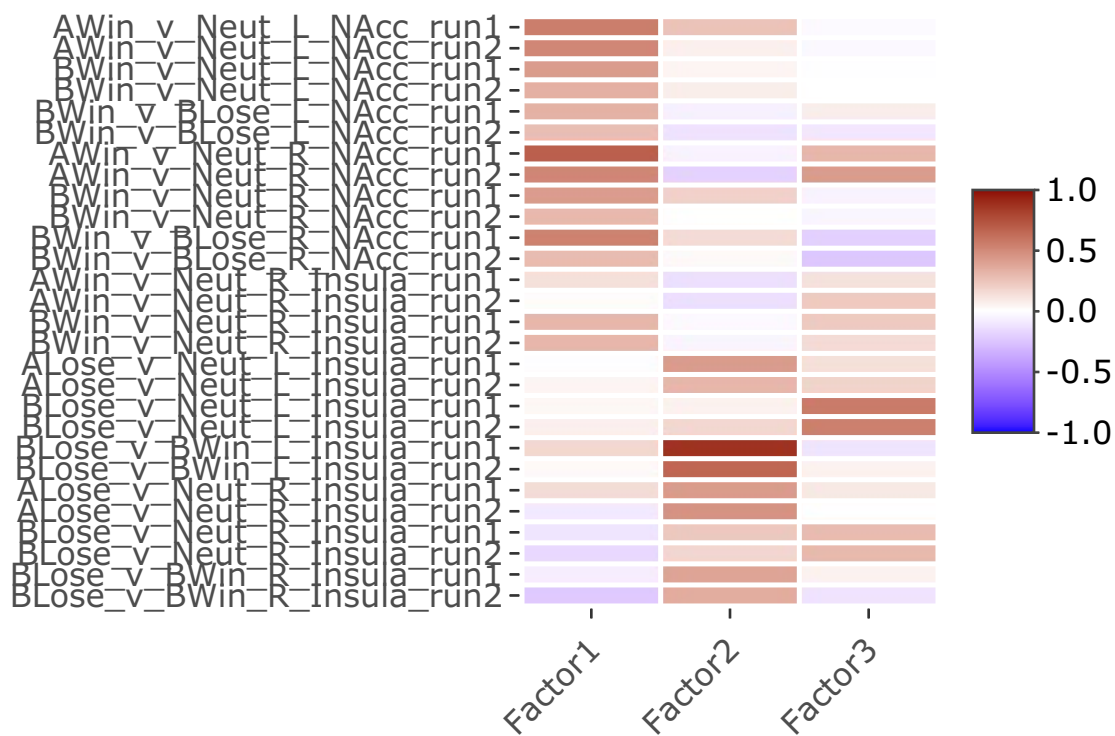
```

```

        space = "Lab",
        midpoint = 0,
        limits = c(-1, 1)
    ),
    dendrogram = "none",
    xlab = "", ylab = "",
    main = "",
    margins = c(60,100,40,20),
    grid_color = "white",
    grid_width = 0.00001,
    titleX = FALSE,
    hide_colorbar = FALSE,
    branches_lwd = 0.1,
    label_names = c("Brain:", "Feature:", "Value"),
    fontsize_row = 9, fontsize_col = 9,
    labCol = colnames(mls_efa$loadings[,1:3]),
    labRow = rownames(mls_efa$loadings[,1:3]),
    heatmap_layers = theme(axis.line=element_blank()),
)

```

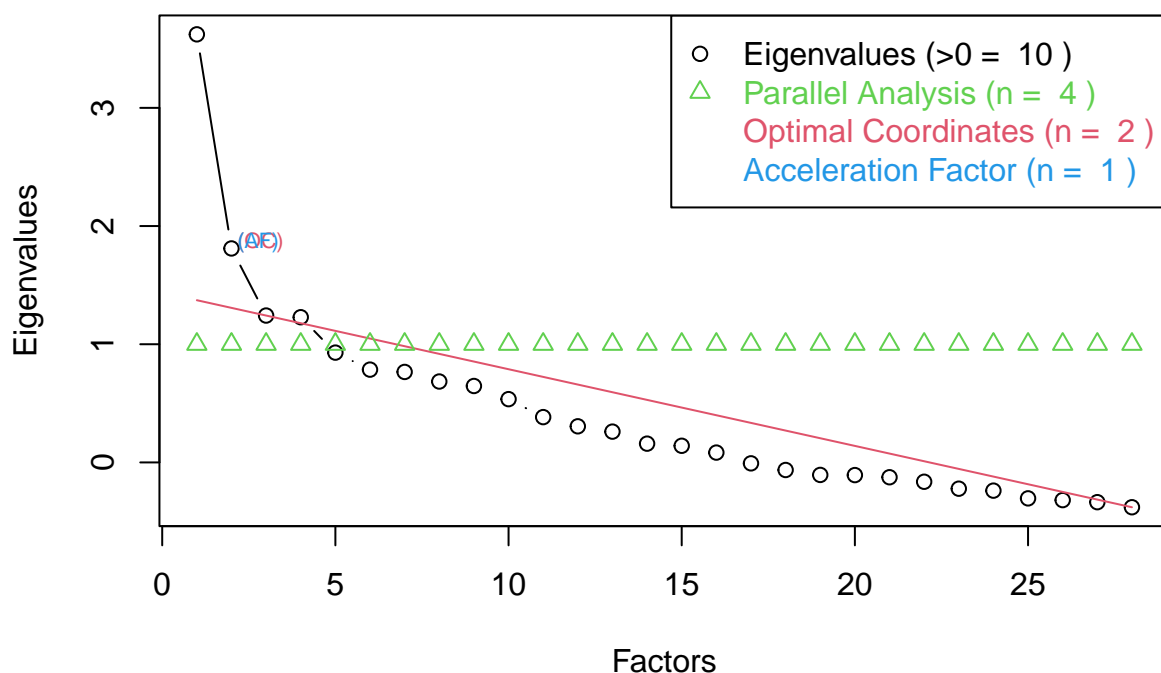
##	Factor1	Factor2	Factor3
## AWin_v_Neut_L_NAcc_run1	0.55	0.25	-0.02
## AWin_v_Neut_L_NAcc_run2	0.51	0.06	-0.03
## BWin_v_Neut_L_NAcc_run1	0.42	0.04	-0.01
## BWin_v_Neut_L_NAcc_run2	0.33	0.07	0.00
## BWin_v_BLose_L_NAcc_run1	0.32	-0.06	0.07
## BWin_v_BLose_L_NAcc_run2	0.27	-0.12	-0.10
## AWin_v_Neut_R_NAcc_run1	0.68	-0.05	0.30
## AWin_v_Neut_R_NAcc_run2	0.51	-0.19	0.42
## BWin_v_Neut_R_NAcc_run1	0.42	0.19	-0.05
## BWin_v_Neut_R_NAcc_run2	0.29	0.00	-0.04
## BWin_v_BLose_R_NAcc_run1	0.53	0.15	-0.20
## BWin_v_BLose_R_NAcc_run2	0.28	0.02	-0.24
## AWin_v_Neut_R_Insula_run1	0.13	-0.13	0.12
## AWin_v_Neut_R_Insula_run2	0.01	-0.13	0.22
## BWin_v_Neut_R_Insula_run1	0.30	-0.02	0.22
## BWin_v_Neut_R_Insula_run2	0.30	-0.04	0.15
## ALose_v_Neut_L_Insula_run1	-0.01	0.42	0.13
## ALose_v_Neut_L_Insula_run2	0.04	0.30	0.18
## BLose_v_Neut_L_Insula_run1	0.03	0.05	0.56
## BLose_v_Neut_L_Insula_run2	0.06	0.16	0.54
## BLose_v_BWin_L_Insula_run1	0.16	0.87	-0.11
## BLose_v_BWin_L_Insula_run2	0.02	0.65	0.05
## ALose_v_Neut_R_Insula_run1	0.14	0.42	0.09
## ALose_v_Neut_R_Insula_run2	-0.09	0.46	0.00
## BLose_v_Neut_R_Insula_run1	-0.11	0.23	0.28
## BLose_v_Neut_R_Insula_run2	-0.16	0.17	0.29
## BLose_v_BWin_R_Insula_run1	-0.07	0.38	0.05
## BLose_v_BWin_R_Insula_run2	-0.23	0.35	-0.12



Sample: *AHRB*

```
fa_ahrb <- subset(ahrb_efa_df[,1:28])
plot(nScree(x=fa_ahrb,model="factors"))
```

Non Graphical Solutions to Scree Test

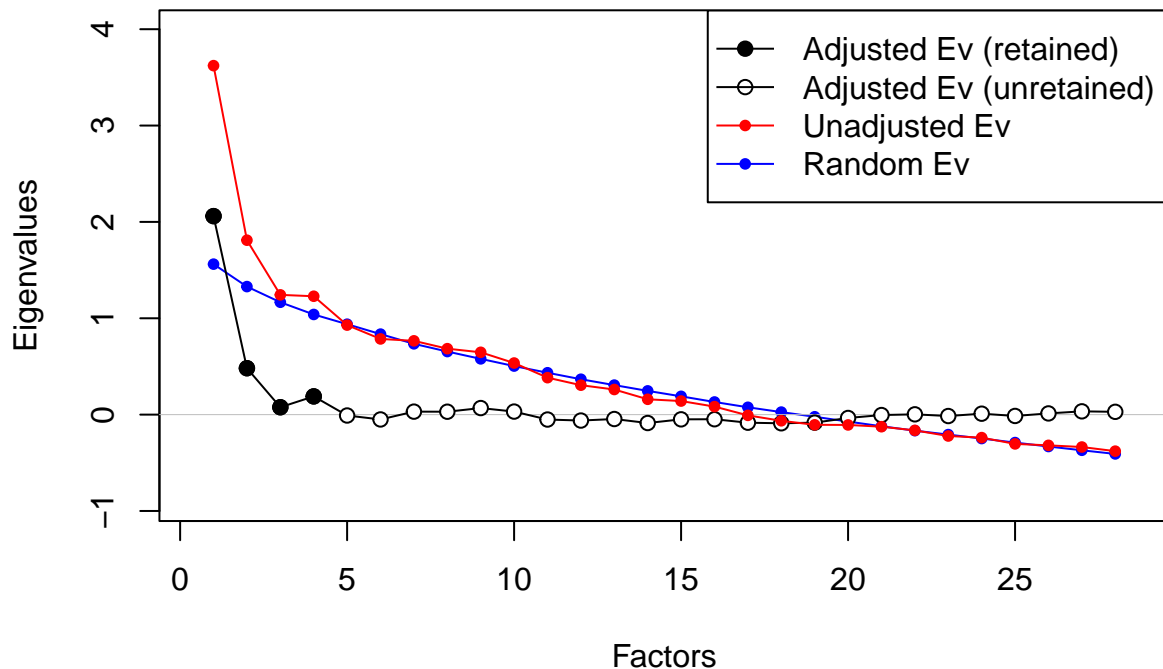


```
paran(x = fa_ahrb,
      iterations = 1000, quietly = FALSE, centile = 95,
      status = FALSE, all = TRUE, cfa = TRUE, graph = TRUE, color = TRUE,
      col = c("black", "red", "blue"), lty = c(1, 2, 3), lwd = 1, legend = TRUE,
      seed = 100)
```

```
##
## Using eigendecomposition of correlation matrix.
##
## Results of Horn's Parallel Analysis for factor retention
## 1000 iterations, using the 95 centile estimate
##
## -----
## Factor      Adjusted      Unadjusted      Estimated
##             Eigenvalue    Eigenvalue      Bias
## -----
## 1           2.060093      3.621941        1.561847
## 2           0.481724      1.810384        1.328660
## 3           0.076629      1.242826        1.166197
## 4           0.189243      1.228597        1.039354
## 5          -0.009360      0.929113        0.938474
## 6          -0.050147      0.785260        0.835407
## 7           0.030826      0.765686        0.734860
## 8           0.030113      0.684464        0.654351
## 9           0.068039      0.646615        0.578576
## 10          0.031492      0.535404        0.503911
```

```
## 11      -0.050998    0.383378    0.434377
## 12      -0.061551    0.305570    0.367122
## 13      -0.045811    0.260598    0.306410
## 14      -0.087183    0.158940    0.246124
## 15      -0.048996    0.140764    0.189761
## 16      -0.047886    0.083183    0.131069
## 17      -0.083831   -0.00786    0.075962
## 18      -0.090172   -0.06397    0.026192
## 19      -0.083539   -0.10575   -0.02221
## 20      -0.034889   -0.10723   -0.07234
## 21      -0.004630   -0.12541   -0.12078
## 22       0.002853   -0.16359   -0.16644
## 23      -0.014767   -0.22211   -0.20734
## 24       0.009603   -0.23872   -0.24833
## 25      -0.014051   -0.30408   -0.29003
## 26       0.012237   -0.31880   -0.33104
## 27       0.033983   -0.33629   -0.37027
## 28       0.028916   -0.37918   -0.40809
## -----
##
## Adjusted eigenvalues > 0 indicate dimensions to retain.
## (4 factors      retained)
```

Parallel Analysis



```
ahrb_rec_factors <- matrix(NA, ncol = 2, nrow = 10)
colnames(ahrb_rec_factors) <- c("Nfactors", "BIC")

for (f in 1:10) {
```

```

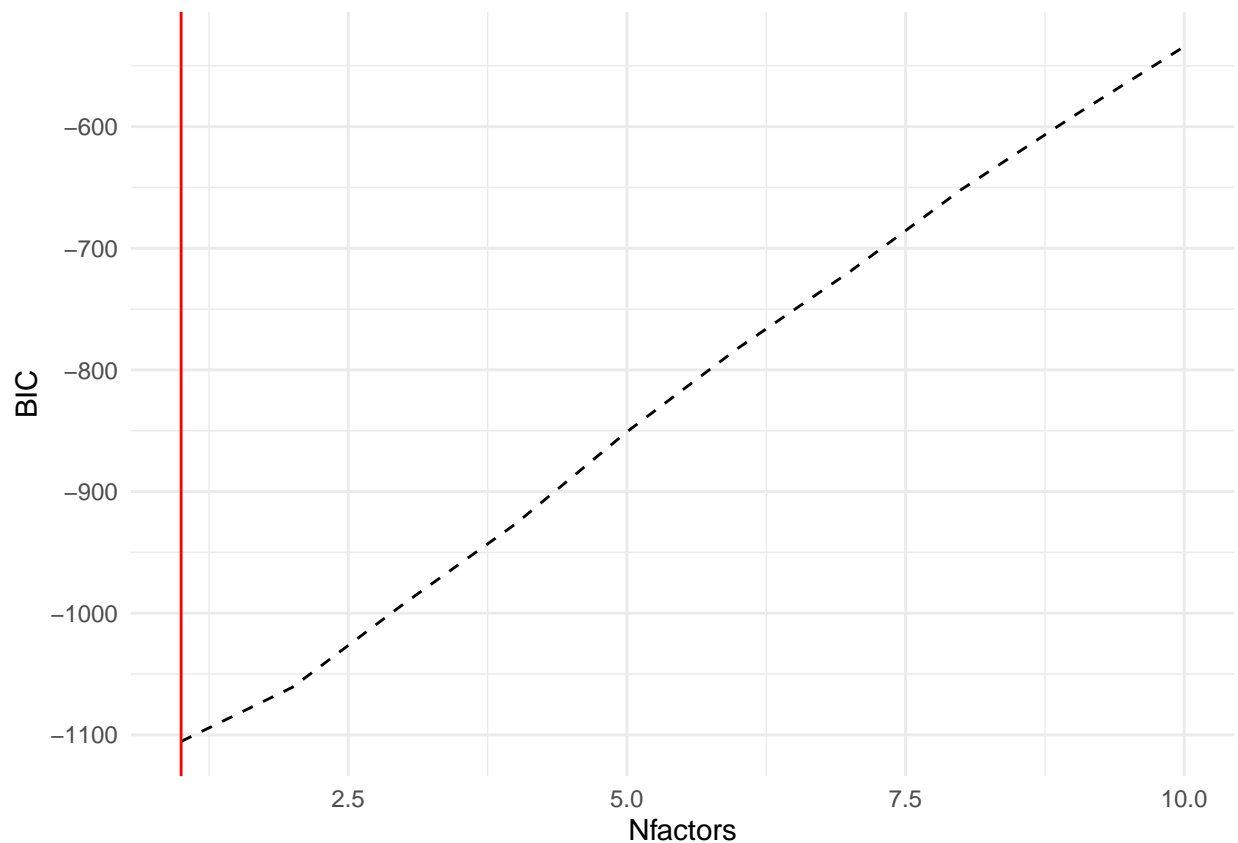
test_fac <- fa(r = fa_ahrb, #raw data
              nfactors = f, fm = 'minres',
              rotate = "promax")
ahrb_rec_factors[f,1] <- f
ahrb_rec_factors[f,2] <- test_fac$BIC
}

ahrb_bic_fact = as.data.frame(ahrb_rec_factors)

ahrb_lowest_bic <- which.min(ahrb_bic_fact$BIC)

ahrb_bic_fact %>%
  ggplot(aes(x = Nfactors, y = BIC)) +
  geom_line(colour = 'black', linetype = 'dashed') +
  geom_vline(xintercept = ahrb_bic_fact$Nfactors[ahrb_lowest_bic], colour = 'red')+
  theme_minimal()

```



Factor Analysis Model

```

ahrb_efa <- factanal(x = fa_ahrb, #raw data
                    factors = 4, rotation = "promax" # oblique rotation allow for non-orthogonal structure
                    )

heatmaply(round(ahrb_efa$loadings[,1:4],2) %>% print(sort = T),
          scale_fill_gradient_fun = ggplot2::scale_fill_gradient2(
            low = "blue",
            high = "darkred",

```

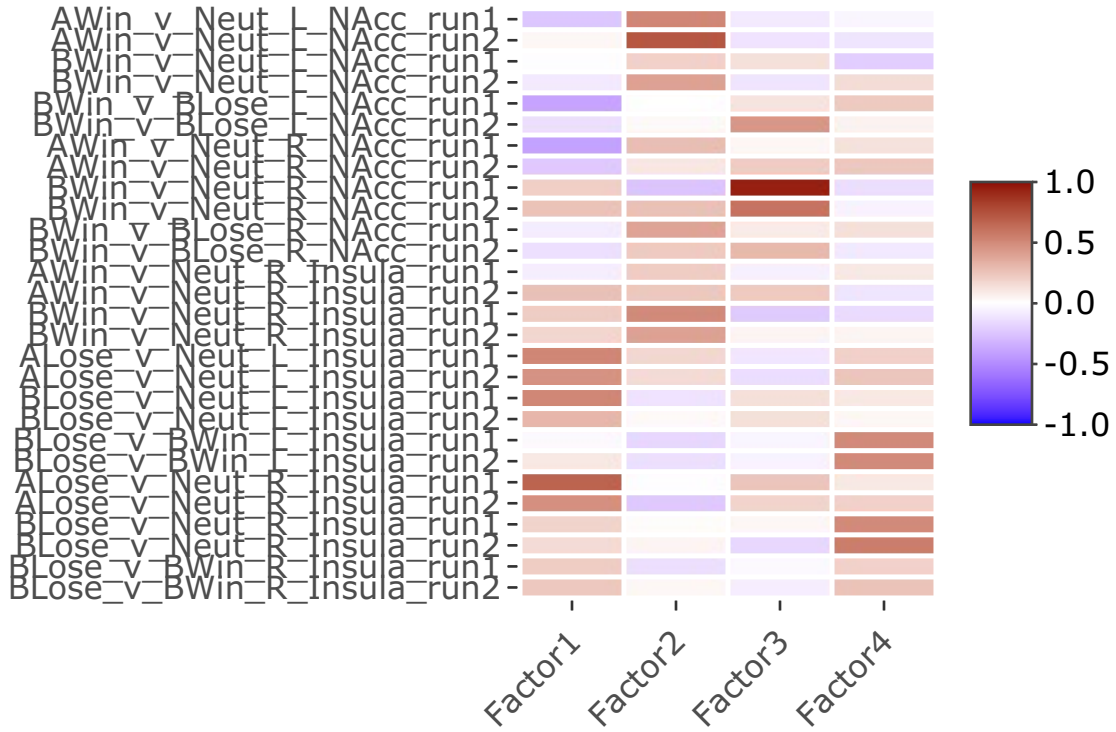


```

        space = "Lab",
        midpoint = 0,
        limits = c(-1, 1)
    ),
    dendrogram = "none",
    xlab = "", ylab = "",
    main = "",
    margins = c(60,100,40,20),
    grid_color = "white",
    grid_width = 0.00001,
    titleX = FALSE,
    hide_colorbar = FALSE,
    branches_lwd = 0.1,
    label_names = c("Brain:", "Feature:", "Value"),
    fontsize_row = 9, fontsize_col = 9,
    labCol = colnames(ahrb_efa$loadings[,1:4]),
    labRow = rownames(ahrb_efa$loadings[,1:4]),
    heatmap_layers = theme(axis.line=element_blank()),
)

```

##	Factor1	Factor2	Factor3	Factor4
## AWin_v_Neut_L_NAcc_run1	-0.24	0.51	-0.09	-0.04
## AWin_v_Neut_L_NAcc_run2	0.03	0.72	-0.12	-0.11
## BWin_v_Neut_L_NAcc_run1	-0.01	0.19	0.13	-0.21
## BWin_v_Neut_L_NAcc_run2	-0.09	0.39	-0.11	0.14
## BWin_v_BLose_L_NAcc_run1	-0.39	0.00	0.11	0.22
## BWin_v_BLose_L_NAcc_run2	-0.13	0.02	0.44	0.05
## AWin_v_Neut_R_NAcc_run1	-0.40	0.27	0.03	0.12
## AWin_v_Neut_R_NAcc_run2	-0.23	0.10	0.21	0.23
## BWin_v_Neut_R_NAcc_run1	0.20	-0.25	0.93	-0.14
## BWin_v_Neut_R_NAcc_run2	0.25	0.26	0.59	-0.05
## BWin_v_BLose_R_NAcc_run1	-0.08	0.38	0.08	0.13
## BWin_v_BLose_R_NAcc_run2	-0.13	0.22	0.29	-0.09
## AWin_v_Neut_R_Insula_run1	-0.07	0.21	-0.06	0.09
## AWin_v_Neut_R_Insula_run2	0.26	0.23	0.22	-0.11
## BWin_v_Neut_R_Insula_run1	0.21	0.49	-0.22	-0.15
## BWin_v_Neut_R_Insula_run2	0.17	0.39	0.04	0.04
## ALose_v_Neut_L_Insula_run1	0.51	0.16	-0.10	0.19
## ALose_v_Neut_L_Insula_run2	0.46	0.15	-0.14	0.24
## BLose_v_Neut_L_Insula_run1	0.51	-0.12	0.13	0.09
## BLose_v_Neut_L_Insula_run2	0.30	0.02	0.13	0.03
## BLose_v_BWin_L_Insula_run1	-0.02	-0.17	-0.04	0.49
## BLose_v_BWin_L_Insula_run2	0.09	-0.13	-0.05	0.49
## ALose_v_Neut_R_Insula_run1	0.66	-0.01	0.24	0.09
## ALose_v_Neut_R_Insula_run2	0.47	-0.23	0.18	0.19
## BLose_v_Neut_R_Insula_run1	0.18	0.01	0.03	0.49
## BLose_v_Neut_R_Insula_run2	0.15	0.04	-0.17	0.55
## BLose_v_BWin_R_Insula_run1	0.21	-0.13	-0.02	0.19
## BLose_v_BWin_R_Insula_run2	0.23	0.03	-0.07	0.25



Quant. Convergence

Calculating a coefficient of factor congruence across the three sample's EFA models. Using function `fa.congruence`

```
fa.congruence(x = list(abcd_efa, mls_efa, ahrb_efa), digits = 2) %>%
  knitr::kable(
    col.names = c("1. ABCD F1", "2. ABCD F2",
                  "3. MLS F1", "4. MLS F2", "5. MLS F3",
                  "6. AHRB F1", "7. AHRB F2", "8. AHRB F3", "8. AHRB F4"),
    caption = "ABCD, MLS and AHRB EFA Factor Congruence",
    booktabs = TRUE
  )
```

Table 2: ABCD, MLS and AHRB EFA Factor Congruence

	1. ABCD F1	2. ABCD F2	3. MLS F1	4. MLS F2	5. MLS F3	6. AHRB F1	7. AHRB F2	8. AHRB F3	8. AHRB F4
Factor1	1.00	0.03	0.94	0.03	0.20	-0.17	0.68	0.47	0.07
Factor2	0.03	1.00	-0.02	0.76	0.58	0.68	-0.11	0.04	0.77
Factor1	0.94	-0.02	1.00	0.07	0.15	-0.21	0.68	0.38	-0.01
Factor2	0.03	0.76	0.07	1.00	0.05	0.51	-0.07	0.02	0.67
Factor3	0.20	0.58	0.15	0.05	1.00	0.37	0.13	0.00	0.34
Factor1	-0.17	0.68	-0.21	0.51	0.37	1.00	-0.06	0.17	0.22
Factor2	0.68	-0.11	0.68	-0.07	0.13	-0.06	1.00	-0.12	-0.09

	1. ABCD F1	2. ABCD F2	3. MLS F1	4. MLS F2	5. MLS F3	6. AHRB F1	7. AHRB F2	8. AHRB F3	8. AHRB F4
Factor3	0.47	0.04	0.38	0.02	0.00	0.17	-0.12	1.00	-0.13
Factor4	0.07	0.77	-0.01	0.67	0.34	0.22	-0.09	-0.13	1.00

Running Local SEM

Running CFA for the pubertal variables in the ABCD sample using the local SEM framework described in Olaru et al (2020) implemented using the sirt package

Run LSEM

Specifying the model for the ABCD data below. For now, using the CFA model. In future [real data] implementation, will apply the EFA CFA from $n = 1000$ ABCD sample in the held out $n = 1000$ ABCD sample. Piloting on the simulated PDS scale

```
#first adding random PDS variable
set.seed(110)
sim_ABCD_data$PDS <- as.integer(rtnorm(n=1000, mean = 3.5, sd = 1.5,
                                       lower = 1, upper = 7))

lsem.MID <- sirt::lsem.estimate(data = sim_ABCD_data, moderator = 'PDS', # moderator variable
                              moderator.grid = seq(1,5,1), # moderator levels, PDS 1 - 5
                              lavmodel = MID_model, # model
                              h = 2, # bandwidth parameter
                              residualize = FALSE, # allow mean level differences
                              meanstructure = TRUE,
                              std.lv=TRUE
                              )

## ** Fit lavaan model
## |*****|
## |-----|
## ** Parameter summary
```

Summary LSEM

Summarizing output of the lsem.estimate

```
summary(lsem.MID)

## -----
## Local Structural Equation Model
##
## sirt 3.12-66 (2022-05-16 12:27:54)
## lavaan 0.6-12 (2022-07-04 16:40:02 UTC)
##
## R version 4.2.1 (2022-06-23) x86_64, darwin17.0 | nodename=DN0a241705.SUNet | login=root
##
## Function 'sirt::lsem.estimate', type='LSEM'
##
##
## Call:
```

```

## sirt::lsem.estimate(data = sim_ABCD_data, moderator = "PDS",
##   moderator.grid = seq(1, 5, 1), lavmodel = MID_model, h = 2,
##   residualize = FALSE, meanstructure = TRUE, std.lv = TRUE)
##
## Date of Analysis: 2022-11-17 16:10:50
## Time difference of 4.933373 secs
## Computation Time: 4.933373
##
## Number of observations in datasets = 1000
## Used observations in analysis = 1000
## Used sampling weights: FALSE
## Bandwidth factor = 2
## Bandwidth = 0.648
## Number of focal points for moderator = 5
##
## Used joint estimation: FALSE
## Used sufficient statistics: FALSE
## Used local linear smoothing: FALSE
## Used pseudo weights: FALSE
## Used lavaan package: TRUE
## Used lavaan.survey package: FALSE
##
## Mean structure modelled: TRUE
##
## lavaan Model
##
## # Factor loadings
## Approach =~ AWin_v_Neut_L_NAcc_run1 + AWin_v_Neut_R_NAcc_run1 + AWin_v_Neut_R_Insula_run1 +
##             AWin_v_Neut_L_NAcc_run2 + AWin_v_Neut_R_NAcc_run2 + AWin_v_Neut_R_Insula_run2 +
##             BWin_v_Neut_L_NAcc_run1 + BWin_v_Neut_R_NAcc_run1 + BWin_v_Neut_R_Insula_run1 +
##             BWin_v_Neut_L_NAcc_run2 + BWin_v_Neut_R_NAcc_run2 + BWin_v_Neut_R_Insula_run2 +
##             BWin_v_BLose_L_NAcc_run1 + BWin_v_BLose_R_NAcc_run1 +
##             BWin_v_BLose_L_NAcc_run2 + BWin_v_BLose_R_NAcc_run2
##
## Avoid =~     ALose_v_Neut_L_Insula_run1 + ALose_v_Neut_R_Insula_run1 +
##             ALose_v_Neut_L_Insula_run2 + ALose_v_Neut_R_Insula_run2 +
##             BLose_v_Neut_L_Insula_run1 + BLose_v_Neut_R_Insula_run1 +
##             BLose_v_Neut_L_Insula_run2 + BLose_v_Neut_R_Insula_run2 +
##             BLose_v_BWin_L_Insula_run1 + BLose_v_BWin_R_Insula_run1 +
##             BLose_v_BWin_L_Insula_run2 + BLose_v_BWin_R_Insula_run2
##
##
## Parameter Estimate Summary
##
##
##               par parindex      M      SD
## 1      Approach=~AWin_v_Neut_L_NAcc_run1      1 0.082 0.006
## 2      Approach=~AWin_v_Neut_R_NAcc_run1      2 0.086 0.004
## 3      Approach=~AWin_v_Neut_R_Insula_run1      3 0.041 0.004
## 4      Approach=~AWin_v_Neut_L_NAcc_run2      4 0.090 0.002
## 5      Approach=~AWin_v_Neut_R_NAcc_run2      5 0.096 0.006
## 6      Approach=~AWin_v_Neut_R_Insula_run2      6 0.049 0.007
## 7      Approach=~BWin_v_Neut_L_NAcc_run1      7 0.095 0.009
## 8      Approach=~BWin_v_Neut_R_NAcc_run1      8 0.070 0.007

```

## 9	Approach=~BWin_v_Neut_R_Insula_run1	9	0.044	0.005
## 10	Approach=~BWin_v_Neut_L_NAcc_run2	10	0.083	0.007
## 11	Approach=~BWin_v_Neut_R_NAcc_run2	11	0.079	0.015
## 12	Approach=~BWin_v_Neut_R_Insula_run2	12	0.050	0.011
## 13	Approach=~BWin_v_BLose_L_NAcc_run1	13	0.092	0.006
## 14	Approach=~BWin_v_BLose_R_NAcc_run1	14	0.089	0.005
## 15	Approach=~BWin_v_BLose_L_NAcc_run2	15	0.079	0.008
## 16	Approach=~BWin_v_BLose_R_NAcc_run2	16	0.095	0.004
## 17	Avoid=~ALose_v_Neut_L_Insula_run1	17	0.087	0.003
## 18	Avoid=~ALose_v_Neut_R_Insula_run1	18	0.094	0.009
## 19	Avoid=~ALose_v_Neut_L_Insula_run2	19	0.078	0.008
## 20	Avoid=~ALose_v_Neut_R_Insula_run2	20	0.090	0.012
## 21	Avoid=~BLose_v_Neut_L_Insula_run1	21	0.086	0.013
## 22	Avoid=~BLose_v_Neut_R_Insula_run1	22	0.092	0.006
## 23	Avoid=~BLose_v_Neut_L_Insula_run2	23	0.082	0.006
## 24	Avoid=~BLose_v_Neut_R_Insula_run2	24	0.091	0.005
## 25	Avoid=~BLose_v_BWin_L_Insula_run1	25	0.085	0.016
## 26	Avoid=~BLose_v_BWin_R_Insula_run1	26	0.040	0.010
## 27	Avoid=~BLose_v_BWin_L_Insula_run2	27	0.077	0.014
## 28	Avoid=~BLose_v_BWin_R_Insula_run2	28	0.041	0.004
## 29	AWin_v_Neut_L_NAcc_run1~~AWin_v_Neut_L_NAcc_run1	29	0.028	0.001
## 30	AWin_v_Neut_R_NAcc_run1~~AWin_v_Neut_R_NAcc_run1	30	0.031	0.002
## 31	AWin_v_Neut_R_Insula_run1~~AWin_v_Neut_R_Insula_run1	31	0.029	0.002
## 32	AWin_v_Neut_L_NAcc_run2~~AWin_v_Neut_L_NAcc_run2	32	0.027	0.001
## 33	AWin_v_Neut_R_NAcc_run2~~AWin_v_Neut_R_NAcc_run2	33	0.028	0.002
## 34	AWin_v_Neut_R_Insula_run2~~AWin_v_Neut_R_Insula_run2	34	0.028	0.002
## 35	BWin_v_Neut_L_NAcc_run1~~BWin_v_Neut_L_NAcc_run1	35	0.025	0.001
## 36	BWin_v_Neut_R_NAcc_run1~~BWin_v_Neut_R_NAcc_run1	36	0.029	0.001
## 37	BWin_v_Neut_R_Insula_run1~~BWin_v_Neut_R_Insula_run1	37	0.031	0.001
## 38	BWin_v_Neut_L_NAcc_run2~~BWin_v_Neut_L_NAcc_run2	38	0.030	0.001
## 39	BWin_v_Neut_R_NAcc_run2~~BWin_v_Neut_R_NAcc_run2	39	0.030	0.002
## 40	BWin_v_Neut_R_Insula_run2~~BWin_v_Neut_R_Insula_run2	40	0.031	0.001
## 41	BWin_v_BLose_L_NAcc_run1~~BWin_v_BLose_L_NAcc_run1	41	0.028	0.002
## 42	BWin_v_BLose_R_NAcc_run1~~BWin_v_BLose_R_NAcc_run1	42	0.030	0.003
## 43	BWin_v_BLose_L_NAcc_run2~~BWin_v_BLose_L_NAcc_run2	43	0.028	0.001
## 44	BWin_v_BLose_R_NAcc_run2~~BWin_v_BLose_R_NAcc_run2	44	0.028	0.002
## 45	ALose_v_Neut_L_Insula_run1~~ALose_v_Neut_L_Insula_run1	45	0.030	0.002
## 46	ALose_v_Neut_R_Insula_run1~~ALose_v_Neut_R_Insula_run1	46	0.027	0.002
## 47	ALose_v_Neut_L_Insula_run2~~ALose_v_Neut_L_Insula_run2	47	0.029	0.001
## 48	ALose_v_Neut_R_Insula_run2~~ALose_v_Neut_R_Insula_run2	48	0.029	0.003
## 49	BLose_v_Neut_L_Insula_run1~~BLose_v_Neut_L_Insula_run1	49	0.029	0.001
## 50	BLose_v_Neut_R_Insula_run1~~BLose_v_Neut_R_Insula_run1	50	0.025	0.001
## 51	BLose_v_Neut_L_Insula_run2~~BLose_v_Neut_L_Insula_run2	51	0.030	0.002
## 52	BLose_v_Neut_R_Insula_run2~~BLose_v_Neut_R_Insula_run2	52	0.029	0.002
## 53	BLose_v_BWin_L_Insula_run1~~BLose_v_BWin_L_Insula_run1	53	0.030	0.002
## 54	BLose_v_BWin_R_Insula_run1~~BLose_v_BWin_R_Insula_run1	54	0.031	0.002
## 55	BLose_v_BWin_L_Insula_run2~~BLose_v_BWin_L_Insula_run2	55	0.029	0.002
## 56	BLose_v_BWin_R_Insula_run2~~BLose_v_BWin_R_Insula_run2	56	0.029	0.001
## 57	Approach~~Approach	57	1.000	0.000
## 58	Avoid~~Avoid	58	1.000	0.000
## 59	Approach~~Avoid	59	-0.577	0.037
## 60	AWin_v_Neut_L_NAcc_run1~1	60	-0.003	0.016
## 61	AWin_v_Neut_R_NAcc_run1~1	61	-0.004	0.009
## 62	AWin_v_Neut_R_Insula_run1~1	62	0.007	0.004

##	63				AWin_v_Neut_L_NAcc_run2~1	63	-0.002	0.009
##	64				AWin_v_Neut_R_NAcc_run2~1	64	0.003	0.007
##	65				AWin_v_Neut_R_Insula_run2~1	65	0.005	0.003
##	66				BWin_v_Neut_L_NAcc_run1~1	66	-0.011	0.003
##	67				BWin_v_Neut_R_NAcc_run1~1	67	-0.006	0.008
##	68				BWin_v_Neut_R_Insula_run1~1	68	0.013	0.004
##	69				BWin_v_Neut_L_NAcc_run2~1	69	-0.005	0.003
##	70				BWin_v_Neut_R_NAcc_run2~1	70	-0.001	0.007
##	71				BWin_v_Neut_R_Insula_run2~1	71	-0.001	0.006
##	72				BWin_v_BLose_L_NAcc_run1~1	72	-0.003	0.005
##	73				BWin_v_BLose_R_NAcc_run1~1	73	-0.005	0.011
##	74				BWin_v_BLose_L_NAcc_run2~1	74	-0.008	0.010
##	75				BWin_v_BLose_R_NAcc_run2~1	75	-0.008	0.008
##	76				ALose_v_Neut_L_Insula_run1~1	76	-0.005	0.004
##	77				ALose_v_Neut_R_Insula_run1~1	77	0.003	0.005
##	78				ALose_v_Neut_L_Insula_run2~1	78	-0.001	0.005
##	79				ALose_v_Neut_R_Insula_run2~1	79	0.002	0.005
##	80				BLose_v_Neut_L_Insula_run1~1	80	0.002	0.009
##	81				BLose_v_Neut_R_Insula_run1~1	81	0.014	0.008
##	82				BLose_v_Neut_L_Insula_run2~1	82	0.013	0.003
##	83				BLose_v_Neut_R_Insula_run2~1	83	0.005	0.006
##	84				BLose_v_BWin_L_Insula_run1~1	84	0.008	0.006
##	85				BLose_v_BWin_R_Insula_run1~1	85	0.002	0.005
##	86				BLose_v_BWin_L_Insula_run2~1	86	0.005	0.004
##	87				BLose_v_BWin_R_Insula_run2~1	87	-0.003	0.012
##	88				Approach~1	88	0.000	0.000
##	89				Avoid~1	89	0.000	0.000
##	90				rmsea	90	0.075	0.005
##	91				cfi	91	0.598	0.039
##	92				tli	92	0.564	0.042
##	93				gfi	93	0.860	0.014
##	94				srmr	94	0.061	0.004
##		MAD	Min	Max	lin_int	lin_slo	SD_nonlin	
##	1	0.006	0.074	0.090	0.092	-0.003	0.005	
##	2	0.003	0.081	0.092	0.087	0.000	0.003	
##	3	0.003	0.036	0.047	0.043	-0.001	0.003	
##	4	0.002	0.085	0.093	0.095	-0.002	0.001	
##	5	0.003	0.086	0.109	0.107	-0.004	0.004	
##	6	0.006	0.039	0.061	0.033	0.005	0.003	
##	7	0.005	0.071	0.105	0.115	-0.007	0.004	
##	8	0.006	0.063	0.080	0.060	0.003	0.005	
##	9	0.003	0.034	0.055	0.056	-0.004	0.003	
##	10	0.005	0.069	0.090	0.080	0.001	0.006	
##	11	0.012	0.056	0.100	0.042	0.013	0.002	
##	12	0.009	0.037	0.072	0.074	-0.008	0.006	
##	13	0.005	0.076	0.097	0.101	-0.003	0.005	
##	14	0.004	0.083	0.099	0.082	0.002	0.004	
##	15	0.005	0.061	0.086	0.070	0.003	0.007	
##	16	0.003	0.086	0.099	0.100	-0.002	0.003	
##	17	0.002	0.082	0.092	0.088	0.000	0.003	
##	18	0.008	0.080	0.105	0.072	0.007	0.003	
##	19	0.007	0.068	0.088	0.096	-0.006	0.003	
##	20	0.009	0.068	0.104	0.079	0.004	0.011	
##	21	0.012	0.058	0.097	0.067	0.006	0.011	

##	22	0.005	0.083	0.099	0.103	-0.004	0.004
##	23	0.005	0.070	0.088	0.073	0.003	0.005
##	24	0.004	0.081	0.095	0.089	0.001	0.005
##	25	0.015	0.062	0.104	0.048	0.013	0.007
##	26	0.009	0.028	0.056	0.057	-0.006	0.008
##	27	0.012	0.057	0.097	0.044	0.011	0.006
##	28	0.004	0.037	0.050	0.049	-0.003	0.003
##	29	0.001	0.026	0.029	0.027	0.000	0.001
##	30	0.002	0.028	0.033	0.026	0.002	0.001
##	31	0.002	0.026	0.032	0.029	0.000	0.002
##	32	0.001	0.025	0.027	0.026	0.000	0.001
##	33	0.001	0.027	0.033	0.025	0.001	0.001
##	34	0.002	0.025	0.032	0.034	-0.002	0.001
##	35	0.001	0.023	0.028	0.022	0.001	0.001
##	36	0.001	0.027	0.030	0.029	0.000	0.001
##	37	0.001	0.029	0.032	0.033	-0.001	0.001
##	38	0.001	0.028	0.032	0.033	-0.001	0.001
##	39	0.001	0.024	0.032	0.032	-0.001	0.002
##	40	0.001	0.029	0.033	0.031	0.000	0.001
##	41	0.001	0.025	0.033	0.032	-0.001	0.001
##	42	0.002	0.021	0.032	0.035	-0.002	0.002
##	43	0.001	0.027	0.029	0.030	-0.001	0.001
##	44	0.002	0.024	0.029	0.023	0.002	0.001
##	45	0.002	0.026	0.032	0.032	-0.001	0.002
##	46	0.002	0.024	0.030	0.032	-0.002	0.001
##	47	0.001	0.028	0.032	0.028	0.000	0.001
##	48	0.003	0.026	0.036	0.036	-0.002	0.002
##	49	0.001	0.025	0.029	0.031	-0.001	0.001
##	50	0.001	0.024	0.026	0.026	0.000	0.001
##	51	0.002	0.027	0.033	0.031	0.000	0.002
##	52	0.001	0.026	0.031	0.024	0.002	0.000
##	53	0.002	0.027	0.031	0.032	-0.001	0.002
##	54	0.002	0.028	0.035	0.036	-0.002	0.001
##	55	0.001	0.025	0.031	0.032	-0.001	0.001
##	56	0.001	0.028	0.030	0.030	0.000	0.001
##	57	0.000	1.000	1.000	1.000	0.000	0.000
##	58	0.000	1.000	1.000	1.000	0.000	0.000
##	59	0.033	-0.621	-0.517	-0.517	-0.020	0.028
##	60	0.014	-0.017	0.028	0.034	-0.013	0.006
##	61	0.008	-0.020	0.007	0.003	-0.002	0.008
##	62	0.003	-0.001	0.010	0.005	0.001	0.004
##	63	0.007	-0.012	0.013	0.018	-0.007	0.002
##	64	0.005	-0.011	0.011	0.016	-0.004	0.004
##	65	0.003	0.001	0.009	0.010	-0.002	0.003
##	66	0.003	-0.014	-0.005	-0.013	0.001	0.003
##	67	0.006	-0.011	0.014	0.007	-0.004	0.006
##	68	0.003	0.008	0.017	0.008	0.002	0.003
##	69	0.003	-0.009	0.000	-0.007	0.001	0.003
##	70	0.005	-0.006	0.018	0.010	-0.004	0.006
##	71	0.004	-0.010	0.015	0.011	-0.004	0.004
##	72	0.007	-0.010	0.016	0.005	-0.003	0.008
##	73	0.010	-0.019	0.010	0.018	-0.008	0.007
##	74	0.010	-0.017	0.011	0.007	-0.005	0.008
##	75	0.008	-0.017	0.005	0.003	-0.004	0.007

```

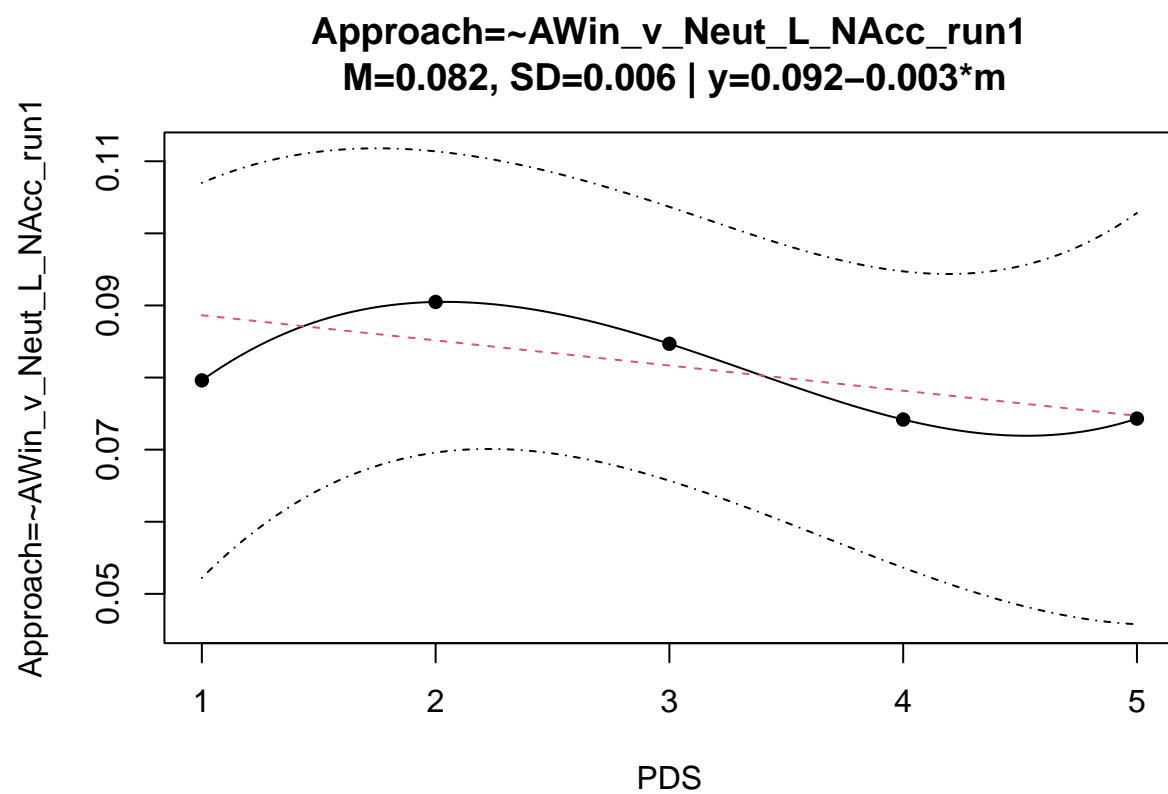
## 76 0.004 -0.009 0.003 -0.008 0.001 0.004
## 77 0.004 -0.008 0.009 0.015 -0.004 0.001
## 78 0.004 -0.008 0.007 -0.005 0.001 0.005
## 79 0.003 -0.010 0.007 0.009 -0.002 0.004
## 80 0.007 -0.006 0.024 0.012 -0.003 0.008
## 81 0.007 0.003 0.025 -0.006 0.007 0.003
## 82 0.003 0.009 0.016 0.011 0.001 0.003
## 83 0.006 -0.003 0.011 -0.002 0.002 0.005
## 84 0.005 0.001 0.016 0.018 -0.003 0.004
## 85 0.003 -0.011 0.006 0.008 -0.002 0.004
## 86 0.003 -0.002 0.009 0.007 -0.001 0.004
## 87 0.010 -0.030 0.005 -0.028 0.008 0.007
## 88 0.000 0.000 0.000 0.000 0.000 0.000
## 89 0.000 0.000 0.000 0.000 0.000 0.000
## 90 0.005 0.071 0.086 0.071 0.001 0.005
## 91 0.030 0.527 0.635 0.594 0.001 0.038
## 92 0.033 0.488 0.605 0.560 0.001 0.042
## 93 0.013 0.830 0.872 0.872 -0.004 0.013
## 94 0.003 0.058 0.069 0.060 0.000 0.004
##
## Distribution of Moderator: Density and Effective Sample Size
##
## M=3.083 | SD=1.289
##
##      moderator    wgt    Neff
## 1           1 0.125 188.687
## 2           2 0.227 345.959
## 3           3 0.307 433.724
## 4           4 0.243 353.279
## 5           5 0.098 179.382
##
##      variable      M      SD      min      max
## 1 moderator    3.083    1.289    1.000    6.000
## 2          wgt    0.200    0.087    0.098    0.307
## 3         Neff 300.206 111.548 179.382 433.724

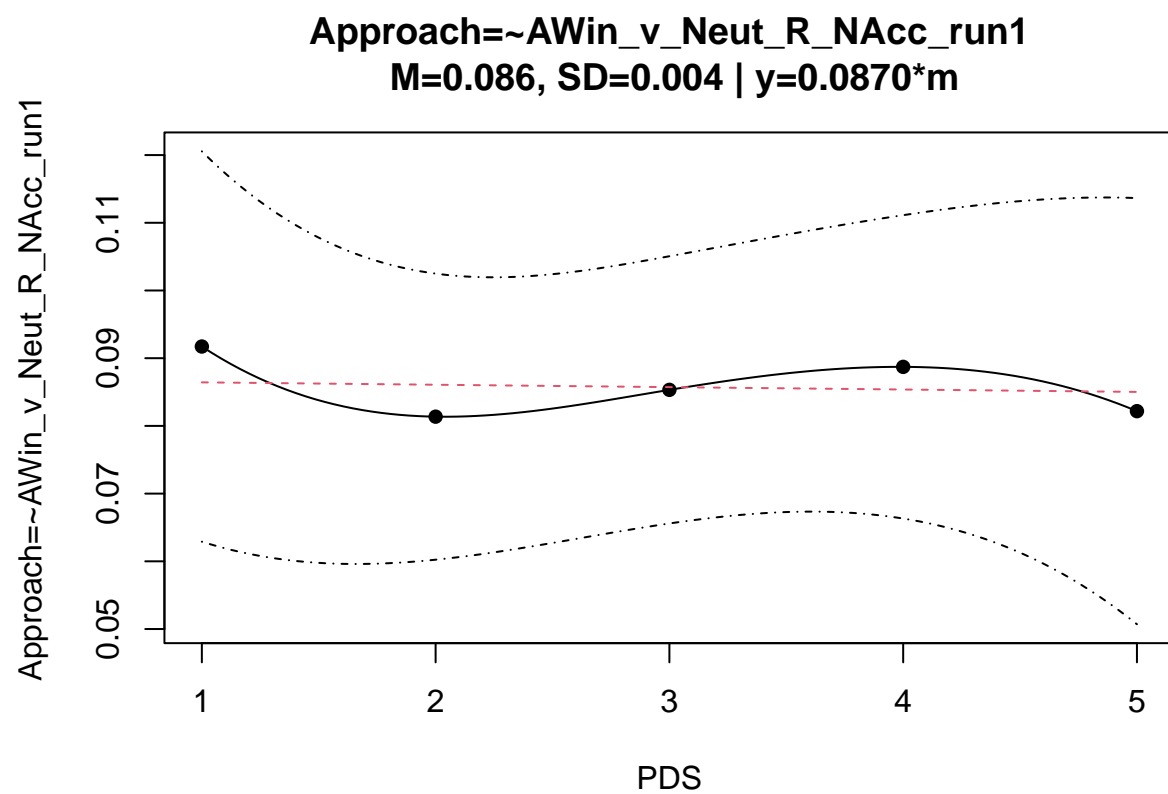
```

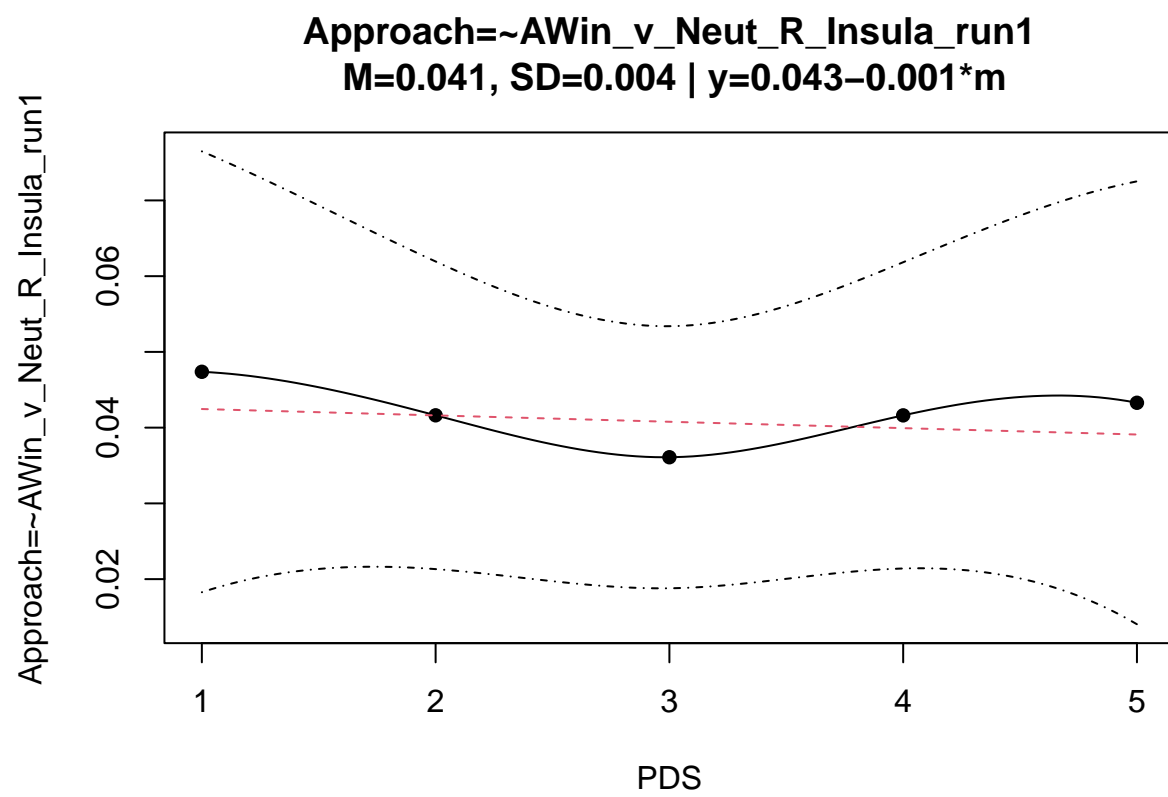
Plot LSEM

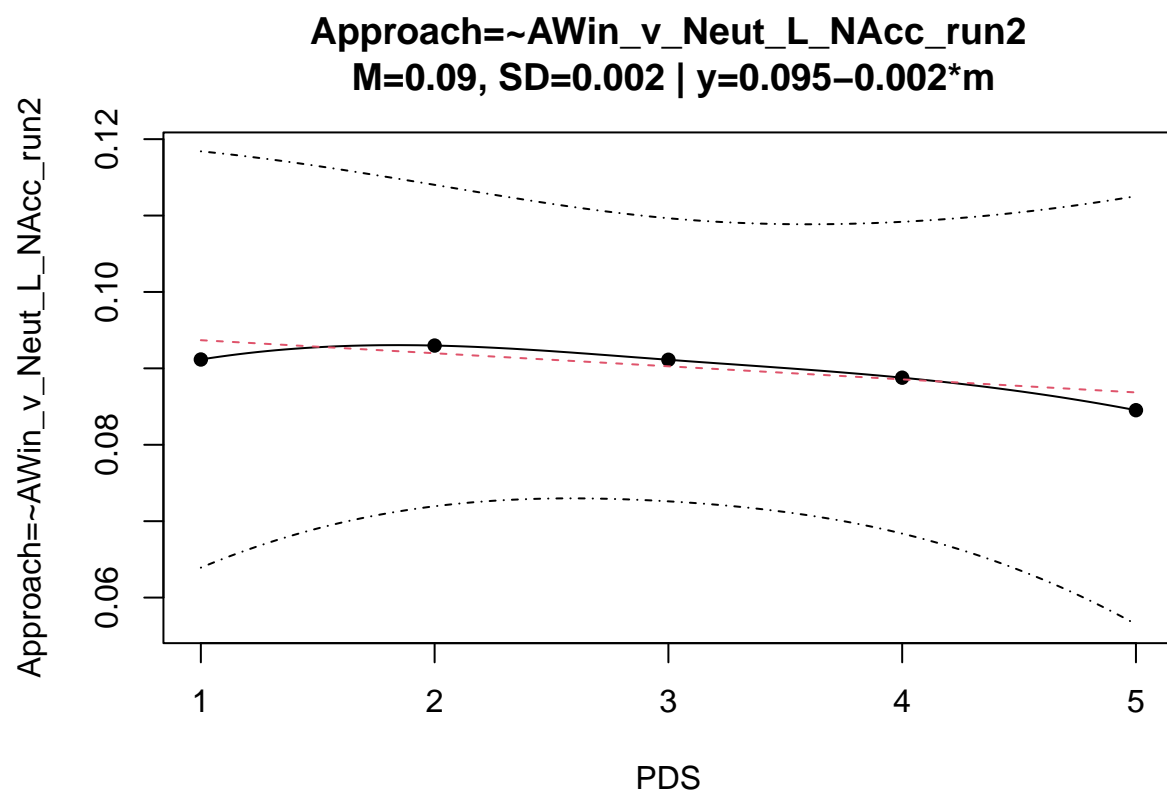
Plotting the `lsem.estimate` for the first 20 indexes.

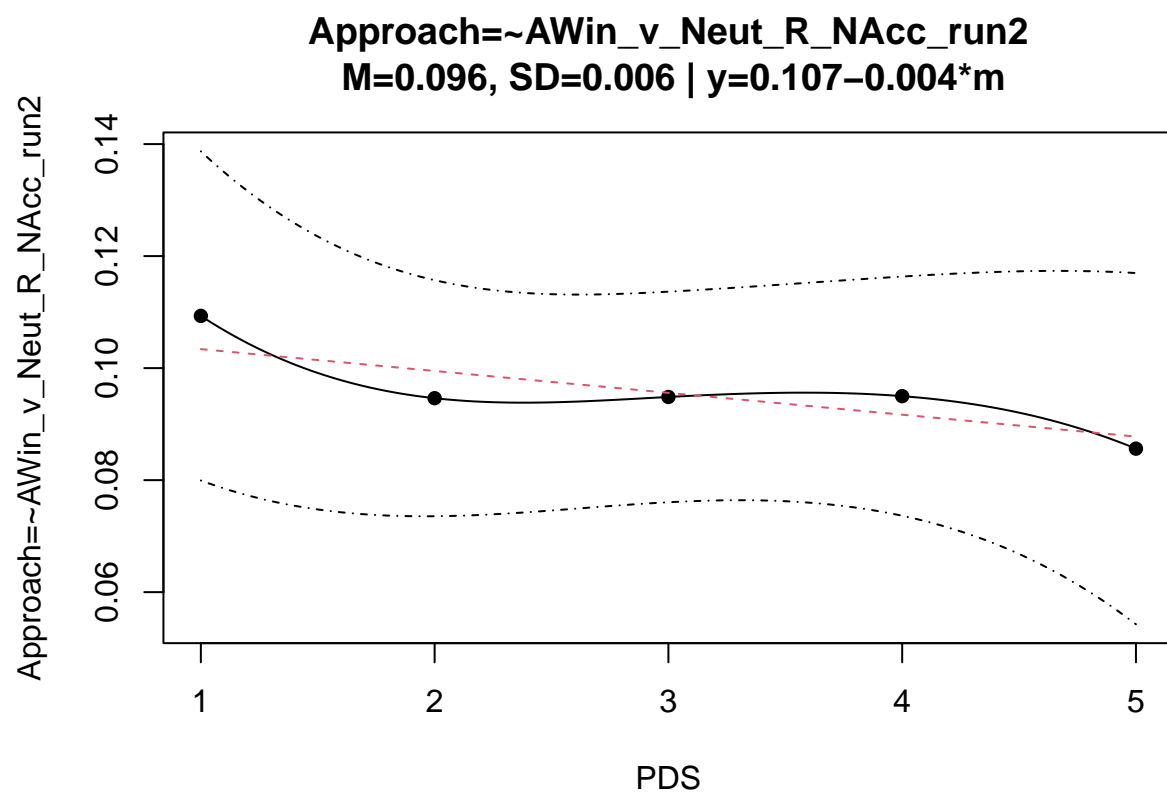
```
plot(lsem.MID, parindex=1:20)
```

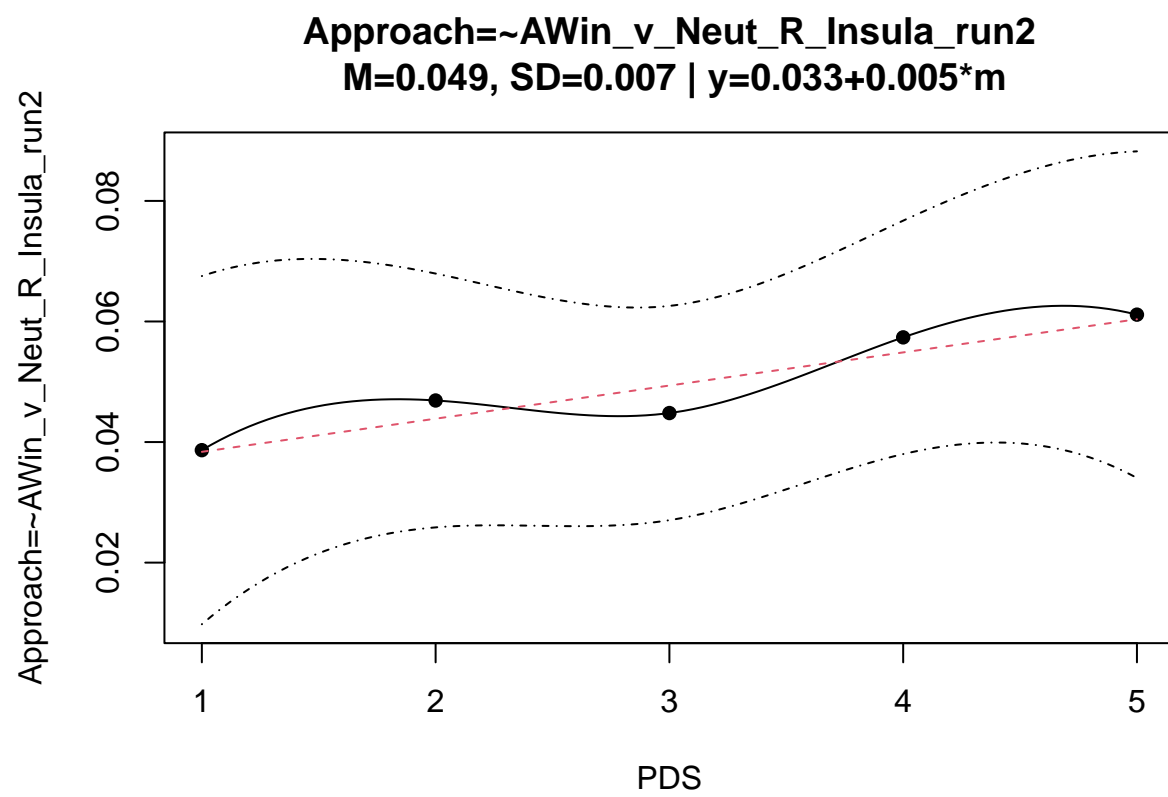



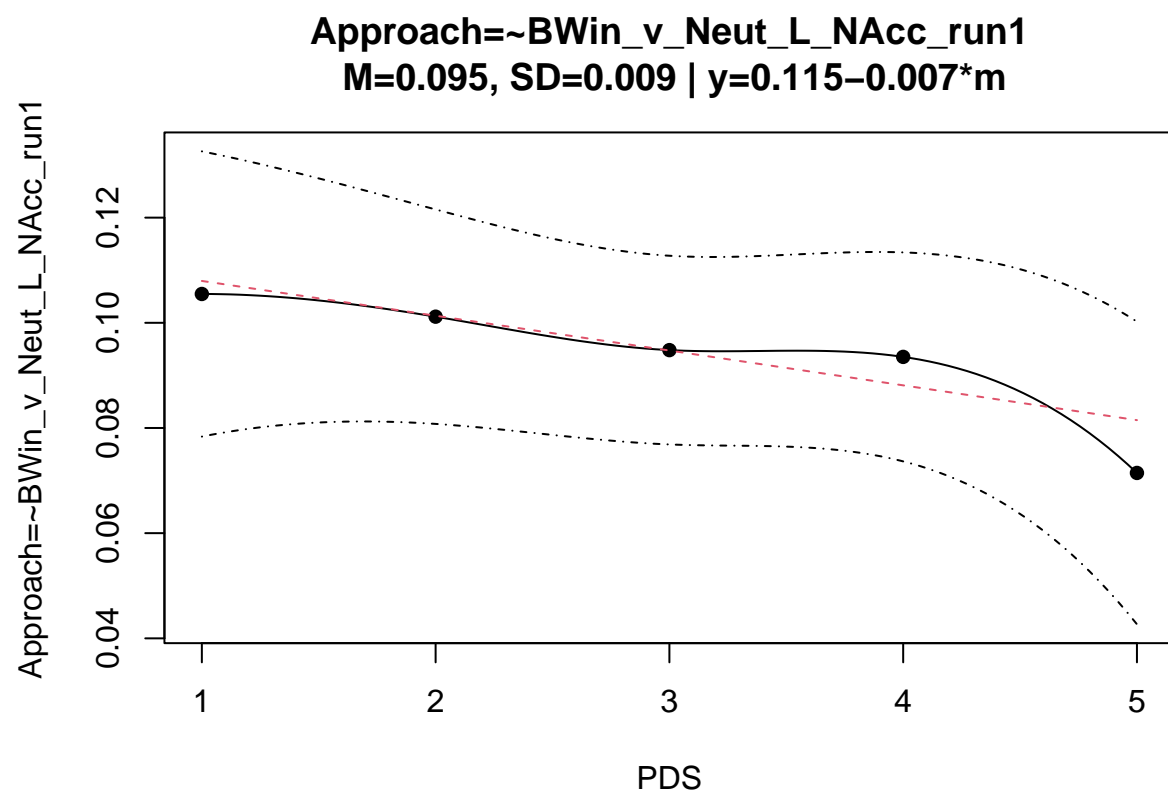


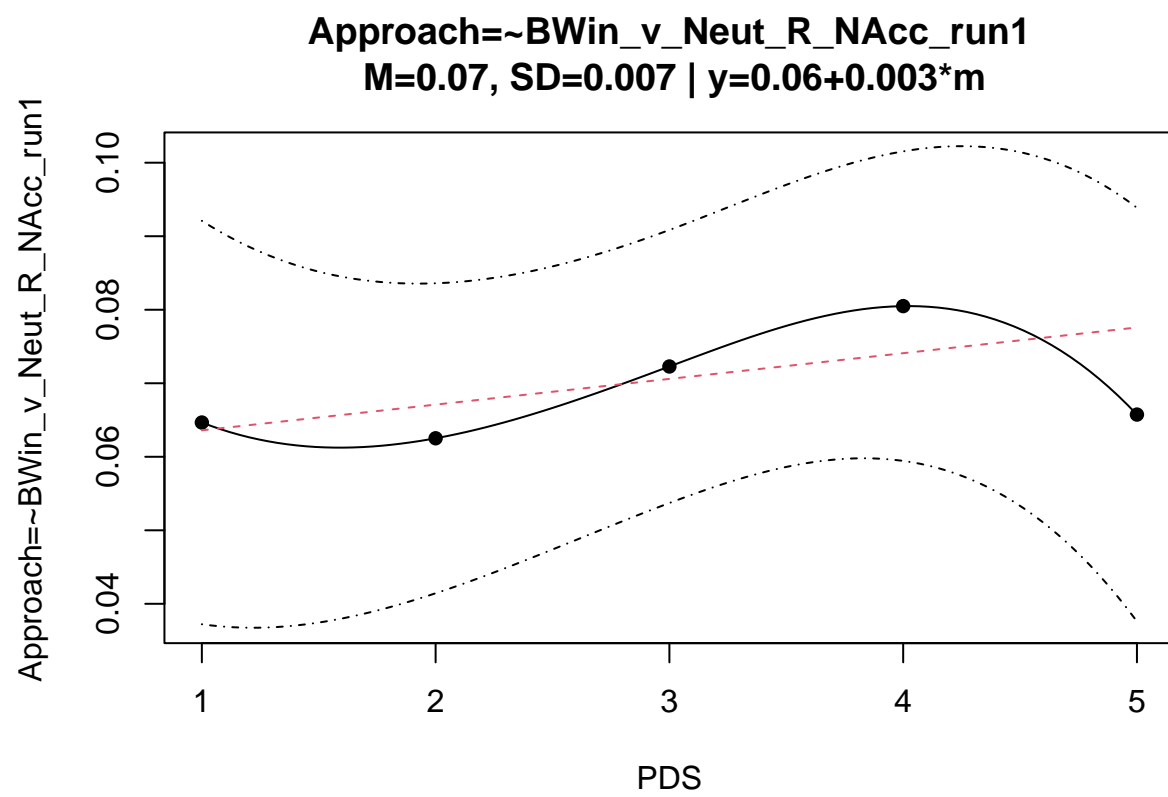


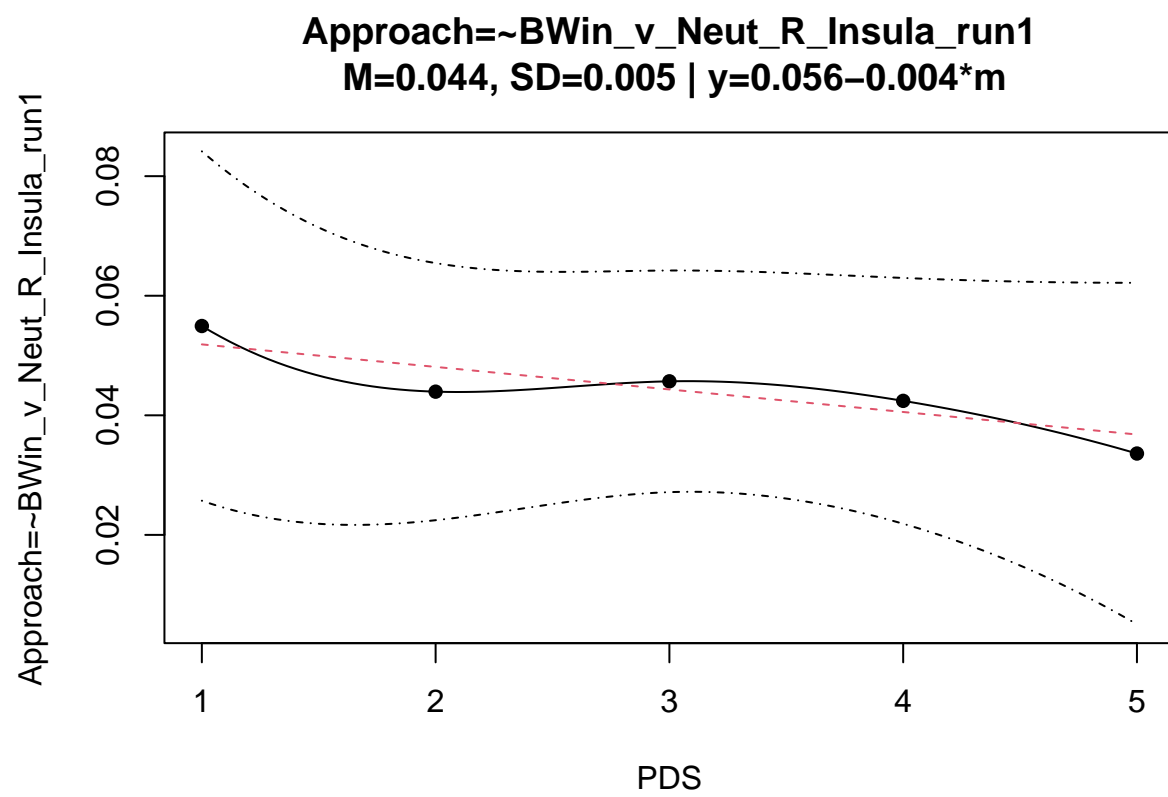


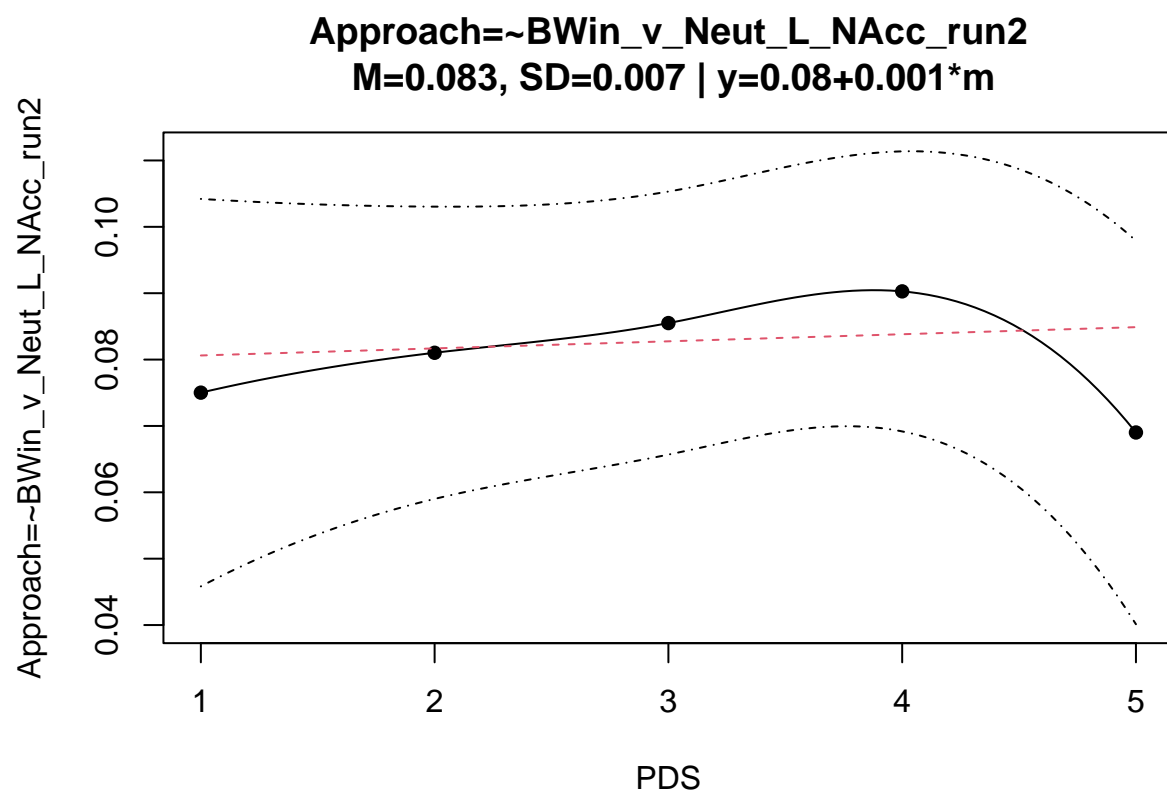


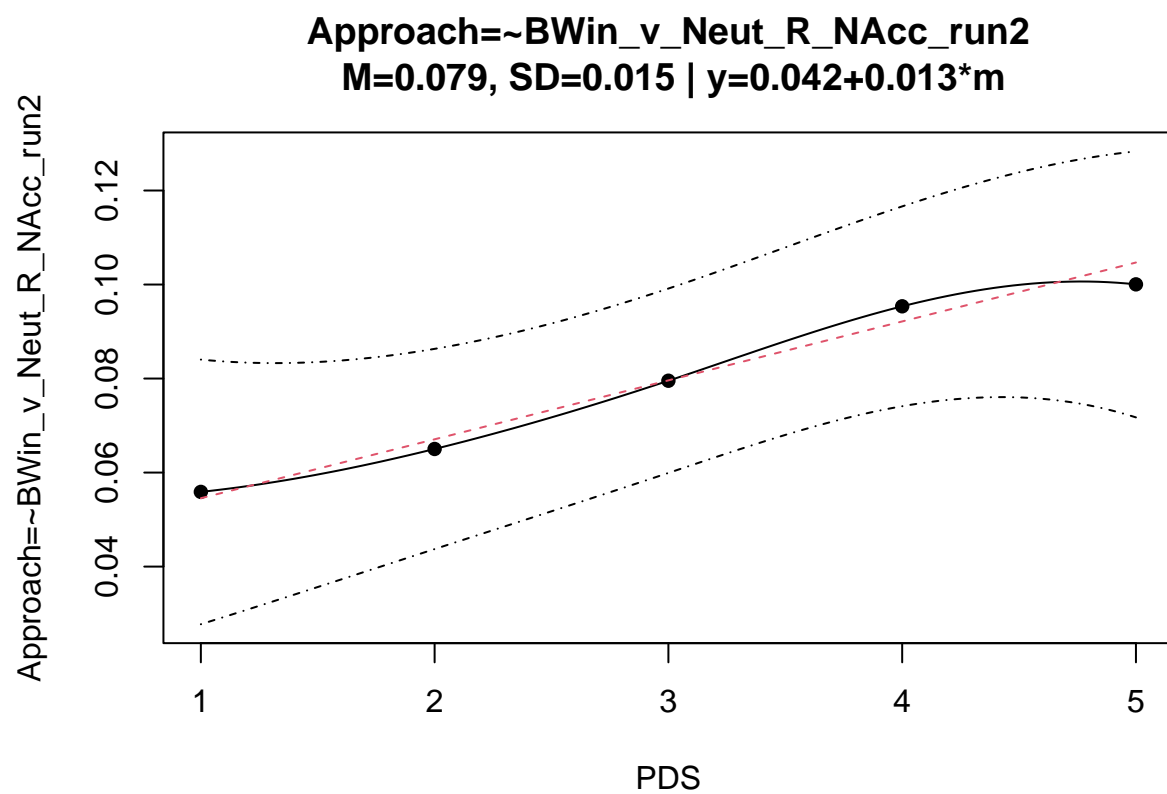


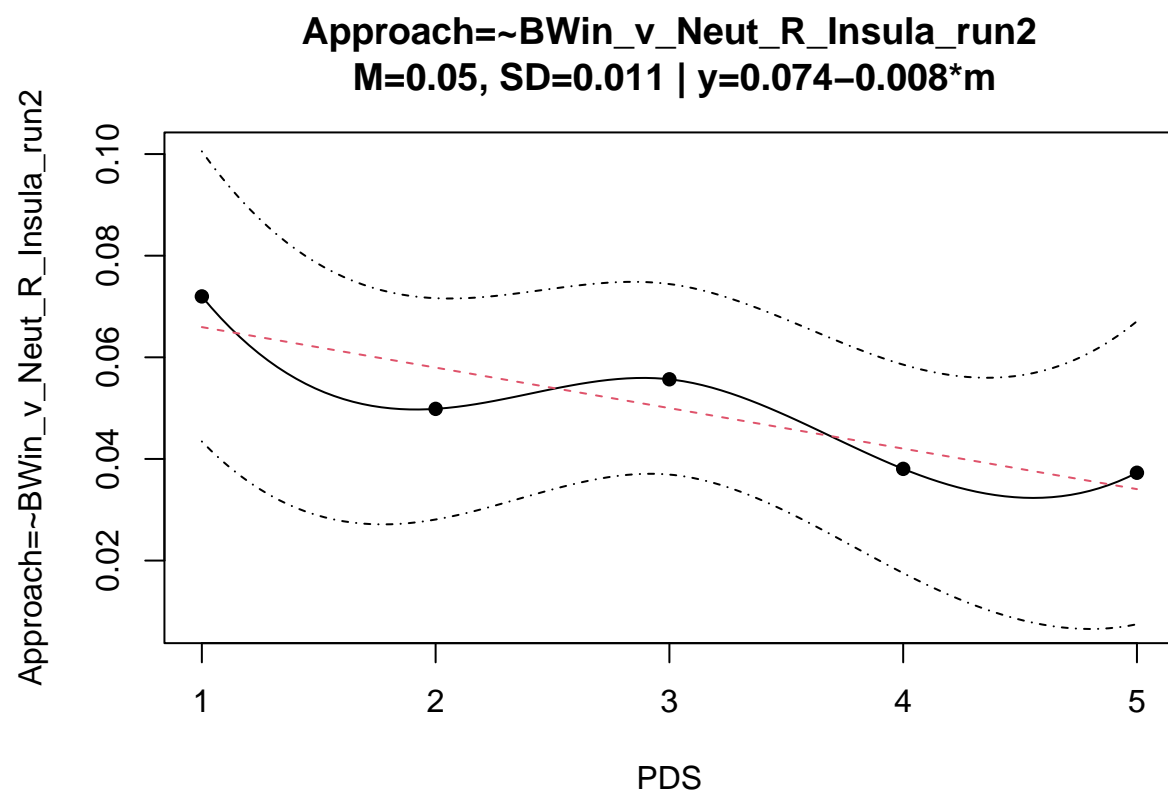


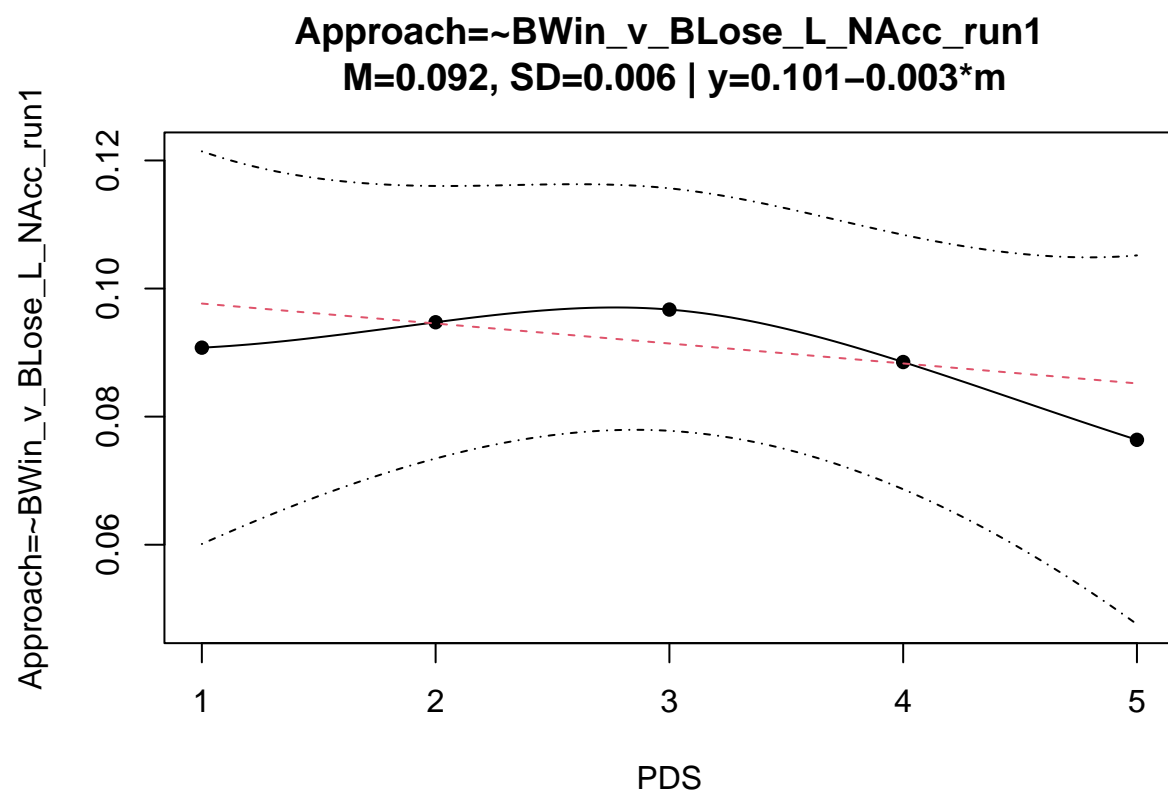


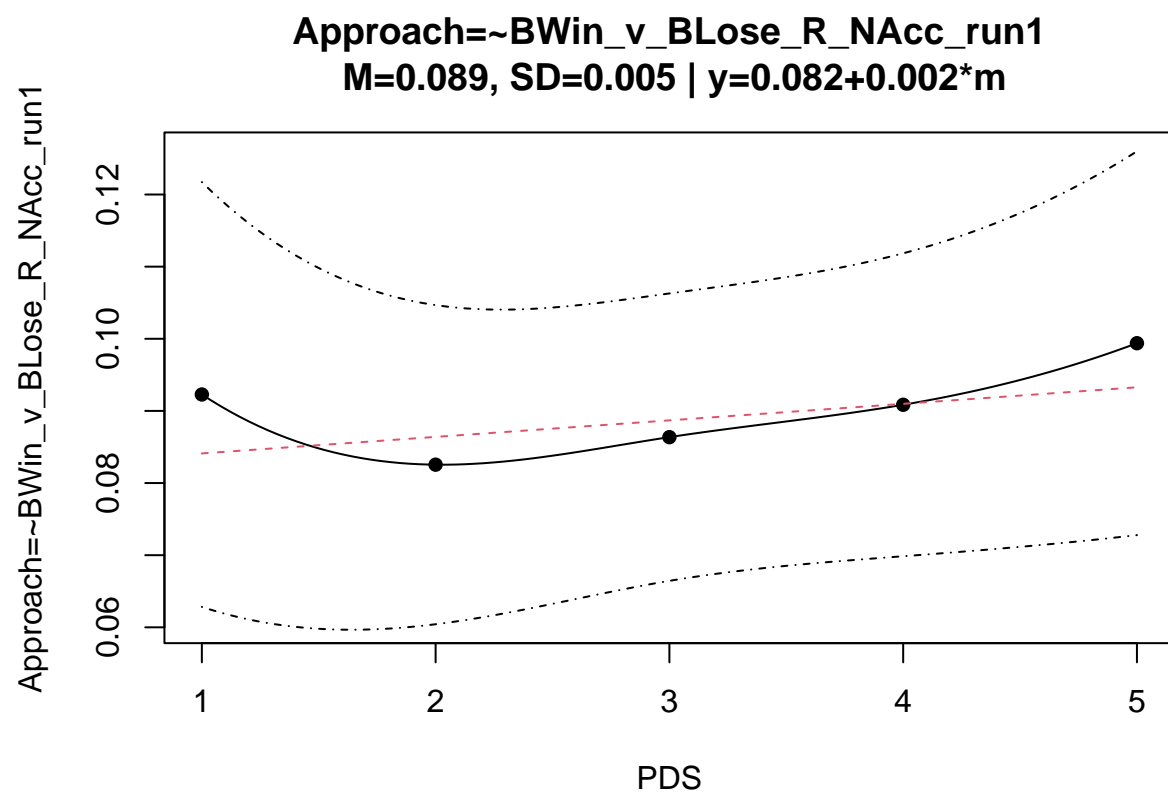


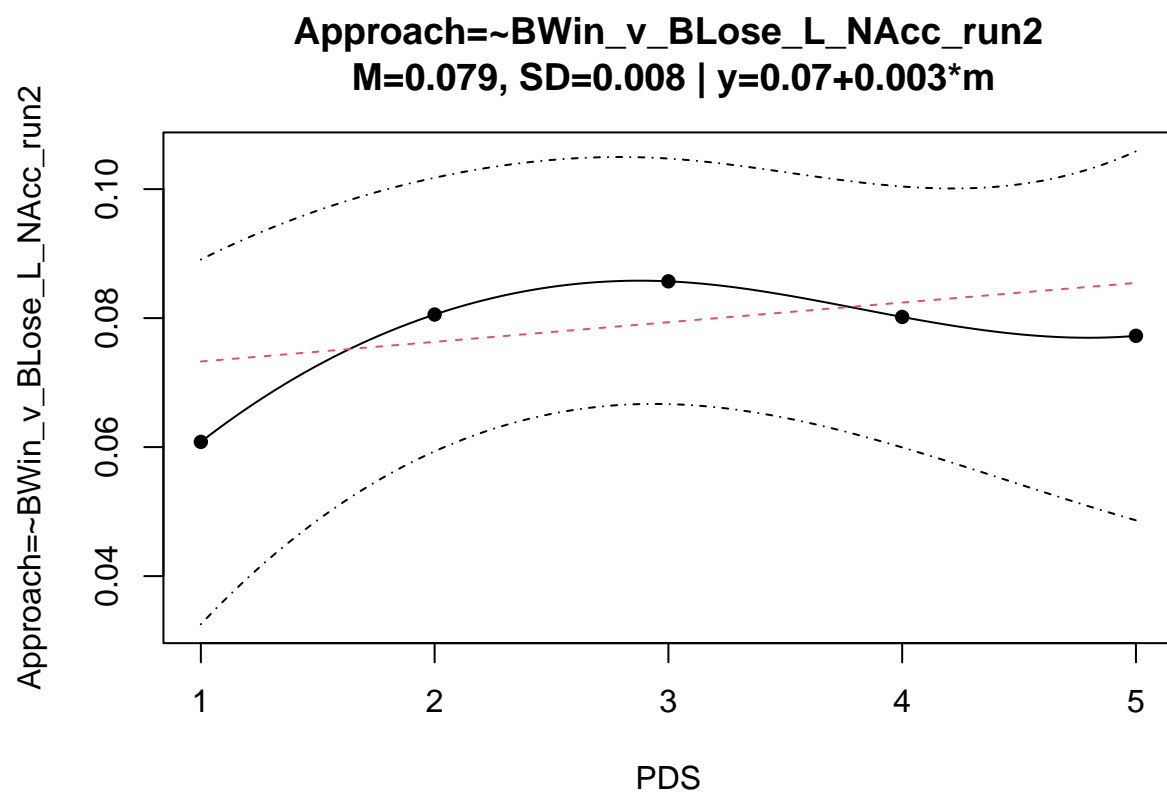


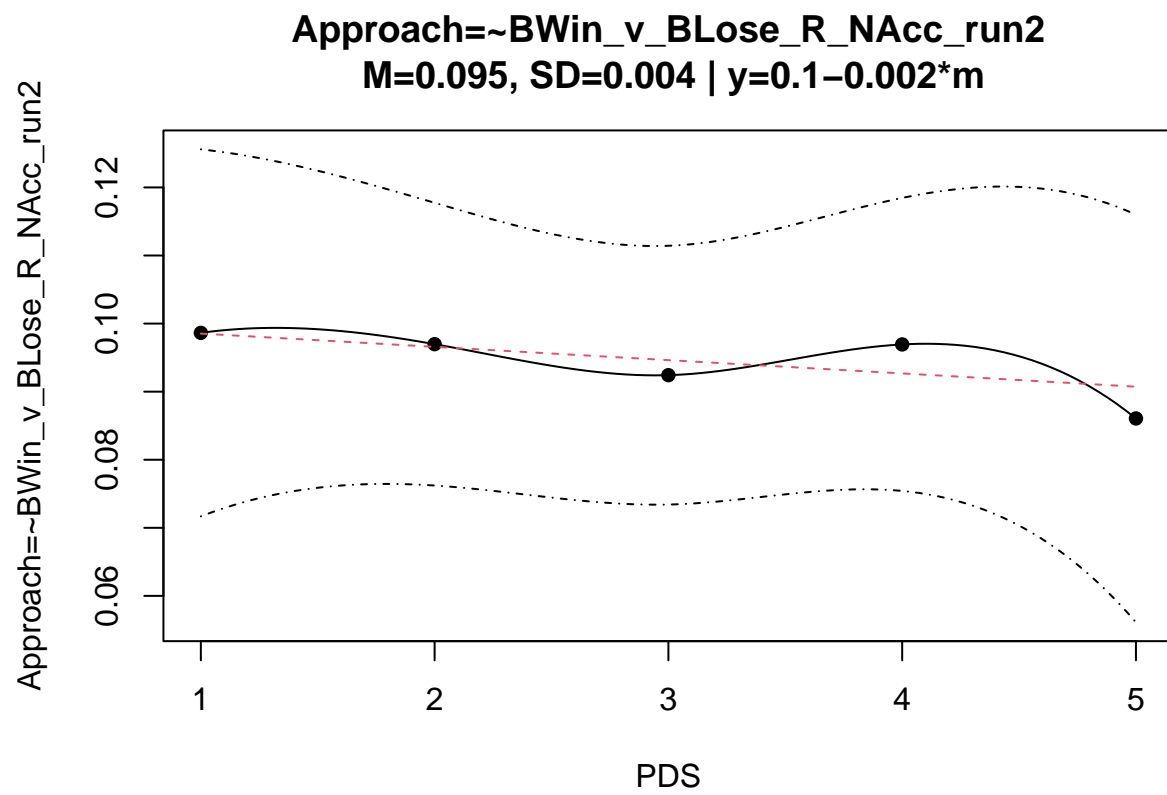




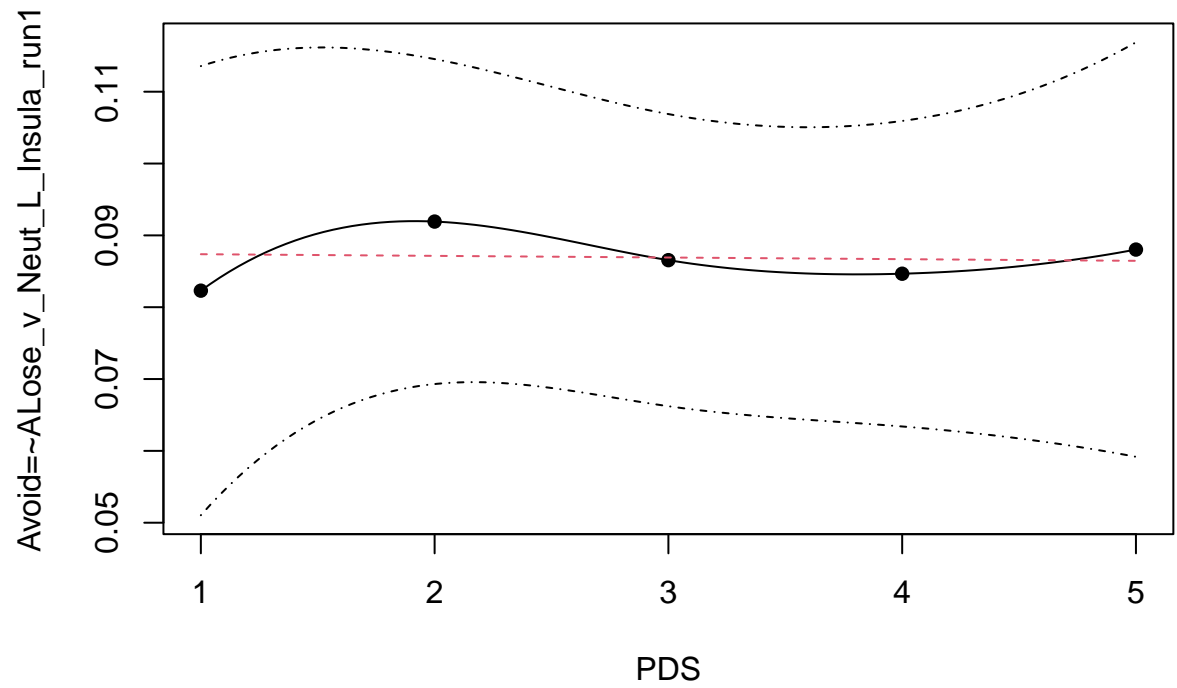




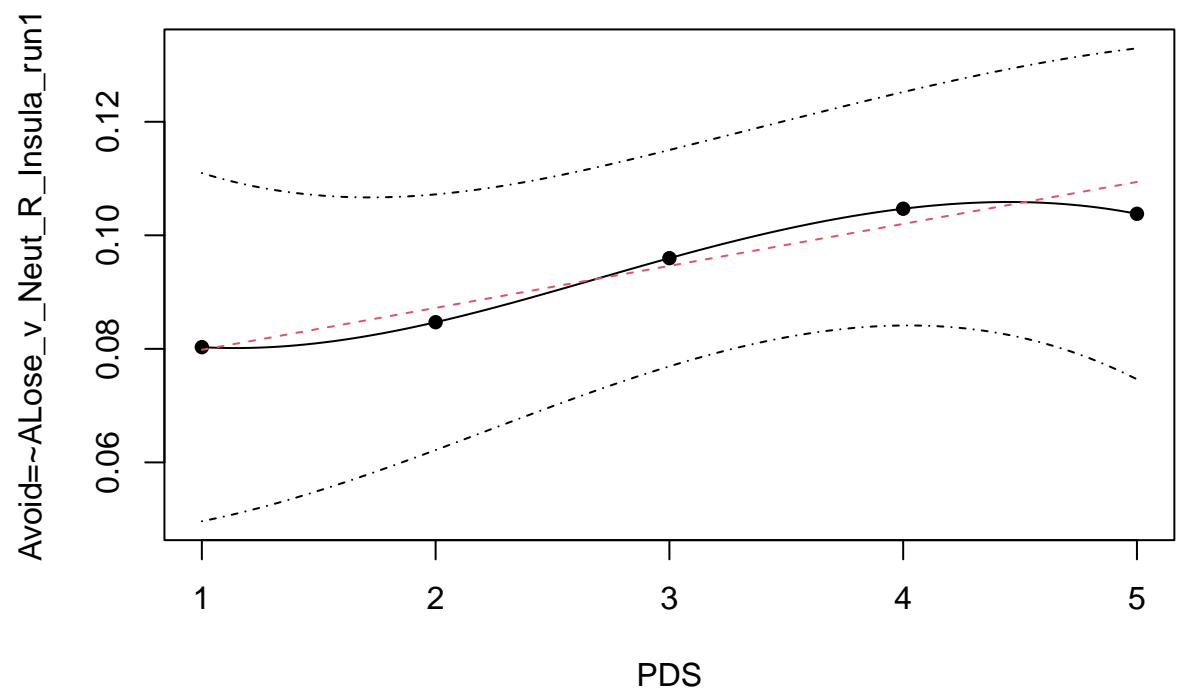


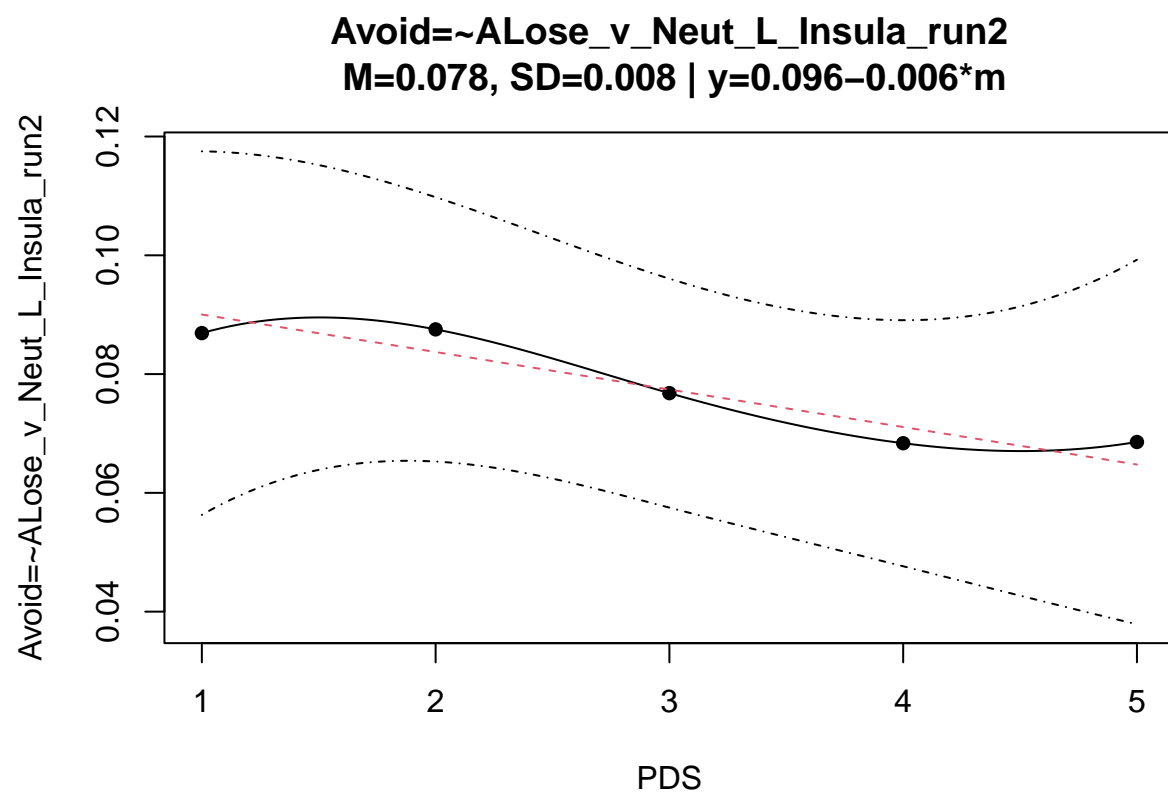


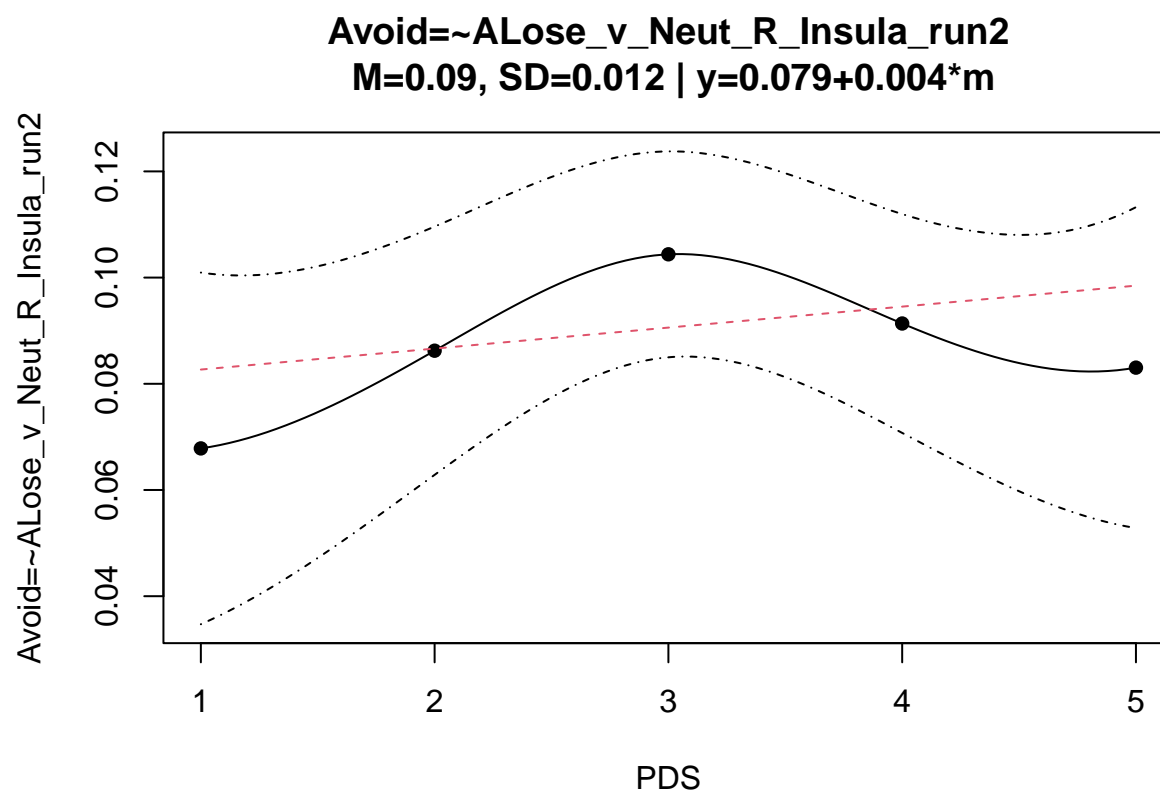
Avoid= \sim ALose_v_Neut_L_Insula_run1
M=0.087, SD=0.003 | $y=0.0880 \cdot m$



Avoid=~ALose_v_Neut_R_Insula_run1
M=0.094, SD=0.009 | $y=0.072+0.007*m$







Permutation Test LSEM

Running permutation test of LSEM model. In this case, using 10 permutation to save on time. In future iterations, permutations will be 1000.

```
lsem.permuted <- sirt::lsem.permutationTest(lsem.object = lsem.MID,
                                           B = 10, # permutations
                                           residualize = FALSE)
```

```
## Permutation test LSEM
## 1 2 3 4 5 6 7 8 9 10
```

```
summary(lsem.permuted) # examine results
```

```
## -----
## Permutation Test for Local Structural Equation Model
##
## sirt 3.12-66 (2022-05-16 12:27:54)
## lavaan 0.6-12 (2022-07-04 16:40:02 UTC)
##
## Function 'sirt::lsem.permutationTest'
##
## Call:
## sirt::lsem.permutationTest(lsem.object = lsem.MID, B = 10, residualize = FALSE)
##
## Date of Analysis: 2022-11-17 16:11:41
```

```

## Time difference of 50.20561 secs
## Computation Time: 50.20561
##
## Number of permutations = 10
## Percentage of non-converged datasets = 0
## Number of observations=1000
## Bandwidth factor=2
## Bandwidth=0.648
## Number of focal points for moderator=5
##
## lavaan Model
##
##
## # Factor loadings
## Approach =~ AWin_v_Neut_L_NAcc_run1 + AWin_v_Neut_R_NAcc_run1 + AWin_v_Neut_R_Insula_run1 +
##              AWin_v_Neut_L_NAcc_run2 + AWin_v_Neut_R_NAcc_run2 + AWin_v_Neut_R_Insula_run2 +
##              BWin_v_Neut_L_NAcc_run1 + BWin_v_Neut_R_NAcc_run1 + BWin_v_Neut_R_Insula_run1 +
##              BWin_v_Neut_L_NAcc_run2 + BWin_v_Neut_R_NAcc_run2 + BWin_v_Neut_R_Insula_run2 +
##              BWin_v_BLose_L_NAcc_run1 + BWin_v_BLose_R_NAcc_run1 +
##              BWin_v_BLose_L_NAcc_run2 + BWin_v_BLose_R_NAcc_run2
##
## Avoid =~      ALose_v_Neut_L_Insula_run1 + ALose_v_Neut_R_Insula_run1 +
##              ALose_v_Neut_L_Insula_run2 + ALose_v_Neut_R_Insula_run2 +
##              BLose_v_Neut_L_Insula_run1 + BLose_v_Neut_R_Insula_run1 +
##              BLose_v_Neut_L_Insula_run2 + BLose_v_Neut_R_Insula_run2 +
##              BLose_v_BWin_L_Insula_run1 + BLose_v_BWin_R_Insula_run1 +
##              BLose_v_BWin_L_Insula_run2 + BLose_v_BWin_R_Insula_run2
##
##
## Global Test Statistics
##
##
##              par      M    SD SD_p
## 1      Approach=~AWin_v_Neut_L_NAcc_run1  0.082 0.006 0.6
## 2      Approach=~AWin_v_Neut_R_NAcc_run1  0.086 0.004 0.9
## 3      Approach=~AWin_v_Neut_R_Insula_run1 0.041 0.004 0.8
## 4      Approach=~AWin_v_Neut_L_NAcc_run2  0.090 0.002 1.0
## 5      Approach=~AWin_v_Neut_R_NAcc_run2  0.096 0.006 0.8
## 6      Approach=~AWin_v_Neut_R_Insula_run2 0.049 0.007 0.3
## 7      Approach=~BWin_v_Neut_L_NAcc_run1  0.095 0.009 0.2
## 8      Approach=~BWin_v_Neut_R_NAcc_run1  0.070 0.007 0.2
## 9      Approach=~BWin_v_Neut_R_Insula_run1 0.044 0.005 0.7
## 10     Approach=~BWin_v_Neut_L_NAcc_run2  0.083 0.007 0.7
## 11     Approach=~BWin_v_Neut_R_NAcc_run2  0.079 0.015 0.0
## 12     Approach=~BWin_v_Neut_R_Insula_run2 0.050 0.011 0.1
## 13     Approach=~BWin_v_BLose_L_NAcc_run1 0.092 0.006 0.6
## 14     Approach=~BWin_v_BLose_R_NAcc_run1  0.089 0.005 0.8
## 15     Approach=~BWin_v_BLose_L_NAcc_run2 0.079 0.008 0.5
## 16     Approach=~BWin_v_BLose_R_NAcc_run2 0.095 0.004 1.0
## 17     Avoid=~ALose_v_Neut_L_Insula_run1  0.087 0.003 1.0
## 18     Avoid=~ALose_v_Neut_R_Insula_run1  0.094 0.009 0.4
## 19     Avoid=~ALose_v_Neut_L_Insula_run2 0.078 0.008 0.4
## 20     Avoid=~ALose_v_Neut_R_Insula_run2 0.090 0.012 0.2
## 21     Avoid=~BLose_v_Neut_L_Insula_run1  0.086 0.013 0.0
## 22     Avoid=~BLose_v_Neut_R_Insula_run1  0.092 0.006 0.7

```

## 23	Avoid=~Blose_v_Neut_L_Insula_run2	0.082	0.006	0.7
## 24	Avoid=~Blose_v_Neut_R_Insula_run2	0.091	0.005	0.8
## 25	Avoid=~Blose_v_BWin_L_Insula_run1	0.085	0.016	0.1
## 26	Avoid=~Blose_v_BWin_R_Insula_run1	0.040	0.010	0.1
## 27	Avoid=~Blose_v_BWin_L_Insula_run2	0.077	0.014	0.0
## 28	Avoid=~Blose_v_BWin_R_Insula_run2	0.041	0.004	0.7
## 29	AWin_v_Neut_L_NAcc_run1~~AWin_v_Neut_L_NAcc_run1	0.028	0.001	0.6
## 30	AWin_v_Neut_R_NAcc_run1~~AWin_v_Neut_R_NAcc_run1	0.031	0.002	0.2
## 31	AWin_v_Neut_R_Insula_run1~~AWin_v_Neut_R_Insula_run1	0.029	0.002	0.0
## 32	AWin_v_Neut_L_NAcc_run2~~AWin_v_Neut_L_NAcc_run2	0.027	0.001	0.8
## 33	AWin_v_Neut_R_NAcc_run2~~AWin_v_Neut_R_NAcc_run2	0.028	0.002	0.5
## 34	AWin_v_Neut_R_Insula_run2~~AWin_v_Neut_R_Insula_run2	0.028	0.002	0.3
## 35	BWin_v_Neut_L_NAcc_run1~~BWin_v_Neut_L_NAcc_run1	0.025	0.001	0.6
## 36	BWin_v_Neut_R_NAcc_run1~~BWin_v_Neut_R_NAcc_run1	0.029	0.001	0.8
## 37	BWin_v_Neut_R_Insula_run1~~BWin_v_Neut_R_Insula_run1	0.031	0.001	0.8
## 38	BWin_v_Neut_L_NAcc_run2~~BWin_v_Neut_L_NAcc_run2	0.030	0.001	0.7
## 39	BWin_v_Neut_R_NAcc_run2~~BWin_v_Neut_R_NAcc_run2	0.030	0.002	0.1
## 40	BWin_v_Neut_R_Insula_run2~~BWin_v_Neut_R_Insula_run2	0.031	0.001	0.8
## 41	BWin_v_BLose_L_NAcc_run1~~BWin_v_BLose_L_NAcc_run1	0.028	0.002	0.0
## 42	BWin_v_BLose_R_NAcc_run1~~BWin_v_BLose_R_NAcc_run1	0.030	0.003	0.1
## 43	BWin_v_BLose_L_NAcc_run2~~BWin_v_BLose_L_NAcc_run2	0.028	0.001	0.7
## 44	BWin_v_BLose_R_NAcc_run2~~BWin_v_BLose_R_NAcc_run2	0.028	0.002	0.0
## 45	ALose_v_Neut_L_Insula_run1~~ALose_v_Neut_L_Insula_run1	0.030	0.002	0.2
## 46	ALose_v_Neut_R_Insula_run1~~ALose_v_Neut_R_Insula_run1	0.027	0.002	0.1
## 47	ALose_v_Neut_L_Insula_run2~~ALose_v_Neut_L_Insula_run2	0.029	0.001	0.7
## 48	ALose_v_Neut_R_Insula_run2~~ALose_v_Neut_R_Insula_run2	0.029	0.003	0.0
## 49	BLose_v_Neut_L_Insula_run1~~BLose_v_Neut_L_Insula_run1	0.029	0.001	0.7
## 50	BLose_v_Neut_R_Insula_run1~~BLose_v_Neut_R_Insula_run1	0.025	0.001	0.9
## 51	BLose_v_Neut_L_Insula_run2~~BLose_v_Neut_L_Insula_run2	0.030	0.002	0.1
## 52	BLose_v_Neut_R_Insula_run2~~BLose_v_Neut_R_Insula_run2	0.029	0.002	0.5
## 53	BLose_v_BWin_L_Insula_run1~~BLose_v_BWin_L_Insula_run1	0.030	0.002	0.7
## 54	BLose_v_BWin_R_Insula_run1~~BLose_v_BWin_R_Insula_run1	0.031	0.002	0.0
## 55	BLose_v_BWin_L_Insula_run2~~BLose_v_BWin_L_Insula_run2	0.029	0.002	0.3
## 56	BLose_v_BWin_R_Insula_run2~~BLose_v_BWin_R_Insula_run2	0.029	0.001	1.0
## 57	Approach~~Approach	1.000	0.000	1.0
## 58	Avoid~~Avoid	1.000	0.000	1.0
## 59	Approach~~Avoid	-0.577	0.037	0.6
## 60	AWin_v_Neut_L_NAcc_run1~1	-0.003	0.016	0.0
## 61	AWin_v_Neut_R_NAcc_run1~1	-0.004	0.009	0.2
## 62	AWin_v_Neut_R_Insula_run1~1	0.007	0.004	0.6
## 63	AWin_v_Neut_L_NAcc_run2~1	-0.002	0.009	0.1
## 64	AWin_v_Neut_R_NAcc_run2~1	0.003	0.007	0.5
## 65	AWin_v_Neut_R_Insula_run2~1	0.005	0.003	0.8
## 66	BWin_v_Neut_L_NAcc_run1~1	-0.011	0.003	0.9
## 67	BWin_v_Neut_R_NAcc_run1~1	-0.006	0.008	0.3
## 68	BWin_v_Neut_R_Insula_run1~1	0.013	0.004	0.8
## 69	BWin_v_Neut_L_NAcc_run2~1	-0.005	0.003	0.8
## 70	BWin_v_Neut_R_NAcc_run2~1	-0.001	0.007	0.4
## 71	BWin_v_Neut_R_Insula_run2~1	-0.001	0.006	0.5
## 72	BWin_v_BLose_L_NAcc_run1~1	-0.003	0.009	0.3
## 73	BWin_v_BLose_R_NAcc_run1~1	-0.005	0.011	0.1
## 74	BWin_v_BLose_L_NAcc_run2~1	-0.008	0.010	0.0
## 75	BWin_v_BLose_R_NAcc_run2~1	-0.008	0.008	0.1
## 76	ALose_v_Neut_L_Insula_run1~1	-0.005	0.004	0.8

## 77	ALose_v_Neut_R_Insula_run1~1	0.003	0.005	0.7
## 78	ALose_v_Neut_L_Insula_run2~1	-0.001	0.005	0.8
## 79	ALose_v_Neut_R_Insula_run2~1	0.002	0.005	0.8
## 80	BLose_v_Neut_L_Insula_run1~1	0.002	0.009	0.0
## 81	BLose_v_Neut_R_Insula_run1~1	0.014	0.008	0.5
## 82	BLose_v_Neut_L_Insula_run2~1	0.013	0.003	0.9
## 83	BLose_v_Neut_R_Insula_run2~1	0.005	0.006	0.4
## 84	BLose_v_BWin_L_Insula_run1~1	0.008	0.006	0.3
## 85	BLose_v_BWin_R_Insula_run1~1	0.002	0.005	0.7
## 86	BLose_v_BWin_L_Insula_run2~1	0.005	0.004	0.7
## 87	BLose_v_BWin_R_Insula_run2~1	-0.003	0.012	0.0
## 88	Approach~1	0.000	0.000	1.0
## 89	Avoid~1	0.000	0.000	1.0
## 90	rmsea	0.075	0.005	0.7
## 91	cfi	0.598	0.039	0.4
## 92	tli	0.564	0.042	0.4
## 93	gfi	0.860	0.014	0.6
## 94	srmr	0.061	0.004	0.5
##	MAD	MAD_p	lin_slo	lin_slo_p
## 1	0.006	0.5	-0.003	0.2
## 2	0.003	0.9	0.000	0.4
## 3	0.003	0.8	-0.001	0.8
## 4	0.002	1.0	-0.002	0.6
## 5	0.003	0.8	-0.004	0.6
## 6	0.006	0.2	0.005	0.2
## 7	0.005	0.4	-0.007	0.0
## 8	0.006	0.2	0.003	0.2
## 9	0.003	0.9	-0.004	0.8
## 10	0.005	0.8	0.001	1.0
## 11	0.012	0.0	0.013	0.0
## 12	0.009	0.1	-0.008	0.2
## 13	0.005	0.5	-0.003	0.4
## 14	0.004	0.8	0.002	0.8
## 15	0.005	0.5	0.003	0.6
## 16	0.003	1.0	-0.002	1.0
## 17	0.002	1.0	0.000	0.8
## 18	0.008	0.5	0.007	0.4
## 19	0.007	0.2	-0.006	0.2
## 20	0.009	0.4	0.004	0.8
## 21	0.012	0.0	0.006	0.2
## 22	0.005	0.6	-0.004	0.4
## 23	0.005	0.6	0.003	0.6
## 24	0.004	0.8	0.001	0.8
## 25	0.015	0.0	0.013	0.0
## 26	0.009	0.1	-0.006	0.0
## 27	0.012	0.1	0.011	0.0
## 28	0.004	0.7	-0.003	0.4
## 29	0.001	0.6	0.000	0.8
## 30	0.002	0.1	0.002	0.0
## 31	0.002	0.0	0.000	0.8
## 32	0.001	0.8	0.000	0.8
## 33	0.001	0.6	0.001	0.0
## 34	0.002	0.1	-0.002	0.0
## 35	0.001	0.9	0.001	0.4

## 36	0.001	0.6	0.000	0.8
## 37	0.001	0.6	-0.001	0.2
## 38	0.001	0.7	-0.001	0.8
## 39	0.001	0.3	-0.001	0.2
## 40	0.001	0.6	0.000	1.0
## 41	0.001	0.3	-0.001	0.2
## 42	0.002	0.1	-0.002	0.0
## 43	0.001	0.6	-0.001	0.4
## 44	0.002	0.0	0.002	0.0
## 45	0.002	0.1	-0.001	0.8
## 46	0.002	0.1	-0.002	0.2
## 47	0.001	0.7	0.000	1.0
## 48	0.003	0.0	-0.002	0.0
## 49	0.001	0.7	-0.001	0.4
## 50	0.001	0.8	0.000	1.0
## 51	0.002	0.1	0.000	0.8
## 52	0.001	0.7	0.002	0.2
## 53	0.002	0.6	-0.001	0.4
## 54	0.002	0.0	-0.002	0.0
## 55	0.001	0.3	-0.001	0.0
## 56	0.001	0.9	0.000	1.0
## 57	0.000	1.0	0.000	1.0
## 58	0.000	1.0	0.000	1.0
## 59	0.033	0.6	-0.020	0.2
## 60	0.014	0.0	-0.013	0.0
## 61	0.008	0.2	-0.002	0.2
## 62	0.003	0.6	0.001	1.0
## 63	0.007	0.0	-0.007	0.0
## 64	0.005	0.4	-0.004	0.2
## 65	0.003	0.5	-0.002	0.0
## 66	0.003	0.9	0.001	0.8
## 67	0.006	0.3	-0.004	0.6
## 68	0.003	0.7	0.002	0.2
## 69	0.003	0.8	0.001	0.8
## 70	0.005	0.6	-0.004	0.0
## 71	0.004	0.6	-0.004	0.2
## 72	0.007	0.3	-0.003	0.6
## 73	0.010	0.0	-0.008	0.0
## 74	0.010	0.0	-0.005	0.2
## 75	0.008	0.0	-0.004	0.0
## 76	0.004	0.8	0.001	0.6
## 77	0.004	0.6	-0.004	0.0
## 78	0.004	0.7	0.001	0.8
## 79	0.003	0.9	-0.002	0.8
## 80	0.007	0.2	-0.003	0.4
## 81	0.007	0.3	0.007	0.0
## 82	0.003	0.8	0.001	0.6
## 83	0.006	0.1	0.002	0.2
## 84	0.005	0.3	-0.003	0.2
## 85	0.003	0.8	-0.002	0.6
## 86	0.003	0.7	-0.001	1.0
## 87	0.010	0.0	0.008	0.0
## 88	0.000	1.0	0.000	1.0
## 89	0.000	1.0	0.000	1.0


```

## 90 0.005    0.2    0.001        0.0
## 91 0.030    0.5    0.001        0.8
## 92 0.033    0.5    0.001        0.8
## 93 0.013    0.3   -0.004        0.0
## 94 0.003    0.4    0.000        0.6
##
## Pointwise Test Statistics
##
##
##                                par parindex moderator
## 1      Approach=~AWin_v_Neut_L_NAcc_run1      1      1
## 2      Approach=~AWin_v_Neut_L_NAcc_run1      1      2
## 3      Approach=~AWin_v_Neut_L_NAcc_run1      1      3
## 4      Approach=~AWin_v_Neut_L_NAcc_run1      1      4
## 5      Approach=~AWin_v_Neut_L_NAcc_run1      1      5
## 6      Approach=~AWin_v_Neut_R_NAcc_run1      2      1
## 7      Approach=~AWin_v_Neut_R_NAcc_run1      2      2
## 8      Approach=~AWin_v_Neut_R_NAcc_run1      2      3
## 9      Approach=~AWin_v_Neut_R_NAcc_run1      2      4
## 10     Approach=~AWin_v_Neut_R_NAcc_run1      2      5
## 11     Approach=~AWin_v_Neut_R_Insula_run1     3      1
## 12     Approach=~AWin_v_Neut_R_Insula_run1     3      2
## 13     Approach=~AWin_v_Neut_R_Insula_run1     3      3
## 14     Approach=~AWin_v_Neut_R_Insula_run1     3      4
## 15     Approach=~AWin_v_Neut_R_Insula_run1     3      5
## 16     Approach=~AWin_v_Neut_L_NAcc_run2       4      1
## 17     Approach=~AWin_v_Neut_L_NAcc_run2       4      2
## 18     Approach=~AWin_v_Neut_L_NAcc_run2       4      3
## 19     Approach=~AWin_v_Neut_L_NAcc_run2       4      4
## 20     Approach=~AWin_v_Neut_L_NAcc_run2       4      5
## 21     Approach=~AWin_v_Neut_R_NAcc_run2       5      1
## 22     Approach=~AWin_v_Neut_R_NAcc_run2       5      2
## 23     Approach=~AWin_v_Neut_R_NAcc_run2       5      3
## 24     Approach=~AWin_v_Neut_R_NAcc_run2       5      4
## 25     Approach=~AWin_v_Neut_R_NAcc_run2       5      5
## 26     Approach=~AWin_v_Neut_R_Insula_run2     6      1
## 27     Approach=~AWin_v_Neut_R_Insula_run2     6      2
## 28     Approach=~AWin_v_Neut_R_Insula_run2     6      3
## 29     Approach=~AWin_v_Neut_R_Insula_run2     6      4
## 30     Approach=~AWin_v_Neut_R_Insula_run2     6      5
## 31     Approach=~BWin_v_Neut_L_NAcc_run1       7      1
## 32     Approach=~BWin_v_Neut_L_NAcc_run1       7      2
## 33     Approach=~BWin_v_Neut_L_NAcc_run1       7      3
## 34     Approach=~BWin_v_Neut_L_NAcc_run1       7      4
## 35     Approach=~BWin_v_Neut_L_NAcc_run1       7      5
## 36     Approach=~BWin_v_Neut_R_NAcc_run1       8      1
## 37     Approach=~BWin_v_Neut_R_NAcc_run1       8      2
## 38     Approach=~BWin_v_Neut_R_NAcc_run1       8      3
## 39     Approach=~BWin_v_Neut_R_NAcc_run1       8      4
## 40     Approach=~BWin_v_Neut_R_NAcc_run1       8      5
## 41     Approach=~BWin_v_Neut_R_Insula_run1     9      1
## 42     Approach=~BWin_v_Neut_R_Insula_run1     9      2
## 43     Approach=~BWin_v_Neut_R_Insula_run1     9      3
## 44     Approach=~BWin_v_Neut_R_Insula_run1     9      4
## 45     Approach=~BWin_v_Neut_R_Insula_run1     9      5

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## 46	Approach=~BWin_v_Neut_L_NAcc_run2	10	1
## 47	Approach=~BWin_v_Neut_L_NAcc_run2	10	2
## 48	Approach=~BWin_v_Neut_L_NAcc_run2	10	3
## 49	Approach=~BWin_v_Neut_L_NAcc_run2	10	4
## 50	Approach=~BWin_v_Neut_L_NAcc_run2	10	5
## 51	Approach=~BWin_v_Neut_R_NAcc_run2	11	1
## 52	Approach=~BWin_v_Neut_R_NAcc_run2	11	2
## 53	Approach=~BWin_v_Neut_R_NAcc_run2	11	3
## 54	Approach=~BWin_v_Neut_R_NAcc_run2	11	4
## 55	Approach=~BWin_v_Neut_R_NAcc_run2	11	5
## 56	Approach=~BWin_v_Neut_R_Insula_run2	12	1
## 57	Approach=~BWin_v_Neut_R_Insula_run2	12	2
## 58	Approach=~BWin_v_Neut_R_Insula_run2	12	3
## 59	Approach=~BWin_v_Neut_R_Insula_run2	12	4
## 60	Approach=~BWin_v_Neut_R_Insula_run2	12	5
## 61	Approach=~BWin_v_BLose_L_NAcc_run1	13	1
## 62	Approach=~BWin_v_BLose_L_NAcc_run1	13	2
## 63	Approach=~BWin_v_BLose_L_NAcc_run1	13	3
## 64	Approach=~BWin_v_BLose_L_NAcc_run1	13	4
## 65	Approach=~BWin_v_BLose_L_NAcc_run1	13	5
## 66	Approach=~BWin_v_BLose_R_NAcc_run1	14	1
## 67	Approach=~BWin_v_BLose_R_NAcc_run1	14	2
## 68	Approach=~BWin_v_BLose_R_NAcc_run1	14	3
## 69	Approach=~BWin_v_BLose_R_NAcc_run1	14	4
## 70	Approach=~BWin_v_BLose_R_NAcc_run1	14	5
## 71	Approach=~BWin_v_BLose_L_NAcc_run2	15	1
## 72	Approach=~BWin_v_BLose_L_NAcc_run2	15	2
## 73	Approach=~BWin_v_BLose_L_NAcc_run2	15	3
## 74	Approach=~BWin_v_BLose_L_NAcc_run2	15	4
## 75	Approach=~BWin_v_BLose_L_NAcc_run2	15	5
## 76	Approach=~BWin_v_BLose_R_NAcc_run2	16	1
## 77	Approach=~BWin_v_BLose_R_NAcc_run2	16	2
## 78	Approach=~BWin_v_BLose_R_NAcc_run2	16	3
## 79	Approach=~BWin_v_BLose_R_NAcc_run2	16	4
## 80	Approach=~BWin_v_BLose_R_NAcc_run2	16	5
## 81	Avoid=~ALose_v_Neut_L_Insula_run1	17	1
## 82	Avoid=~ALose_v_Neut_L_Insula_run1	17	2
## 83	Avoid=~ALose_v_Neut_L_Insula_run1	17	3
## 84	Avoid=~ALose_v_Neut_L_Insula_run1	17	4
## 85	Avoid=~ALose_v_Neut_L_Insula_run1	17	5
## 86	Avoid=~ALose_v_Neut_R_Insula_run1	18	1
## 87	Avoid=~ALose_v_Neut_R_Insula_run1	18	2
## 88	Avoid=~ALose_v_Neut_R_Insula_run1	18	3
## 89	Avoid=~ALose_v_Neut_R_Insula_run1	18	4
## 90	Avoid=~ALose_v_Neut_R_Insula_run1	18	5
## 91	Avoid=~ALose_v_Neut_L_Insula_run2	19	1
## 92	Avoid=~ALose_v_Neut_L_Insula_run2	19	2
## 93	Avoid=~ALose_v_Neut_L_Insula_run2	19	3
## 94	Avoid=~ALose_v_Neut_L_Insula_run2	19	4
## 95	Avoid=~ALose_v_Neut_L_Insula_run2	19	5
## 96	Avoid=~ALose_v_Neut_R_Insula_run2	20	1
## 97	Avoid=~ALose_v_Neut_R_Insula_run2	20	2
## 98	Avoid=~ALose_v_Neut_R_Insula_run2	20	3
## 99	Avoid=~ALose_v_Neut_R_Insula_run2	20	4

## 100	Avoid=~ALose_v_Neut_R_Insula_run2	20	5
## 101	Avoid=~BLose_v_Neut_L_Insula_run1	21	1
## 102	Avoid=~BLose_v_Neut_L_Insula_run1	21	2
## 103	Avoid=~BLose_v_Neut_L_Insula_run1	21	3
## 104	Avoid=~BLose_v_Neut_L_Insula_run1	21	4
## 105	Avoid=~BLose_v_Neut_L_Insula_run1	21	5
## 106	Avoid=~BLose_v_Neut_R_Insula_run1	22	1
## 107	Avoid=~BLose_v_Neut_R_Insula_run1	22	2
## 108	Avoid=~BLose_v_Neut_R_Insula_run1	22	3
## 109	Avoid=~BLose_v_Neut_R_Insula_run1	22	4
## 110	Avoid=~BLose_v_Neut_R_Insula_run1	22	5
## 111	Avoid=~BLose_v_Neut_L_Insula_run2	23	1
## 112	Avoid=~BLose_v_Neut_L_Insula_run2	23	2
## 113	Avoid=~BLose_v_Neut_L_Insula_run2	23	3
## 114	Avoid=~BLose_v_Neut_L_Insula_run2	23	4
## 115	Avoid=~BLose_v_Neut_L_Insula_run2	23	5
## 116	Avoid=~BLose_v_Neut_R_Insula_run2	24	1
## 117	Avoid=~BLose_v_Neut_R_Insula_run2	24	2
## 118	Avoid=~BLose_v_Neut_R_Insula_run2	24	3
## 119	Avoid=~BLose_v_Neut_R_Insula_run2	24	4
## 120	Avoid=~BLose_v_Neut_R_Insula_run2	24	5
## 121	Avoid=~BLose_v_BWin_L_Insula_run1	25	1
## 122	Avoid=~BLose_v_BWin_L_Insula_run1	25	2
## 123	Avoid=~BLose_v_BWin_L_Insula_run1	25	3
## 124	Avoid=~BLose_v_BWin_L_Insula_run1	25	4
## 125	Avoid=~BLose_v_BWin_L_Insula_run1	25	5
## 126	Avoid=~BLose_v_BWin_R_Insula_run1	26	1
## 127	Avoid=~BLose_v_BWin_R_Insula_run1	26	2
## 128	Avoid=~BLose_v_BWin_R_Insula_run1	26	3
## 129	Avoid=~BLose_v_BWin_R_Insula_run1	26	4
## 130	Avoid=~BLose_v_BWin_R_Insula_run1	26	5
## 131	Avoid=~BLose_v_BWin_L_Insula_run2	27	1
## 132	Avoid=~BLose_v_BWin_L_Insula_run2	27	2
## 133	Avoid=~BLose_v_BWin_L_Insula_run2	27	3
## 134	Avoid=~BLose_v_BWin_L_Insula_run2	27	4
## 135	Avoid=~BLose_v_BWin_L_Insula_run2	27	5
## 136	Avoid=~BLose_v_BWin_R_Insula_run2	28	1
## 137	Avoid=~BLose_v_BWin_R_Insula_run2	28	2
## 138	Avoid=~BLose_v_BWin_R_Insula_run2	28	3
## 139	Avoid=~BLose_v_BWin_R_Insula_run2	28	4
## 140	Avoid=~BLose_v_BWin_R_Insula_run2	28	5
## 141	AWin_v_Neut_L_NAcc_run1~~AWin_v_Neut_L_NAcc_run1	29	1
## 142	AWin_v_Neut_L_NAcc_run1~~AWin_v_Neut_L_NAcc_run1	29	2
## 143	AWin_v_Neut_L_NAcc_run1~~AWin_v_Neut_L_NAcc_run1	29	3
## 144	AWin_v_Neut_L_NAcc_run1~~AWin_v_Neut_L_NAcc_run1	29	4
## 145	AWin_v_Neut_L_NAcc_run1~~AWin_v_Neut_L_NAcc_run1	29	5
## 146	AWin_v_Neut_R_NAcc_run1~~AWin_v_Neut_R_NAcc_run1	30	1
## 147	AWin_v_Neut_R_NAcc_run1~~AWin_v_Neut_R_NAcc_run1	30	2
## 148	AWin_v_Neut_R_NAcc_run1~~AWin_v_Neut_R_NAcc_run1	30	3
## 149	AWin_v_Neut_R_NAcc_run1~~AWin_v_Neut_R_NAcc_run1	30	4
## 150	AWin_v_Neut_R_NAcc_run1~~AWin_v_Neut_R_NAcc_run1	30	5
## 151	AWin_v_Neut_R_Insula_run1~~AWin_v_Neut_R_Insula_run1	31	1
## 152	AWin_v_Neut_R_Insula_run1~~AWin_v_Neut_R_Insula_run1	31	2
## 153	AWin_v_Neut_R_Insula_run1~~AWin_v_Neut_R_Insula_run1	31	3

## 154	AWin_v_Neut_R_Insula_run1~~AWin_v_Neut_R_Insula_run1	31	4
## 155	AWin_v_Neut_R_Insula_run1~~AWin_v_Neut_R_Insula_run1	31	5
## 156	AWin_v_Neut_L_NAcc_run2~~AWin_v_Neut_L_NAcc_run2	32	1
## 157	AWin_v_Neut_L_NAcc_run2~~AWin_v_Neut_L_NAcc_run2	32	2
## 158	AWin_v_Neut_L_NAcc_run2~~AWin_v_Neut_L_NAcc_run2	32	3
## 159	AWin_v_Neut_L_NAcc_run2~~AWin_v_Neut_L_NAcc_run2	32	4
## 160	AWin_v_Neut_L_NAcc_run2~~AWin_v_Neut_L_NAcc_run2	32	5
## 161	AWin_v_Neut_R_NAcc_run2~~AWin_v_Neut_R_NAcc_run2	33	1
## 162	AWin_v_Neut_R_NAcc_run2~~AWin_v_Neut_R_NAcc_run2	33	2
## 163	AWin_v_Neut_R_NAcc_run2~~AWin_v_Neut_R_NAcc_run2	33	3
## 164	AWin_v_Neut_R_NAcc_run2~~AWin_v_Neut_R_NAcc_run2	33	4
## 165	AWin_v_Neut_R_NAcc_run2~~AWin_v_Neut_R_NAcc_run2	33	5
## 166	AWin_v_Neut_R_Insula_run2~~AWin_v_Neut_R_Insula_run2	34	1
## 167	AWin_v_Neut_R_Insula_run2~~AWin_v_Neut_R_Insula_run2	34	2
## 168	AWin_v_Neut_R_Insula_run2~~AWin_v_Neut_R_Insula_run2	34	3
## 169	AWin_v_Neut_R_Insula_run2~~AWin_v_Neut_R_Insula_run2	34	4
## 170	AWin_v_Neut_R_Insula_run2~~AWin_v_Neut_R_Insula_run2	34	5
## 171	BWin_v_Neut_L_NAcc_run1~~BWin_v_Neut_L_NAcc_run1	35	1
## 172	BWin_v_Neut_L_NAcc_run1~~BWin_v_Neut_L_NAcc_run1	35	2
## 173	BWin_v_Neut_L_NAcc_run1~~BWin_v_Neut_L_NAcc_run1	35	3
## 174	BWin_v_Neut_L_NAcc_run1~~BWin_v_Neut_L_NAcc_run1	35	4
## 175	BWin_v_Neut_L_NAcc_run1~~BWin_v_Neut_L_NAcc_run1	35	5
## 176	BWin_v_Neut_R_NAcc_run1~~BWin_v_Neut_R_NAcc_run1	36	1
## 177	BWin_v_Neut_R_NAcc_run1~~BWin_v_Neut_R_NAcc_run1	36	2
## 178	BWin_v_Neut_R_NAcc_run1~~BWin_v_Neut_R_NAcc_run1	36	3
## 179	BWin_v_Neut_R_NAcc_run1~~BWin_v_Neut_R_NAcc_run1	36	4
## 180	BWin_v_Neut_R_NAcc_run1~~BWin_v_Neut_R_NAcc_run1	36	5
## 181	BWin_v_Neut_R_Insula_run1~~BWin_v_Neut_R_Insula_run1	37	1
## 182	BWin_v_Neut_R_Insula_run1~~BWin_v_Neut_R_Insula_run1	37	2
## 183	BWin_v_Neut_R_Insula_run1~~BWin_v_Neut_R_Insula_run1	37	3
## 184	BWin_v_Neut_R_Insula_run1~~BWin_v_Neut_R_Insula_run1	37	4
## 185	BWin_v_Neut_R_Insula_run1~~BWin_v_Neut_R_Insula_run1	37	5
## 186	BWin_v_Neut_L_NAcc_run2~~BWin_v_Neut_L_NAcc_run2	38	1
## 187	BWin_v_Neut_L_NAcc_run2~~BWin_v_Neut_L_NAcc_run2	38	2
## 188	BWin_v_Neut_L_NAcc_run2~~BWin_v_Neut_L_NAcc_run2	38	3
## 189	BWin_v_Neut_L_NAcc_run2~~BWin_v_Neut_L_NAcc_run2	38	4
## 190	BWin_v_Neut_L_NAcc_run2~~BWin_v_Neut_L_NAcc_run2	38	5
## 191	BWin_v_Neut_R_NAcc_run2~~BWin_v_Neut_R_NAcc_run2	39	1
## 192	BWin_v_Neut_R_NAcc_run2~~BWin_v_Neut_R_NAcc_run2	39	2
## 193	BWin_v_Neut_R_NAcc_run2~~BWin_v_Neut_R_NAcc_run2	39	3
## 194	BWin_v_Neut_R_NAcc_run2~~BWin_v_Neut_R_NAcc_run2	39	4
## 195	BWin_v_Neut_R_NAcc_run2~~BWin_v_Neut_R_NAcc_run2	39	5
## 196	BWin_v_Neut_R_Insula_run2~~BWin_v_Neut_R_Insula_run2	40	1
## 197	BWin_v_Neut_R_Insula_run2~~BWin_v_Neut_R_Insula_run2	40	2
## 198	BWin_v_Neut_R_Insula_run2~~BWin_v_Neut_R_Insula_run2	40	3
## 199	BWin_v_Neut_R_Insula_run2~~BWin_v_Neut_R_Insula_run2	40	4
## 200	BWin_v_Neut_R_Insula_run2~~BWin_v_Neut_R_Insula_run2	40	5
## 201	BWin_v_BLose_L_NAcc_run1~~BWin_v_BLose_L_NAcc_run1	41	1
## 202	BWin_v_BLose_L_NAcc_run1~~BWin_v_BLose_L_NAcc_run1	41	2
## 203	BWin_v_BLose_L_NAcc_run1~~BWin_v_BLose_L_NAcc_run1	41	3
## 204	BWin_v_BLose_L_NAcc_run1~~BWin_v_BLose_L_NAcc_run1	41	4
## 205	BWin_v_BLose_L_NAcc_run1~~BWin_v_BLose_L_NAcc_run1	41	5
## 206	BWin_v_BLose_R_NAcc_run1~~BWin_v_BLose_R_NAcc_run1	42	1
## 207	BWin_v_BLose_R_NAcc_run1~~BWin_v_BLose_R_NAcc_run1	42	2

## 208	BWin_v_BLose_R_NAcc_run1~~BWin_v_BLose_R_NAcc_run1	42	3
## 209	BWin_v_BLose_R_NAcc_run1~~BWin_v_BLose_R_NAcc_run1	42	4
## 210	BWin_v_BLose_R_NAcc_run1~~BWin_v_BLose_R_NAcc_run1	42	5
## 211	BWin_v_BLose_L_NAcc_run2~~BWin_v_BLose_L_NAcc_run2	43	1
## 212	BWin_v_BLose_L_NAcc_run2~~BWin_v_BLose_L_NAcc_run2	43	2
## 213	BWin_v_BLose_L_NAcc_run2~~BWin_v_BLose_L_NAcc_run2	43	3
## 214	BWin_v_BLose_L_NAcc_run2~~BWin_v_BLose_L_NAcc_run2	43	4
## 215	BWin_v_BLose_L_NAcc_run2~~BWin_v_BLose_L_NAcc_run2	43	5
## 216	BWin_v_BLose_R_NAcc_run2~~BWin_v_BLose_R_NAcc_run2	44	1
## 217	BWin_v_BLose_R_NAcc_run2~~BWin_v_BLose_R_NAcc_run2	44	2
## 218	BWin_v_BLose_R_NAcc_run2~~BWin_v_BLose_R_NAcc_run2	44	3
## 219	BWin_v_BLose_R_NAcc_run2~~BWin_v_BLose_R_NAcc_run2	44	4
## 220	BWin_v_BLose_R_NAcc_run2~~BWin_v_BLose_R_NAcc_run2	44	5
## 221	ALose_v_Neut_L_Insula_run1~~ALose_v_Neut_L_Insula_run1	45	1
## 222	ALose_v_Neut_L_Insula_run1~~ALose_v_Neut_L_Insula_run1	45	2
## 223	ALose_v_Neut_L_Insula_run1~~ALose_v_Neut_L_Insula_run1	45	3
## 224	ALose_v_Neut_L_Insula_run1~~ALose_v_Neut_L_Insula_run1	45	4
## 225	ALose_v_Neut_L_Insula_run1~~ALose_v_Neut_L_Insula_run1	45	5
## 226	ALose_v_Neut_R_Insula_run1~~ALose_v_Neut_R_Insula_run1	46	1
## 227	ALose_v_Neut_R_Insula_run1~~ALose_v_Neut_R_Insula_run1	46	2
## 228	ALose_v_Neut_R_Insula_run1~~ALose_v_Neut_R_Insula_run1	46	3
## 229	ALose_v_Neut_R_Insula_run1~~ALose_v_Neut_R_Insula_run1	46	4
## 230	ALose_v_Neut_R_Insula_run1~~ALose_v_Neut_R_Insula_run1	46	5
## 231	ALose_v_Neut_L_Insula_run2~~ALose_v_Neut_L_Insula_run2	47	1
## 232	ALose_v_Neut_L_Insula_run2~~ALose_v_Neut_L_Insula_run2	47	2
## 233	ALose_v_Neut_L_Insula_run2~~ALose_v_Neut_L_Insula_run2	47	3
## 234	ALose_v_Neut_L_Insula_run2~~ALose_v_Neut_L_Insula_run2	47	4
## 235	ALose_v_Neut_L_Insula_run2~~ALose_v_Neut_L_Insula_run2	47	5
## 236	ALose_v_Neut_R_Insula_run2~~ALose_v_Neut_R_Insula_run2	48	1
## 237	ALose_v_Neut_R_Insula_run2~~ALose_v_Neut_R_Insula_run2	48	2
## 238	ALose_v_Neut_R_Insula_run2~~ALose_v_Neut_R_Insula_run2	48	3
## 239	ALose_v_Neut_R_Insula_run2~~ALose_v_Neut_R_Insula_run2	48	4
## 240	ALose_v_Neut_R_Insula_run2~~ALose_v_Neut_R_Insula_run2	48	5
## 241	BLose_v_Neut_L_Insula_run1~~BLose_v_Neut_L_Insula_run1	49	1
## 242	BLose_v_Neut_L_Insula_run1~~BLose_v_Neut_L_Insula_run1	49	2
## 243	BLose_v_Neut_L_Insula_run1~~BLose_v_Neut_L_Insula_run1	49	3
## 244	BLose_v_Neut_L_Insula_run1~~BLose_v_Neut_L_Insula_run1	49	4
## 245	BLose_v_Neut_L_Insula_run1~~BLose_v_Neut_L_Insula_run1	49	5
## 246	BLose_v_Neut_R_Insula_run1~~BLose_v_Neut_R_Insula_run1	50	1
## 247	BLose_v_Neut_R_Insula_run1~~BLose_v_Neut_R_Insula_run1	50	2
## 248	BLose_v_Neut_R_Insula_run1~~BLose_v_Neut_R_Insula_run1	50	3
## 249	BLose_v_Neut_R_Insula_run1~~BLose_v_Neut_R_Insula_run1	50	4
## 250	BLose_v_Neut_R_Insula_run1~~BLose_v_Neut_R_Insula_run1	50	5
## 251	BLose_v_Neut_L_Insula_run2~~BLose_v_Neut_L_Insula_run2	51	1
## 252	BLose_v_Neut_L_Insula_run2~~BLose_v_Neut_L_Insula_run2	51	2
## 253	BLose_v_Neut_L_Insula_run2~~BLose_v_Neut_L_Insula_run2	51	3
## 254	BLose_v_Neut_L_Insula_run2~~BLose_v_Neut_L_Insula_run2	51	4
## 255	BLose_v_Neut_L_Insula_run2~~BLose_v_Neut_L_Insula_run2	51	5
## 256	BLose_v_Neut_R_Insula_run2~~BLose_v_Neut_R_Insula_run2	52	1
## 257	BLose_v_Neut_R_Insula_run2~~BLose_v_Neut_R_Insula_run2	52	2
## 258	BLose_v_Neut_R_Insula_run2~~BLose_v_Neut_R_Insula_run2	52	3
## 259	BLose_v_Neut_R_Insula_run2~~BLose_v_Neut_R_Insula_run2	52	4
## 260	BLose_v_Neut_R_Insula_run2~~BLose_v_Neut_R_Insula_run2	52	5
## 261	BLose_v_BWin_L_Insula_run1~~BLose_v_BWin_L_Insula_run1	53	1

## 262	BLOSE_v_BWin_L_Insula_run1~~BLOSE_v_BWin_L_Insula_run1	53	2
## 263	BLOSE_v_BWin_L_Insula_run1~~BLOSE_v_BWin_L_Insula_run1	53	3
## 264	BLOSE_v_BWin_L_Insula_run1~~BLOSE_v_BWin_L_Insula_run1	53	4
## 265	BLOSE_v_BWin_L_Insula_run1~~BLOSE_v_BWin_L_Insula_run1	53	5
## 266	BLOSE_v_BWin_R_Insula_run1~~BLOSE_v_BWin_R_Insula_run1	54	1
## 267	BLOSE_v_BWin_R_Insula_run1~~BLOSE_v_BWin_R_Insula_run1	54	2
## 268	BLOSE_v_BWin_R_Insula_run1~~BLOSE_v_BWin_R_Insula_run1	54	3
## 269	BLOSE_v_BWin_R_Insula_run1~~BLOSE_v_BWin_R_Insula_run1	54	4
## 270	BLOSE_v_BWin_R_Insula_run1~~BLOSE_v_BWin_R_Insula_run1	54	5
## 271	BLOSE_v_BWin_L_Insula_run2~~BLOSE_v_BWin_L_Insula_run2	55	1
## 272	BLOSE_v_BWin_L_Insula_run2~~BLOSE_v_BWin_L_Insula_run2	55	2
## 273	BLOSE_v_BWin_L_Insula_run2~~BLOSE_v_BWin_L_Insula_run2	55	3
## 274	BLOSE_v_BWin_L_Insula_run2~~BLOSE_v_BWin_L_Insula_run2	55	4
## 275	BLOSE_v_BWin_L_Insula_run2~~BLOSE_v_BWin_L_Insula_run2	55	5
## 276	BLOSE_v_BWin_R_Insula_run2~~BLOSE_v_BWin_R_Insula_run2	56	1
## 277	BLOSE_v_BWin_R_Insula_run2~~BLOSE_v_BWin_R_Insula_run2	56	2
## 278	BLOSE_v_BWin_R_Insula_run2~~BLOSE_v_BWin_R_Insula_run2	56	3
## 279	BLOSE_v_BWin_R_Insula_run2~~BLOSE_v_BWin_R_Insula_run2	56	4
## 280	BLOSE_v_BWin_R_Insula_run2~~BLOSE_v_BWin_R_Insula_run2	56	5
## 281	Approach~~Approach	57	1
## 282	Approach~~Approach	57	2
## 283	Approach~~Approach	57	3
## 284	Approach~~Approach	57	4
## 285	Approach~~Approach	57	5
## 286	Avoid~~Avoid	58	1
## 287	Avoid~~Avoid	58	2
## 288	Avoid~~Avoid	58	3
## 289	Avoid~~Avoid	58	4
## 290	Avoid~~Avoid	58	5
## 291	Approach~~Avoid	59	1
## 292	Approach~~Avoid	59	2
## 293	Approach~~Avoid	59	3
## 294	Approach~~Avoid	59	4
## 295	Approach~~Avoid	59	5
## 296	AWin_v_Neut_L_NAcc_run1~1	60	1
## 297	AWin_v_Neut_L_NAcc_run1~1	60	2
## 298	AWin_v_Neut_L_NAcc_run1~1	60	3
## 299	AWin_v_Neut_L_NAcc_run1~1	60	4
## 300	AWin_v_Neut_L_NAcc_run1~1	60	5
## 301	AWin_v_Neut_R_NAcc_run1~1	61	1
## 302	AWin_v_Neut_R_NAcc_run1~1	61	2
## 303	AWin_v_Neut_R_NAcc_run1~1	61	3
## 304	AWin_v_Neut_R_NAcc_run1~1	61	4
## 305	AWin_v_Neut_R_NAcc_run1~1	61	5
## 306	AWin_v_Neut_R_Insula_run1~1	62	1
## 307	AWin_v_Neut_R_Insula_run1~1	62	2
## 308	AWin_v_Neut_R_Insula_run1~1	62	3
## 309	AWin_v_Neut_R_Insula_run1~1	62	4
## 310	AWin_v_Neut_R_Insula_run1~1	62	5
## 311	AWin_v_Neut_L_NAcc_run2~1	63	1
## 312	AWin_v_Neut_L_NAcc_run2~1	63	2
## 313	AWin_v_Neut_L_NAcc_run2~1	63	3
## 314	AWin_v_Neut_L_NAcc_run2~1	63	4
## 315	AWin_v_Neut_L_NAcc_run2~1	63	5

## 316	AWin_v_Neut_R_NAcc_run2~1	64	1
## 317	AWin_v_Neut_R_NAcc_run2~1	64	2
## 318	AWin_v_Neut_R_NAcc_run2~1	64	3
## 319	AWin_v_Neut_R_NAcc_run2~1	64	4
## 320	AWin_v_Neut_R_NAcc_run2~1	64	5
## 321	AWin_v_Neut_R_Insula_run2~1	65	1
## 322	AWin_v_Neut_R_Insula_run2~1	65	2
## 323	AWin_v_Neut_R_Insula_run2~1	65	3
## 324	AWin_v_Neut_R_Insula_run2~1	65	4
## 325	AWin_v_Neut_R_Insula_run2~1	65	5
## 326	BWin_v_Neut_L_NAcc_run1~1	66	1
## 327	BWin_v_Neut_L_NAcc_run1~1	66	2
## 328	BWin_v_Neut_L_NAcc_run1~1	66	3
## 329	BWin_v_Neut_L_NAcc_run1~1	66	4
## 330	BWin_v_Neut_L_NAcc_run1~1	66	5
## 331	BWin_v_Neut_R_NAcc_run1~1	67	1
## 332	BWin_v_Neut_R_NAcc_run1~1	67	2
## 333	BWin_v_Neut_R_NAcc_run1~1	67	3
## 334	BWin_v_Neut_R_NAcc_run1~1	67	4
## 335	BWin_v_Neut_R_NAcc_run1~1	67	5
## 336	BWin_v_Neut_R_Insula_run1~1	68	1
## 337	BWin_v_Neut_R_Insula_run1~1	68	2
## 338	BWin_v_Neut_R_Insula_run1~1	68	3
## 339	BWin_v_Neut_R_Insula_run1~1	68	4
## 340	BWin_v_Neut_R_Insula_run1~1	68	5
## 341	BWin_v_Neut_L_NAcc_run2~1	69	1
## 342	BWin_v_Neut_L_NAcc_run2~1	69	2
## 343	BWin_v_Neut_L_NAcc_run2~1	69	3
## 344	BWin_v_Neut_L_NAcc_run2~1	69	4
## 345	BWin_v_Neut_L_NAcc_run2~1	69	5
## 346	BWin_v_Neut_R_NAcc_run2~1	70	1
## 347	BWin_v_Neut_R_NAcc_run2~1	70	2
## 348	BWin_v_Neut_R_NAcc_run2~1	70	3
## 349	BWin_v_Neut_R_NAcc_run2~1	70	4
## 350	BWin_v_Neut_R_NAcc_run2~1	70	5
## 351	BWin_v_Neut_R_Insula_run2~1	71	1
## 352	BWin_v_Neut_R_Insula_run2~1	71	2
## 353	BWin_v_Neut_R_Insula_run2~1	71	3
## 354	BWin_v_Neut_R_Insula_run2~1	71	4
## 355	BWin_v_Neut_R_Insula_run2~1	71	5
## 356	BWin_v_BLose_L_NAcc_run1~1	72	1
## 357	BWin_v_BLose_L_NAcc_run1~1	72	2
## 358	BWin_v_BLose_L_NAcc_run1~1	72	3
## 359	BWin_v_BLose_L_NAcc_run1~1	72	4
## 360	BWin_v_BLose_L_NAcc_run1~1	72	5
## 361	BWin_v_BLose_R_NAcc_run1~1	73	1
## 362	BWin_v_BLose_R_NAcc_run1~1	73	2
## 363	BWin_v_BLose_R_NAcc_run1~1	73	3
## 364	BWin_v_BLose_R_NAcc_run1~1	73	4
## 365	BWin_v_BLose_R_NAcc_run1~1	73	5
## 366	BWin_v_BLose_L_NAcc_run2~1	74	1
## 367	BWin_v_BLose_L_NAcc_run2~1	74	2
## 368	BWin_v_BLose_L_NAcc_run2~1	74	3
## 369	BWin_v_BLose_L_NAcc_run2~1	74	4

## 370	BWin_v_BLose_L_NAcc_run2~1	74	5
## 371	BWin_v_BLose_R_NAcc_run2~1	75	1
## 372	BWin_v_BLose_R_NAcc_run2~1	75	2
## 373	BWin_v_BLose_R_NAcc_run2~1	75	3
## 374	BWin_v_BLose_R_NAcc_run2~1	75	4
## 375	BWin_v_BLose_R_NAcc_run2~1	75	5
## 376	ALose_v_Neut_L_Insula_run1~1	76	1
## 377	ALose_v_Neut_L_Insula_run1~1	76	2
## 378	ALose_v_Neut_L_Insula_run1~1	76	3
## 379	ALose_v_Neut_L_Insula_run1~1	76	4
## 380	ALose_v_Neut_L_Insula_run1~1	76	5
## 381	ALose_v_Neut_R_Insula_run1~1	77	1
## 382	ALose_v_Neut_R_Insula_run1~1	77	2
## 383	ALose_v_Neut_R_Insula_run1~1	77	3
## 384	ALose_v_Neut_R_Insula_run1~1	77	4
## 385	ALose_v_Neut_R_Insula_run1~1	77	5
## 386	ALose_v_Neut_L_Insula_run2~1	78	1
## 387	ALose_v_Neut_L_Insula_run2~1	78	2
## 388	ALose_v_Neut_L_Insula_run2~1	78	3
## 389	ALose_v_Neut_L_Insula_run2~1	78	4
## 390	ALose_v_Neut_L_Insula_run2~1	78	5
## 391	ALose_v_Neut_R_Insula_run2~1	79	1
## 392	ALose_v_Neut_R_Insula_run2~1	79	2
## 393	ALose_v_Neut_R_Insula_run2~1	79	3
## 394	ALose_v_Neut_R_Insula_run2~1	79	4
## 395	ALose_v_Neut_R_Insula_run2~1	79	5
## 396	BLose_v_Neut_L_Insula_run1~1	80	1
## 397	BLose_v_Neut_L_Insula_run1~1	80	2
## 398	BLose_v_Neut_L_Insula_run1~1	80	3
## 399	BLose_v_Neut_L_Insula_run1~1	80	4
## 400	BLose_v_Neut_L_Insula_run1~1	80	5
## 401	BLose_v_Neut_R_Insula_run1~1	81	1
## 402	BLose_v_Neut_R_Insula_run1~1	81	2
## 403	BLose_v_Neut_R_Insula_run1~1	81	3
## 404	BLose_v_Neut_R_Insula_run1~1	81	4
## 405	BLose_v_Neut_R_Insula_run1~1	81	5
## 406	BLose_v_Neut_L_Insula_run2~1	82	1
## 407	BLose_v_Neut_L_Insula_run2~1	82	2
## 408	BLose_v_Neut_L_Insula_run2~1	82	3
## 409	BLose_v_Neut_L_Insula_run2~1	82	4
## 410	BLose_v_Neut_L_Insula_run2~1	82	5
## 411	BLose_v_Neut_R_Insula_run2~1	83	1
## 412	BLose_v_Neut_R_Insula_run2~1	83	2
## 413	BLose_v_Neut_R_Insula_run2~1	83	3
## 414	BLose_v_Neut_R_Insula_run2~1	83	4
## 415	BLose_v_Neut_R_Insula_run2~1	83	5
## 416	BLose_v_BWin_L_Insula_run1~1	84	1
## 417	BLose_v_BWin_L_Insula_run1~1	84	2
## 418	BLose_v_BWin_L_Insula_run1~1	84	3
## 419	BLose_v_BWin_L_Insula_run1~1	84	4
## 420	BLose_v_BWin_L_Insula_run1~1	84	5
## 421	BLose_v_BWin_R_Insula_run1~1	85	1
## 422	BLose_v_BWin_R_Insula_run1~1	85	2
## 423	BLose_v_BWin_R_Insula_run1~1	85	3

## 424	BLose_v_BWin_R_Insula_run1~1	85	4
## 425	BLose_v_BWin_R_Insula_run1~1	85	5
## 426	BLose_v_BWin_L_Insula_run2~1	86	1
## 427	BLose_v_BWin_L_Insula_run2~1	86	2
## 428	BLose_v_BWin_L_Insula_run2~1	86	3
## 429	BLose_v_BWin_L_Insula_run2~1	86	4
## 430	BLose_v_BWin_L_Insula_run2~1	86	5
## 431	BLose_v_BWin_R_Insula_run2~1	87	1
## 432	BLose_v_BWin_R_Insula_run2~1	87	2
## 433	BLose_v_BWin_R_Insula_run2~1	87	3
## 434	BLose_v_BWin_R_Insula_run2~1	87	4
## 435	BLose_v_BWin_R_Insula_run2~1	87	5
## 436	Approach~1	88	1
## 437	Approach~1	88	2
## 438	Approach~1	88	3
## 439	Approach~1	88	4
## 440	Approach~1	88	5
## 441	Avoid~1	89	1
## 442	Avoid~1	89	2
## 443	Avoid~1	89	3
## 444	Avoid~1	89	4
## 445	Avoid~1	89	5
## 446	rmsea	90	1
## 447	rmsea	90	2
## 448	rmsea	90	3
## 449	rmsea	90	4
## 450	rmsea	90	5
## 451	cfi	91	1
## 452	cfi	91	2
## 453	cfi	91	3
## 454	cfi	91	4
## 455	cfi	91	5
## 456	tli	92	1
## 457	tli	92	2
## 458	tli	92	3
## 459	tli	92	4
## 460	tli	92	5
## 461	gfi	93	1
## 462	gfi	93	2
## 463	gfi	93	3
## 464	gfi	93	4
## 465	gfi	93	5
## 466	srmr	94	1
## 467	srmr	94	2
## 468	srmr	94	3
## 469	srmr	94	4
## 470	srmr	94	5
##	est	p	
## 1	-0.002	0.8	
## 2	0.009	0.4	
## 3	0.003	0.6	
## 4	-0.008	0.4	
## 5	-0.007	0.0	
## 6	0.006	0.2	

```
## 7   -0.004 0.6
## 8    0.000 0.8
## 9    0.003 1.0
## 10  -0.004 0.8
## 11   0.007 0.4
## 12   0.001 0.8
## 13  -0.005 0.6
## 14   0.001 0.8
## 15   0.002 0.8
## 16   0.001 0.6
## 17   0.003 0.6
## 18   0.001 1.0
## 19  -0.002 0.4
## 20  -0.006 0.4
## 21   0.014 0.2
## 22  -0.001 0.8
## 23  -0.001 0.2
## 24  -0.001 0.8
## 25  -0.010 0.6
## 26  -0.011 0.0
## 27  -0.002 1.0
## 28  -0.004 0.4
## 29   0.008 0.0
## 30   0.012 0.2
## 31   0.011 0.4
## 32   0.006 0.4
## 33   0.000 1.0
## 34  -0.001 0.6
## 35  -0.024 0.0
## 36  -0.006 0.8
## 37  -0.008 0.0
## 38   0.002 0.6
## 39   0.010 0.0
## 40  -0.005 1.0
## 41   0.010 0.6
## 42  -0.001 1.0
## 43   0.001 0.6
## 44  -0.002 1.0
## 45  -0.011 0.8
## 46  -0.008 0.8
## 47  -0.002 0.8
## 48   0.003 0.6
## 49   0.008 0.4
## 50  -0.014 0.2
## 51  -0.023 0.0
## 52  -0.014 0.0
## 53   0.000 1.0
## 54   0.016 0.0
## 55   0.021 0.2
## 56   0.022 0.0
## 57   0.000 0.8
## 58   0.005 0.0
## 59  -0.012 0.0
## 60  -0.013 0.4
```

```
## 61 -0.001 0.8
## 62  0.003 0.6
## 63  0.005 0.6
## 64 -0.003 0.8
## 65 -0.015 0.0
## 66  0.004 0.8
## 67 -0.006 0.2
## 68 -0.002 1.0
## 69  0.002 0.6
## 70  0.011 0.4
## 71 -0.018 0.2
## 72  0.001 0.4
## 73  0.006 0.6
## 74  0.001 0.8
## 75 -0.002 0.8
## 76  0.004 0.8
## 77  0.002 0.8
## 78 -0.002 1.0
## 79  0.002 0.8
## 80 -0.009 0.8
## 81 -0.005 0.8
## 82  0.005 1.0
## 83  0.000 1.0
## 84 -0.002 0.8
## 85  0.001 0.8
## 86 -0.014 0.4
## 87 -0.010 0.4
## 88  0.002 0.8
## 89  0.010 0.2
## 90  0.009 0.4
## 91  0.009 0.8
## 92  0.010 0.2
## 93 -0.001 0.8
## 94 -0.009 0.2
## 95 -0.009 0.4
## 96 -0.023 0.2
## 97 -0.004 1.0
## 98  0.014 0.0
## 99  0.001 0.8
## 100 -0.007 0.6
## 101 -0.028 0.0
## 102 -0.006 0.6
## 103  0.011 0.0
## 104  0.010 0.4
## 105 -0.010 0.4
## 106  0.006 0.6
## 107  0.003 0.6
## 108  0.003 0.6
## 109 -0.009 0.2
## 110 -0.002 1.0
## 111 -0.011 0.6
## 112 -0.004 0.6
## 113  0.006 0.2
## 114  0.003 0.8
```

```
## 115 -0.003 0.6
## 116 -0.010 0.4
## 117  0.002 0.8
## 118  0.003 0.6
## 119  0.002 0.8
## 120 -0.006 1.0
## 121 -0.023 0.0
## 122 -0.021 0.0
## 123  0.006 0.4
## 124  0.019 0.0
## 125  0.011 0.0
## 126  0.016 0.0
## 127  0.009 0.4
## 128 -0.006 0.0
## 129 -0.012 0.0
## 130  0.007 0.4
## 131 -0.013 0.2
## 132 -0.020 0.0
## 133  0.001 1.0
## 134  0.015 0.0
## 135  0.020 0.4
## 136  0.009 0.2
## 137  0.003 0.4
## 138 -0.003 0.6
## 139 -0.004 0.4
## 140  0.000 0.8
## 141 -0.002 0.6
## 142 -0.001 1.0
## 143  0.001 0.0
## 144  0.001 0.8
## 145 -0.001 0.6
## 146 -0.003 0.2
## 147 -0.002 0.0
## 148  0.001 0.8
## 149  0.002 0.0
## 150  0.002 0.4
## 151  0.003 0.0
## 152  0.000 0.8
## 153 -0.002 0.0
## 154  0.001 0.4
## 155  0.002 0.2
## 156 -0.002 0.6
## 157  0.001 1.0
## 158  0.000 0.6
## 159  0.000 0.8
## 160 -0.001 0.8
## 161  0.000 0.4
## 162 -0.001 0.4
## 163 -0.001 0.8
## 164  0.001 0.4
## 165  0.004 0.0
## 166  0.004 0.2
## 167  0.002 0.2
## 168 -0.001 0.4
```

```
## 169 -0.002 0.2
## 170 -0.003 0.0
## 171 -0.001 0.6
## 172 0.000 0.8
## 173 0.000 0.6
## 174 0.000 0.8
## 175 0.003 0.0
## 176 -0.001 0.6
## 177 0.001 0.8
## 178 0.000 0.6
## 179 0.001 0.4
## 180 -0.002 0.6
## 181 0.001 0.6
## 182 0.001 0.0
## 183 -0.001 0.6
## 184 0.000 0.8
## 185 -0.001 0.4
## 186 0.000 0.8
## 187 0.001 0.8
## 188 0.001 0.6
## 189 -0.002 0.4
## 190 -0.002 0.8
## 191 0.000 0.4
## 192 0.001 0.4
## 193 0.002 0.0
## 194 -0.001 0.2
## 195 -0.005 0.2
## 196 -0.001 0.4
## 197 0.002 0.2
## 198 -0.001 1.0
## 199 -0.001 0.6
## 200 0.001 1.0
## 201 0.005 0.0
## 202 0.000 1.0
## 203 0.000 0.2
## 204 -0.003 0.0
## 205 0.000 0.8
## 206 0.000 0.8
## 207 0.002 0.0
## 208 0.002 0.2
## 209 -0.001 0.4
## 210 -0.009 0.0
## 211 0.001 1.0
## 212 0.000 0.4
## 213 0.001 0.4
## 214 -0.001 0.4
## 215 -0.001 0.8
## 216 -0.004 0.2
## 217 -0.002 0.0
## 218 0.001 0.8
## 219 0.002 0.0
## 220 0.002 0.4
## 221 0.001 0.6
## 222 -0.001 0.0
```

```
## 223  0.002 0.0
## 224 -0.001 1.0
## 225 -0.004 0.4
## 226  0.002 0.2
## 227  0.003 0.2
## 228  0.000 0.6
## 229 -0.003 0.0
## 230 -0.002 0.8
## 231 -0.001 0.6
## 232  0.000 1.0
## 233  0.000 0.6
## 234 -0.001 0.4
## 235  0.002 0.2
## 236  0.007 0.2
## 237  0.003 0.0
## 238 -0.002 0.0
## 239 -0.003 0.0
## 240  0.000 0.8
## 241  0.000 1.0
## 242  0.001 0.8
## 243  0.001 0.6
## 244  0.000 0.8
## 245 -0.003 0.4
## 246 -0.001 0.4
## 247  0.001 0.6
## 248  0.001 0.6
## 249  0.000 1.0
## 250 -0.001 1.0
## 251 -0.003 0.0
## 252  0.000 1.0
## 253  0.003 0.0
## 254 -0.001 0.8
## 255 -0.004 0.0
## 256 -0.003 0.4
## 257 -0.002 0.8
## 258  0.000 0.8
## 259  0.002 0.4
## 260  0.003 0.6
## 261  0.002 0.4
## 262  0.002 0.2
## 263 -0.001 0.4
## 264 -0.003 0.0
## 265  0.002 0.2
## 266  0.005 0.0
## 267  0.002 0.0
## 268  0.000 0.8
## 269 -0.003 0.0
## 270 -0.002 0.6
## 271  0.002 0.6
## 272  0.002 0.0
## 273 -0.001 0.2
## 274  0.000 1.0
## 275 -0.003 0.0
## 276  0.000 1.0
```

```
## 277 0.001 0.8
## 278 0.001 0.6
## 279 -0.001 0.4
## 280 0.000 0.8
## 281 0.000 1.0
## 282 0.000 1.0
## 283 0.000 1.0
## 284 0.000 1.0
## 285 0.000 1.0
## 286 0.000 1.0
## 287 0.000 1.0
## 288 0.000 1.0
## 289 0.000 1.0
## 290 0.000 1.0
## 291 0.061 0.2
## 292 0.028 0.2
## 293 -0.040 0.0
## 294 0.010 1.0
## 295 -0.043 0.4
## 296 0.031 0.0
## 297 0.014 0.2
## 298 -0.008 0.0
## 299 -0.013 0.0
## 300 -0.015 0.2
## 301 -0.007 1.0
## 302 0.002 0.8
## 303 0.011 0.0
## 304 -0.006 0.2
## 305 -0.016 0.2
## 306 -0.008 0.4
## 307 0.003 0.6
## 308 0.003 0.6
## 309 -0.002 0.8
## 310 -0.001 0.6
## 311 0.015 0.0
## 312 0.008 0.2
## 313 -0.002 0.4
## 314 -0.010 0.0
## 315 -0.008 0.4
## 316 0.000 0.6
## 317 0.008 0.4
## 318 0.003 1.0
## 319 -0.005 0.2
## 320 -0.014 0.2
## 321 0.003 0.0
## 322 -0.001 0.8
## 323 0.004 0.6
## 324 -0.004 0.2
## 325 -0.003 0.0
## 326 0.002 0.4
## 327 -0.003 0.8
## 328 -0.002 0.4
## 329 0.005 0.2
## 330 -0.003 0.8
```

```
## 331  0.020 0.0
## 332  0.000 0.4
## 333 -0.005 0.2
## 334 -0.006 0.6
## 335  0.003 0.4
## 336  0.003 0.6
## 337 -0.005 0.4
## 338 -0.002 0.6
## 339  0.004 0.2
## 340  0.004 0.6
## 341  0.002 0.6
## 342  0.000 1.0
## 343 -0.004 0.2
## 344  0.005 0.8
## 345  0.000 0.6
## 346  0.019 0.0
## 347 -0.001 0.8
## 348 -0.005 0.2
## 349 -0.003 0.2
## 350  0.001 1.0
## 351  0.016 0.0
## 352 -0.003 0.8
## 353 -0.001 0.6
## 354 -0.001 1.0
## 355 -0.009 0.4
## 356  0.019 0.2
## 357 -0.002 0.4
## 358 -0.005 0.4
## 359 -0.007 0.4
## 360  0.012 0.2
## 361  0.013 0.0
## 362  0.015 0.0
## 363 -0.005 0.0
## 364 -0.014 0.0
## 365 -0.001 0.6
## 366  0.018 0.0
## 367  0.008 0.4
## 368 -0.009 0.0
## 369 -0.009 0.2
## 370  0.008 0.2
## 371  0.013 0.0
## 372  0.008 0.0
## 373 -0.010 0.0
## 374 -0.004 0.0
## 375  0.006 0.4
## 376  0.005 1.0
## 377 -0.003 0.6
## 378 -0.004 1.0
## 379  0.002 1.0
## 380  0.008 0.8
## 381  0.006 0.8
## 382  0.005 0.2
## 383  0.001 0.8
## 384 -0.005 0.2
```



```
## 385 -0.011 0.4
## 386 0.001 0.8
## 387 0.001 0.8
## 388 -0.007 0.6
## 389 0.004 0.6
## 390 0.008 0.6
## 391 -0.002 0.4
## 392 0.005 0.4
## 393 0.002 0.4
## 394 0.000 0.6
## 395 -0.013 0.0
## 396 0.022 0.0
## 397 -0.002 0.6
## 398 -0.008 0.0
## 399 -0.001 1.0
## 400 0.005 0.8
## 401 -0.009 0.4
## 402 -0.011 0.0
## 403 0.002 0.6
## 404 0.008 0.2
## 405 0.012 0.0
## 406 -0.003 0.4
## 407 -0.002 1.0
## 408 0.002 0.2
## 409 0.003 0.8
## 410 -0.004 0.8
## 411 -0.008 0.4
## 412 -0.006 0.2
## 413 0.006 0.0
## 414 0.005 0.4
## 415 -0.006 0.4
## 416 -0.002 0.8
## 417 0.008 0.0
## 418 0.002 0.2
## 419 -0.007 0.2
## 420 -0.007 0.6
## 421 0.002 0.4
## 422 0.000 0.8
## 423 0.004 0.4
## 424 0.000 0.8
## 425 -0.013 0.6
## 426 0.000 0.8
## 427 0.000 0.8
## 428 0.004 0.2
## 429 -0.007 0.0
## 430 0.004 0.8
## 431 -0.027 0.0
## 432 -0.007 0.2
## 433 0.008 0.0
## 434 0.008 0.0
## 435 0.006 0.4
## 436 0.000 1.0
## 437 0.000 1.0
## 438 0.000 1.0
```

```
## 439 0.000 1.0
## 440 0.000 1.0
## 441 0.000 1.0
## 442 0.000 1.0
## 443 0.000 1.0
## 444 0.000 1.0
## 445 0.000 1.0
## 446 0.006 0.0
## 447 -0.004 0.2
## 448 -0.004 0.8
## 449 0.002 0.0
## 450 0.011 0.2
## 451 -0.065 0.8
## 452 0.010 1.0
## 453 0.037 0.2
## 454 0.006 0.8
## 455 -0.071 0.4
## 456 -0.071 0.8
## 457 0.010 1.0
## 458 0.040 0.2
## 459 0.007 0.8
## 460 -0.077 0.4
## 461 -0.016 0.0
## 462 0.012 0.4
## 463 0.012 0.8
## 464 -0.005 0.0
## 465 -0.029 0.2
## 466 0.006 0.4
## 467 -0.002 0.8
## 468 -0.004 0.8
## 469 0.000 0.6
## 470 0.008 0.4
```

Sensitivity Analyses

In the manuscript, several sensitivity analyses are proposed:

1. Differences (CFA/EFA) in effects across data collected only at UM site.
2. Resampling of CFA/EFA in effects across larger ABCD data to evaluate the *stability* of model effects

UM Specific CFA/ESEM/EFA

In the chunk below, the PDS simulated data will be used as the filter, assuming that PDS 3 are the values reflecting the site of interest.

```
ABCD_site_spec <- sim_ABCD_data %>% filter(PDS == 3)

sim_AHRB_data$PDS <- NA
sim_MLS_data$PDS <- NA

UM_site_spec <- rbind(sim_AHRB_data,
                      sim_MLS_data,
                      ABCD_site_spec # UM specified ABCD data
)
```

CFA

```
UM_all_sample <- cfa(model = MID_model, data = UM_site_spec,
  estimator = "MLR", std.lv = TRUE, meanstructure = TRUE)

UM_config_cfa <- cfa(model = MID_model, data = UM_site_spec, group = 'set',
  estimator = "MLR", std.lv = TRUE, meanstructure = TRUE)

UM_metric_cfa <- cfa(model = MID_model, data = UM_site_spec,
  group = 'set', group.equal=c("loadings"),
  estimator = "MLR", std.lv = TRUE, meanstructure = TRUE)

UMsite_out <- matrix(NA, ncol = 9, nrow = 3)
colnames(UMsite_out) <- c("model", "chisq", "df", "pvalue", "rmsea", "cfi", "srmr",
  "AIC", "BIC")

# save fit measures from models
UMsite_out[1,2:7] <- round(data.matrix(fitmeasures(UM_all_sample,
  fit.measures = c("chisq", "df", "pvalue",
    "rmsea", "cfi", "srmr"))),
  digits=3)

UMsite_out[2,2:7] <- round(data.matrix(fitmeasures(UM_config_cfa,
  fit.measures = c("chisq", "df", "pvalue",
    "rmsea", "cfi", "srmr"))),
  digits=3)

UMsite_out[3,2:7] <- round(data.matrix(fitmeasures(UM_metric_cfa,
  fit.measures = c("chisq", "df", "pvalue",
    "rmsea", "cfi", "srmr"))),
  digits=3)

UMsite_out[1,8] <- round(AIC(UM_all_sample), 3)
UMsite_out[2,8] <- round(AIC(UM_config_cfa), 3)
UMsite_out[3,8] <- round(AIC(UM_metric_cfa), 3)

# BIC models
UMsite_out[1,9] <- round(BIC(UM_all_sample), 3)
UMsite_out[2,9] <- round(BIC(UM_config_cfa), 3)
UMsite_out[3,9] <- round(BIC(UM_metric_cfa), 3)

UMsite_out[1:3,1] <- c("Overall CFA", "Config MG-CFA", "Metric MG-CFA")

UMsite_out %>%
  knitr::kable(
    caption = "Fit statistics from MG-CFA and ESEM models",
    booktabs = TRUE
  )
```

Table 3: Fit statistics from MG-CFA and ESEM models

model	chisq	df	pvalue	rmsea	cfi	srmr	AIC	BIC
Overall CFA	1167.367	349	0	0.065	0.665	0.057	-9462.261	-9094.083
Config MG-CFA	1915.884	1047	0	0.067	0.654	0.073	-9275.648	-8171.115
Metric MG-CFA	1974.859	1099	0	0.065	0.651	0.078	-9320.674	-8441.379

EFA

In this case, only the data for the ABCD sample changes (reduced from 21 → 1 site [UM]). Hence, rerunning only the ABCD EFA. Factor congruence estimation will use EFA from above for AHRB and MLS.

```
UM_abcd_efadata = subset(UM_site_spec[,c("AWin_v_Neut_L_NAcc_run1" , "AWin_v_Neut_L_NAcc_run2" ,
    "BWin_v_Neut_L_NAcc_run1" , "BWin_v_Neut_L_NAcc_run2" ,
    "BWin_v_BLose_L_NAcc_run1" , "BWin_v_BLose_L_NAcc_run2",
    "AWin_v_Neut_R_NAcc_run1" , "AWin_v_Neut_R_NAcc_run2",
    "BWin_v_Neut_R_NAcc_run1" , "BWin_v_Neut_R_NAcc_run2",
    "BWin_v_BLose_R_NAcc_run1" , "BWin_v_BLose_R_NAcc_run2",
    # insula values approach
    "AWin_v_Neut_R_Insula_run1", "AWin_v_Neut_R_Insula_run2",
    "BWin_v_Neut_R_Insula_run1", "BWin_v_Neut_R_Insula_run2",
    # avoidance
    "ALose_v_Neut_L_Insula_run1", "ALose_v_Neut_L_Insula_run2",
    "BLose_v_Neut_L_Insula_run1", "BLose_v_Neut_L_Insula_run2",
    "BLose_v_BWin_L_Insula_run1", "BLose_v_BWin_L_Insula_run2",
    "ALose_v_Neut_R_Insula_run1", "ALose_v_Neut_R_Insula_run2",
    "BLose_v_Neut_R_Insula_run1", "BLose_v_Neut_R_Insula_run2",
    "BLose_v_BWin_R_Insula_run1", "BLose_v_BWin_R_Insula_run2",
    "set")] %>% filter(set==1))

paran(x = UM_abcd_efadata[,1:28],
    iterations = 1000, quietly = FALSE, centile = 95,
    status = FALSE, all = TRUE, cfa = TRUE, graph = TRUE, color = TRUE,
    col = c("black", "red", "blue"), lty = c(1, 2, 3), lwd = 1, legend = TRUE,
    seed = 100)

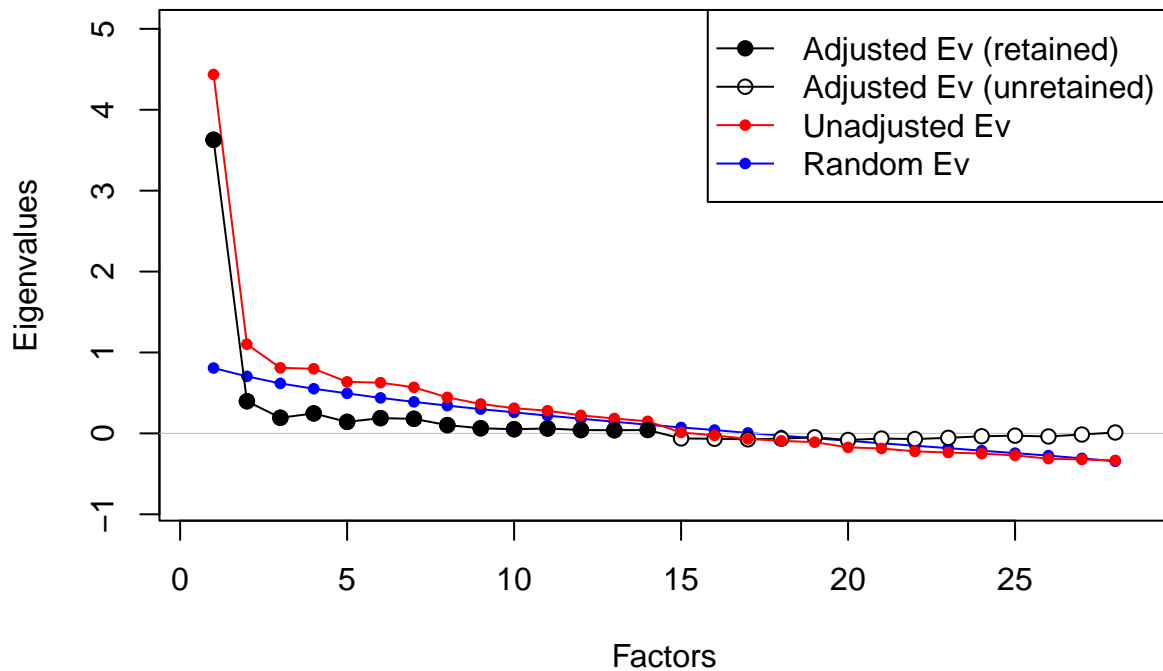
##
## Using eigendecomposition of correlation matrix.
##
## Results of Horn's Parallel Analysis for factor retention
## 1000 iterations, using the 95 centile estimate
##
## -----
## Factor      Adjusted      Unadjusted      Estimated
##             Eigenvalue    Eigenvalue      Bias
## -----
## No components passed.
## -----
## 1           3.628570      4.436251      0.807681
## 2           0.397564      1.102327      0.704762
## 3           0.192784      0.810407      0.617622
## 4           0.246962      0.798866      0.551903
## 5           0.142275      0.636878      0.494602
## 6           0.188467      0.627145      0.438678
## 7           0.179993      0.569582      0.389589
```

```

## 8      0.102263    0.445540    0.343277
## 9      0.063590    0.363779    0.300188
## 10     0.051445    0.311718    0.260272
## 11     0.060812    0.280569    0.219757
## 12     0.041416    0.222636    0.181219
## 13     0.038991    0.183582    0.144591
## 14     0.042094    0.149378    0.107284
## 15     -0.062117   0.011462    0.073580
## 16     -0.066274   -0.02475   0.041519
## 17     -0.073236   -0.06771   0.005522
## 18     -0.065103   -0.09174   -0.02664
## 19     -0.048208   -0.10881   -0.06060
## 20     -0.081346   -0.17265   -0.09130
## 21     -0.063693   -0.18635   -0.12266
## 22     -0.069725   -0.22210   -0.15238
## 23     -0.054965   -0.23672   -0.18176
## 24     -0.035486   -0.24929   -0.21380
## 25     -0.027979   -0.27179   -0.24381
## 26     -0.038075   -0.31251   -0.27444
## 27     -0.013812   -0.32266   -0.30885
## 28      0.011275   -0.33361   -0.34489
## -----
##
## Adjusted eigenvalues > 0 indicate dimensions to retain.
## (14 factors      retained)

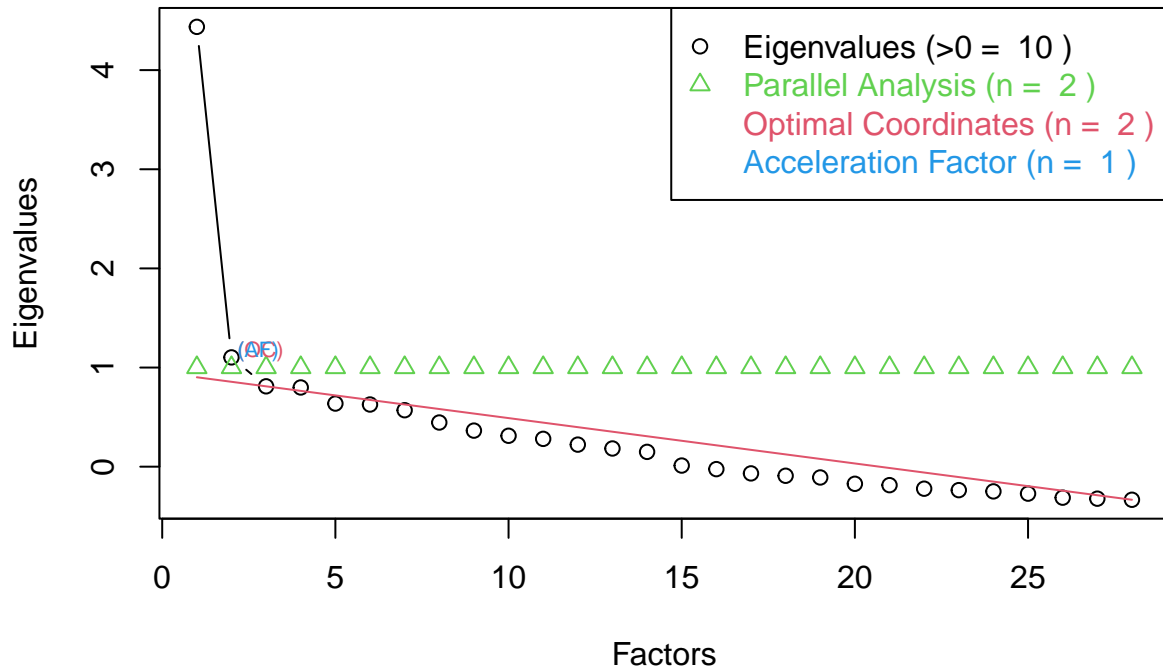
```

Parallel Analysis



```
plot(nScrees(x=UM_abcd_efadata[,1:28],model="factors"))
```

Non Graphical Solutions to Scree Test



```
UM_abcd_efa <- factanal(x = UM_abcd_efadata[,1:28], #raw data
  factors = 2, rotation = "promax" # oblique rotation allow for non-orthogonal structure
)

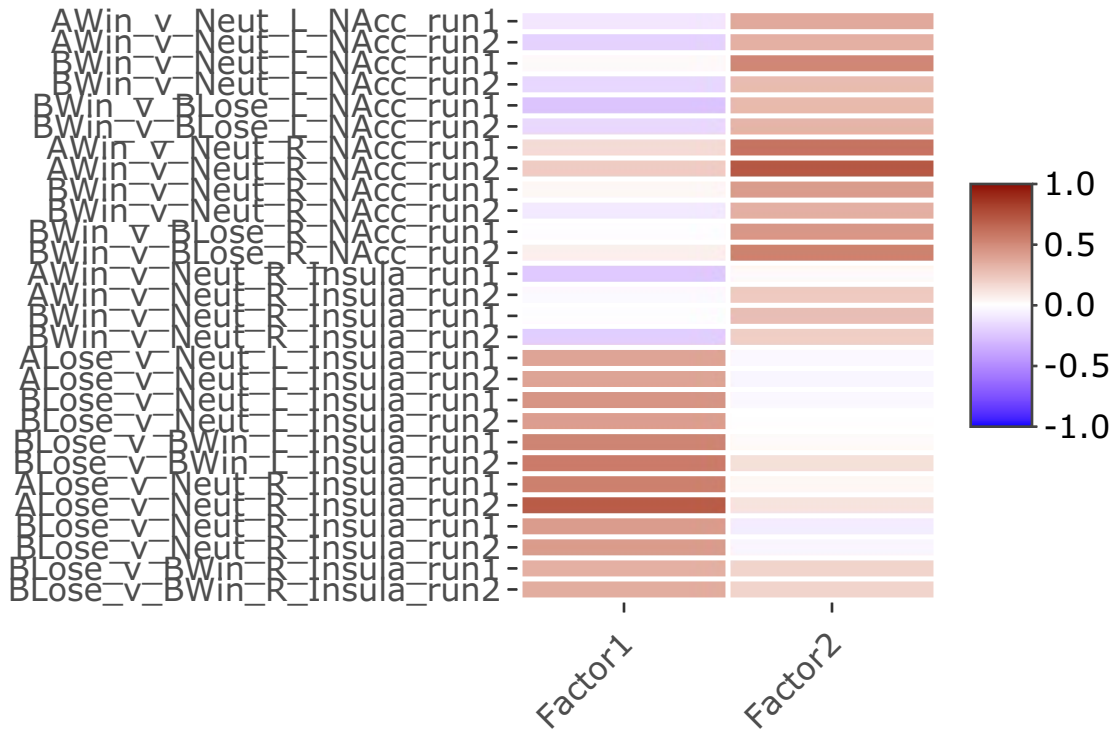
heatmaply(round(UM_abcd_efa$loadings[,1:2],2) %>% print(sort = T),
  scale_fill_gradient_fun = ggplot2::scale_fill_gradient2(
    low = "blue",
    high = "darkred",
    space = "Lab",
    midpoint = 0,
    limits = c(-1, 1)
  ),
  dendrogram = "none",
  xlab = "", ylab = "",
  main = "",
  margins = c(60,100,40,20),
  grid_color = "white",
  grid_width = 0.00001,
  titleX = FALSE,
  hide_colorbar = FALSE,
  branches_lwd = 0.1,
  label_names = c("Brain:", "Feature:", "Value"),
  fontsize_row = 9, fontsize_col = 9,
  labCol = colnames(UM_abcd_efa$loadings[,1:2]),
```

```

labRow = rownames(UM_abcd_efa$loadings[,1:2]),
heatmap_layers = theme(axis.line=element_blank()),
)

```

##	Factor1	Factor2
## AWin_v_Neut_L_NAcc_run1	-0.10	0.36
## AWin_v_Neut_L_NAcc_run2	-0.19	0.33
## BWin_v_Neut_L_NAcc_run1	0.02	0.51
## BWin_v_Neut_L_NAcc_run2	-0.17	0.28
## BWin_v_BLose_L_NAcc_run1	-0.25	0.29
## BWin_v_BLose_L_NAcc_run2	-0.16	0.31
## AWin_v_Neut_R_NAcc_run1	0.15	0.59
## AWin_v_Neut_R_NAcc_run2	0.21	0.71
## BWin_v_Neut_R_NAcc_run1	0.03	0.42
## BWin_v_Neut_R_NAcc_run2	-0.09	0.33
## BWin_v_BLose_R_NAcc_run1	-0.01	0.44
## BWin_v_BLose_R_NAcc_run2	0.06	0.53
## AWin_v_Neut_R_Insula_run1	-0.23	0.02
## AWin_v_Neut_R_Insula_run2	-0.02	0.22
## BWin_v_Neut_R_Insula_run1	-0.01	0.27
## BWin_v_Neut_R_Insula_run2	-0.21	0.20
## ALose_v_Neut_L_Insula_run1	0.38	-0.03
## ALose_v_Neut_L_Insula_run2	0.38	-0.04
## BLose_v_Neut_L_Insula_run1	0.45	-0.03
## BLose_v_Neut_L_Insula_run2	0.41	0.00
## BLose_v_BWin_L_Insula_run1	0.52	0.02
## BLose_v_BWin_L_Insula_run2	0.56	0.13
## ALose_v_Neut_R_Insula_run1	0.54	0.03
## ALose_v_Neut_R_Insula_run2	0.69	0.11
## BLose_v_Neut_R_Insula_run1	0.42	-0.08
## BLose_v_Neut_R_Insula_run2	0.42	-0.04
## BLose_v_BWin_R_Insula_run1	0.33	0.18
## BLose_v_BWin_R_Insula_run2	0.35	0.18



```
fa.congruence(x = list(UM_abcd_efa, mls_efa, ahrb_efa), digits = 2) %>%
  knitr::kable(
    col.names = c("1. ABCD F1", "2. ABCD F2",
                  "3. MLS F1", "4. MLS F2", "5. MLS F3",
                  "6. AHRB F1", "7. AHRB F2", "8. AHRB F3", "8. AHRB F4"),
    caption = "ABCD, MLS and AHRB EFA Factor Congruence",
    booktabs = TRUE
  )
```

Table 4: ABCD, MLS and AHRB EFA Factor Congruence

	1. ABCD F1	2. ABCD F2	3. MLS F1	4. MLS F2	5. MLS F3	6. AHRB F1	7. AHRB F2	8. AHRB F3	8. AHRB F4
Factor1	1.00	0.05	-0.10	0.79	0.44	0.66	-0.25	0.06	0.69
Factor2	0.05	1.00	0.89	0.12	0.14	-0.15	0.57	0.48	0.08
Factor1	-0.10	0.89	1.00	0.07	0.15	-0.21	0.68	0.38	-0.01
Factor2	0.79	0.12	0.07	1.00	0.05	0.51	-0.07	0.02	0.67
Factor3	0.44	0.14	0.15	0.05	1.00	0.37	0.13	0.00	0.34
Factor1	0.66	-0.15	-0.21	0.51	0.37	1.00	-0.06	0.17	0.22
Factor2	-0.25	0.57	0.68	-0.07	0.13	-0.06	1.00	-0.12	-0.09
Factor3	0.06	0.48	0.38	0.02	0.00	0.17	-0.12	1.00	-0.13
Factor4	0.69	0.08	-0.01	0.67	0.34	0.22	-0.09	-0.13	1.00

Resampling ABCD CFA/EFA

```
abcd_resamp_df <- subset(brain_set[,c("AWin_v_Neut_L_NAcc_run1" , "AWin_v_Neut_L_NAcc_run2" ,
    "BWin_v_Neut_L_NAcc_run1" , "BWin_v_Neut_L_NAcc_run2" ,
    "BWin_v_BLose_L_NAcc_run1" , "BWin_v_BLose_L_NAcc_run2",
    "AWin_v_Neut_R_NAcc_run1" , "AWin_v_Neut_R_NAcc_run2",
    "BWin_v_Neut_R_NAcc_run1" , "BWin_v_Neut_R_NAcc_run2",
    "BWin_v_BLose_R_NAcc_run1" , "BWin_v_BLose_R_NAcc_run2",
    # insula values approach
    "AWin_v_Neut_R_Insula_run1", "AWin_v_Neut_R_Insula_run2",
    "BWin_v_Neut_R_Insula_run1", "BWin_v_Neut_R_Insula_run2",
    # avoidance
    "ALose_v_Neut_L_Insula_run1", "ALose_v_Neut_L_Insula_run2",
    "BLose_v_Neut_L_Insula_run1", "BLose_v_Neut_L_Insula_run2",
    "BLose_v_BWin_L_Insula_run1", "BLose_v_BWin_L_Insula_run2",
    "ALose_v_Neut_R_Insula_run1", "ALose_v_Neut_R_Insula_run2",
    "BLose_v_Neut_R_Insula_run1", "BLose_v_Neut_R_Insula_run2",
    "BLose_v_BWin_R_Insula_run1", "BLose_v_BWin_R_Insula_run2",
    "set")] %>% filter(set==1))
```

For resampling of the CFA model, using `bootstrapLavaan` for `lavaan`. First fitting the cfa model, then creating 1000 bootstrapped sampled and extracting fit statistics

```
set.seed(1111)
boot_cfa <- cfa(model = MID_model, data = abcd_resamp_df,
    estimator = "MLR", std.lv = TRUE, meanstructure = TRUE)

out_cfaboot <- bootstrapLavaan(object = boot_cfa, R = 1000,
    FUN = fitMeasures,
    fit.measures=c("chisq", "rmsea", "cfi",
        "tli", "srmr", "AIC", "BIC"),
    parallel="multicore", ncpus=4)

cfaboot_df <- data.frame(out_cfaboot)
```

For the resampling of the EFA model, using `sample_n` to get [1000] samples of $N = 1000$ (w/ replacement) of ABCD data. Then, estimating recommended N of factors per parallel analysis. Saving this to dataframe

```
# create empty df where data will be saved and compiled for models
#resampled_cfa <- matrix(NA, ncol = 11, nrow = 50) # resampling 50x for Config + #Metric models
#colnames(resampled_cfa) <- c("Sample",
#    "chisq", "df", "pvalue",
#    "rmsea", "cfi", "tli", "srmr", "AIC", "BIC", "Factors")
#
#n_loop = 50

resampled_cfa <- matrix(NA, ncol = 2, nrow = 1000) # resampling 50x for Config + Metric models
colnames(resampled_cfa) <- c("Sample", "Factors")

samples = 1000

for (s in 1:samples) {

    sub_df <- sample_n(tbl = abcd_resamp_df, size = 1000, replace = TRUE)
```

```

resampled_cfa[s,1] <- s

val <- nScree(x=sub_df[,1:28],model="factors")
resampled_cfa[s,2] <- as.integer(val$Components[3])

}

resampled_res <- data.frame(resampled_cfa)

```

Plotting the mean + 95% Confidence interval of values

```

n = 1000

ci_plt1 <- cfaboot_df %>%
  gather(key = "FitIndex", value = "Statistic",
    srmr,rmsea) %>%
  dplyr::group_by(FitIndex) %>%
  dplyr::summarize(m = mean(Statistic), stdev = sd(Statistic)) %>%
  ggplot(aes(x =FitIndex, y = m, fill = FitIndex, color=FitIndex)) +
  geom_point()+
  geom_errorbar(aes(ymin=m-(1.96*stdev/sqrt(n)),
    ymax=m+(1.96*stdev/sqrt(n))),
    width=.2,
    position=position_dodge(0.05))+
  labs(
    title = 'CFA: RMSEA & SRMR',
    x = 'Fit Stats',
    y = 'Type',
  )+
  theme_minimal()

ci_plt2 <- cfaboot_df %>%
  gather(key = "FitIndex", value = "Statistic",
    cfi,tli) %>%
  dplyr::group_by(FitIndex) %>%
  dplyr::summarize(m = mean(Statistic), stdev = sd(Statistic)) %>%
  ggplot(aes(x =FitIndex, y = m, fill = FitIndex, color=FitIndex)) +
  geom_point()+
  geom_errorbar(aes(ymin=m-(1.96*stdev/sqrt(n)),
    ymax=m+(1.96*stdev/sqrt(n))),
    width=.2,
    position=position_dodge(0.05))+
  labs(
    title = 'CFA: CFI & TLI',
    x = 'Fit Stats',
    y = 'Type',
  )+
  theme_minimal()

ci_plt3 <- cfaboot_df %>%

```

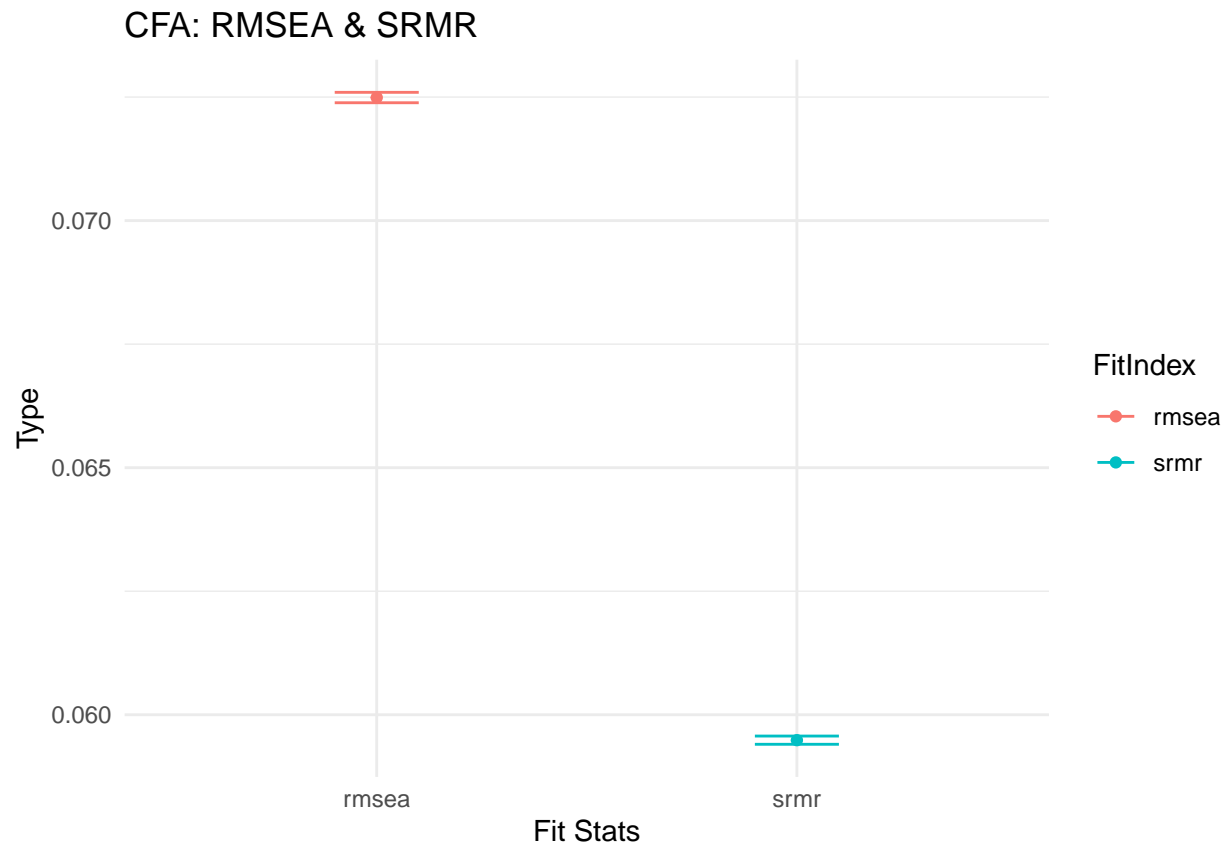
```

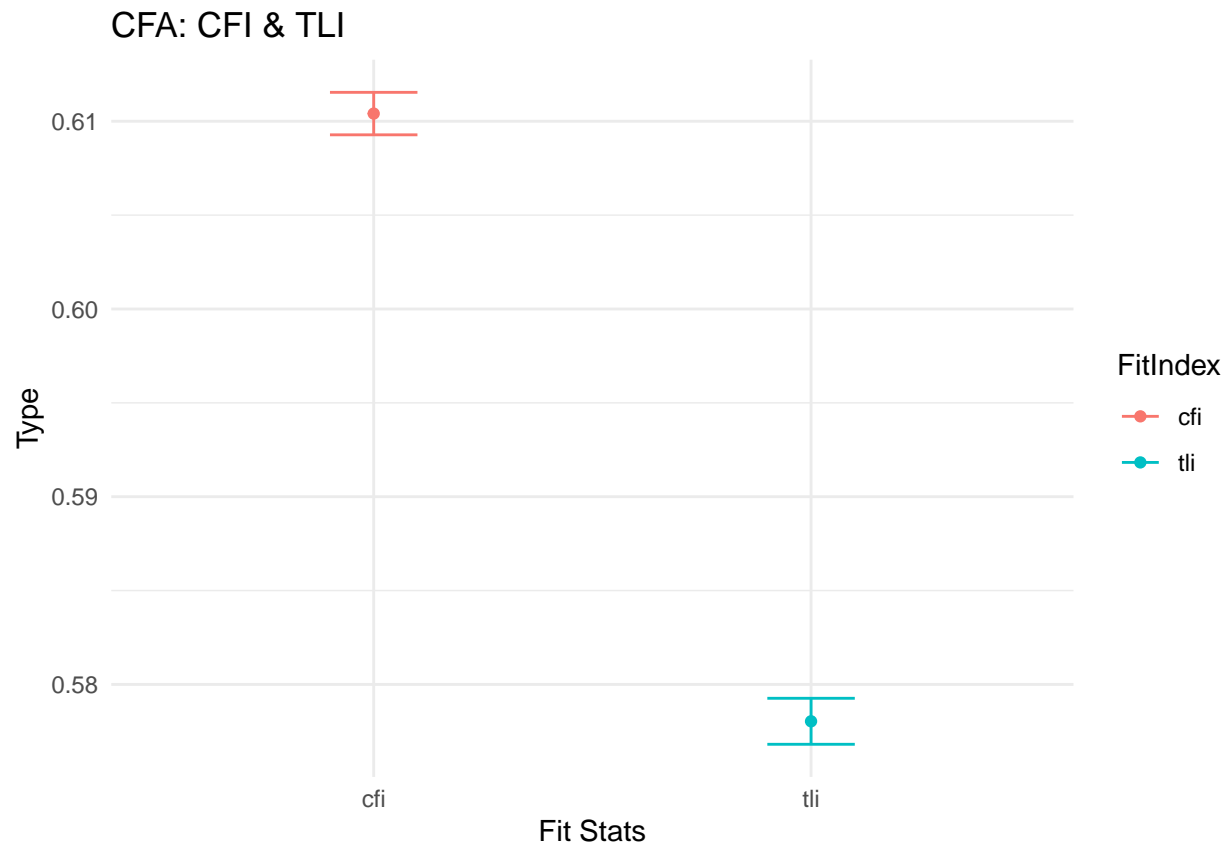
gather(key = "FitIndex", value = "Statistic",
        aic,bic) %>%
dplyr::group_by(FitIndex) %>%
dplyr::summarize(m = mean(Statistic), stdev = sd(Statistic)) %>%
ggplot(aes(x =FitIndex, y = m, fill = FitIndex, color=FitIndex)) +
geom_point()+
geom_errorbar(aes(ymin=m-(1.96*stdev/sqrt(n)),
                  ymax=m+(1.96*stdev/sqrt(n))),
              width=.2,
              position=position_dodge(0.05))+
labs(
  title = 'CFA: AIC & BIC',
  x = 'Fit Stats',
  y = 'Type',
)+
theme_minimal()

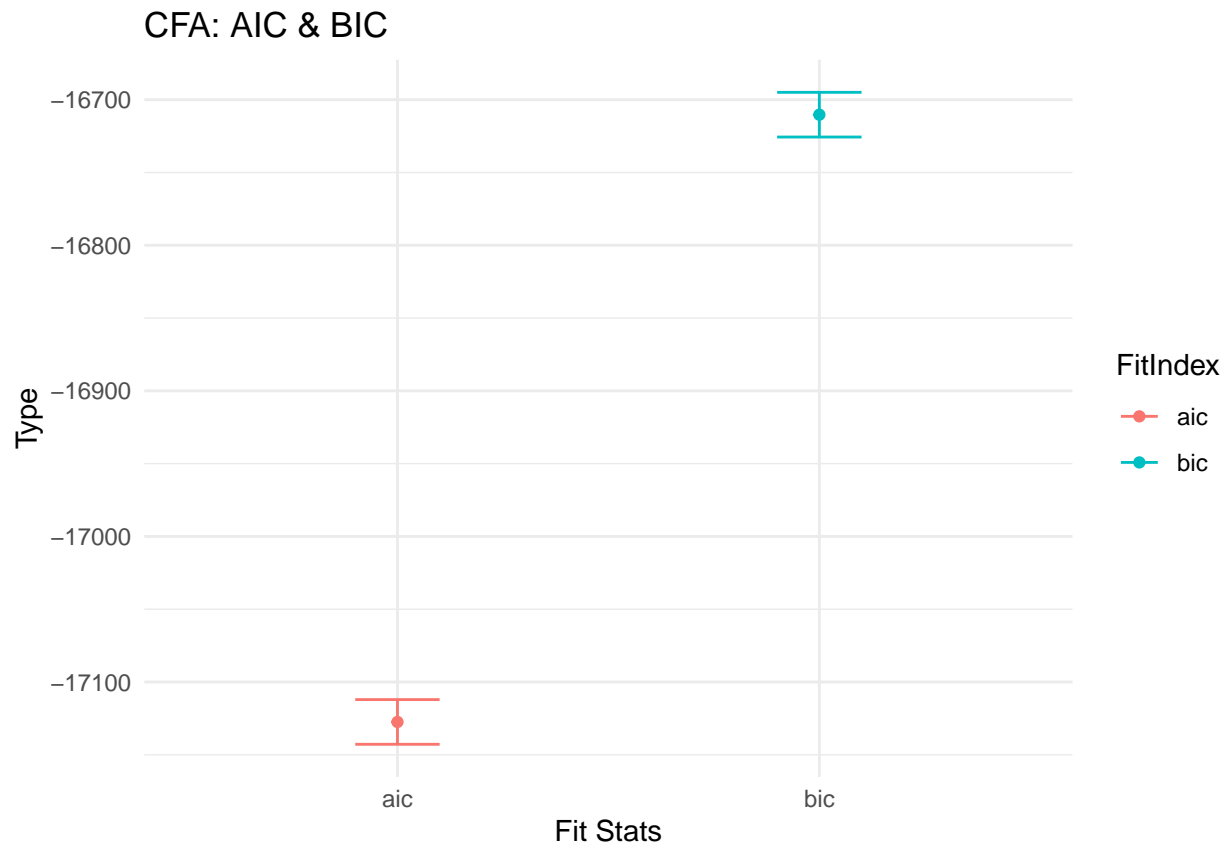
ci_plt4 = resampled_res %>%
dplyr::summarize(m = mean(Factors), stdev = sd(Factors)) %>%
ggplot(aes(x = "", y = m)) +
geom_point()+
geom_errorbar(aes(ymin=m-(1.96*stdev/sqrt(n)),
                  ymax=m+(1.96*stdev/sqrt(n))),
              width=.2,
              position=position_dodge(0.05))+
ylim(0,5)+
labs(
  title = 'EFA Factors',
  x = '',
  y = 'Number of Factors',
)+
theme_minimal()

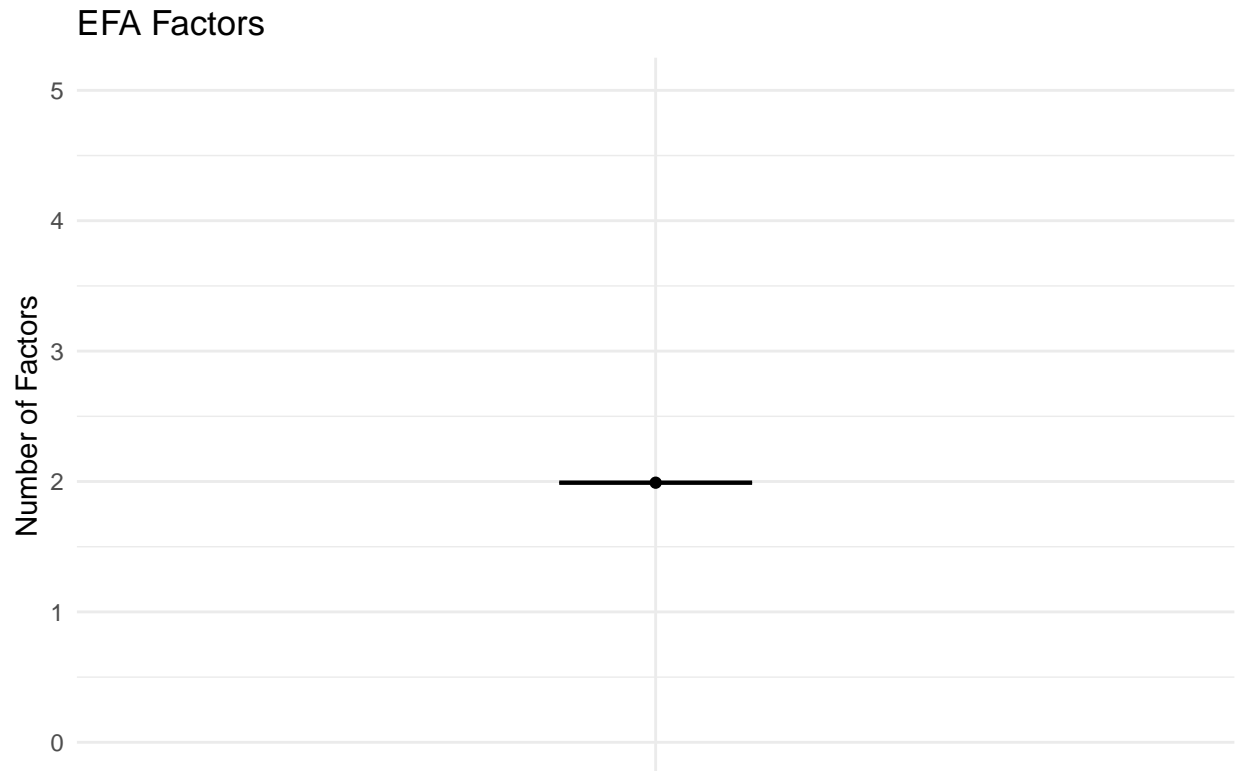
ci_plt1;ci_plt2;ci_plt3;ci_plt4

```









Plotting the distribution of values

```
plt1 <- cfaboot_df %>%
  gather(key = "FitIndex", value = "Statistic",
    srmr,rmsea) %>%
  ggplot(aes(x =FitIndex, y = Statistic, fill = FitIndex, color=FitIndex)) +
  ggdist::stat_halfeye(adjust = .5, width = .7, .width = 0, justification = -.2,
    point_colour = NA, alpha = .5) +
  geom_boxplot(width = .2, outlier.shape = NA, alpha = .3) +
  geom_jitter(width = .05, alpha = .5) +
  theme_minimal()+
  labs(
    title = 'CFA: RMSEA & SRMR',
    x = 'Fit Stats',
    y = 'Type',
  )

plt2 = cfaboot_df %>%
  gather(key = "FitIndex", value = "Statistic",
    cfi,tli) %>%
  ggplot(aes(x =FitIndex, y = Statistic, fill = FitIndex, color=FitIndex)) +
  ggdist::stat_halfeye(adjust = .5, width = .7, .width = 0, justification = -.2,
    point_colour = NA, alpha = .5) +
  geom_boxplot(width = .2, outlier.shape = NA, alpha = .3) +
  geom_jitter(width = .05, alpha = .5) +
  theme_minimal()+
  labs(
```

```

    title = 'CFA: CFI & TLI',
    x = 'Fit Stats',
    y = 'Type',
  )

plt3 = cfaboot_df %>%
  gather(key = "FitIndex", value = "Statistic",
         aic,bic) %>%
  ggplot(aes(x =FitIndex, y = Statistic, fill = FitIndex, color=FitIndex)) +
  ggdist::stat_halfeye(adjust = .5, width = .7, .width = 0, justification = -.2,
                      point_colour = NA, alpha = .5) +
  geom_boxplot(width = .2, outlier.shape = NA, alpha = .3) +
  geom_jitter(width = .05, alpha = .5) +
  theme_minimal()+
  labs(
    title = 'CFA: AIC & BIC',
    x = 'Fit Stats',
    y = 'Type',
  )

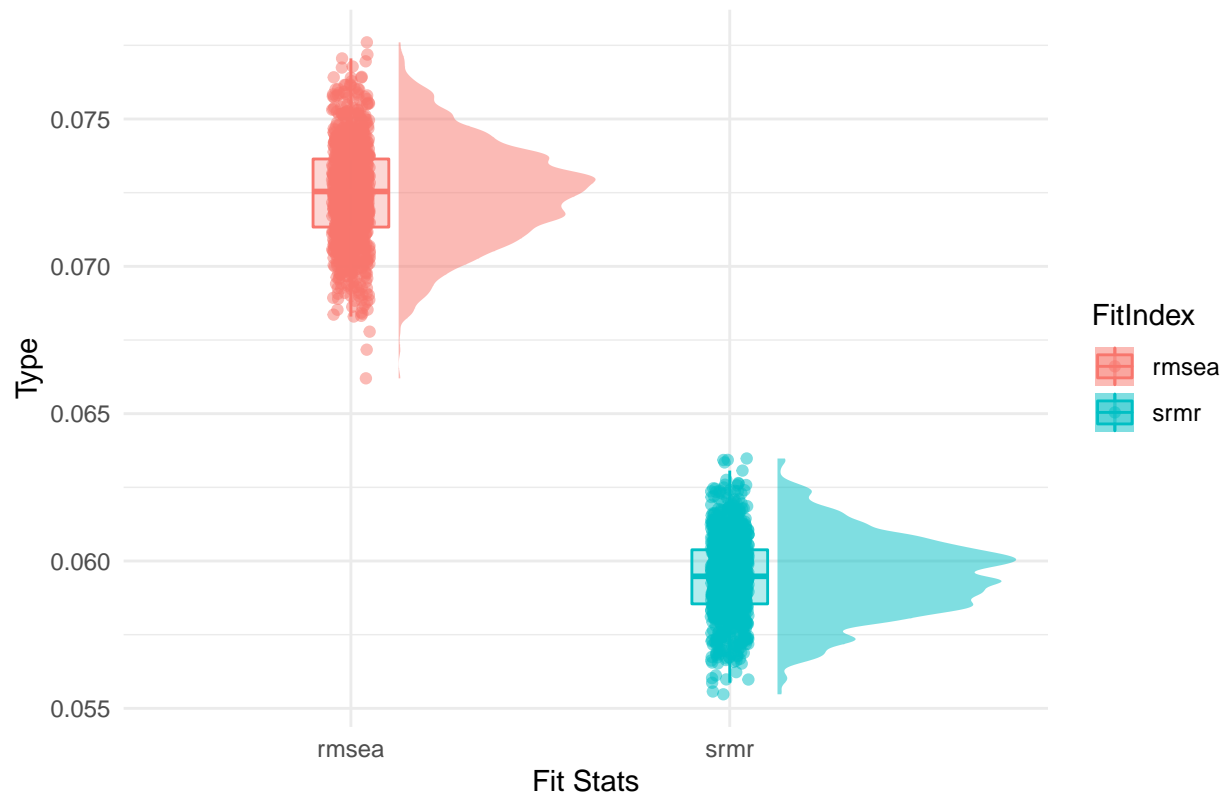
avg = round(mean(resampled_res$Factors),1)
minimum = min(resampled_res$Factors)
maximum = max(resampled_res$Factors)

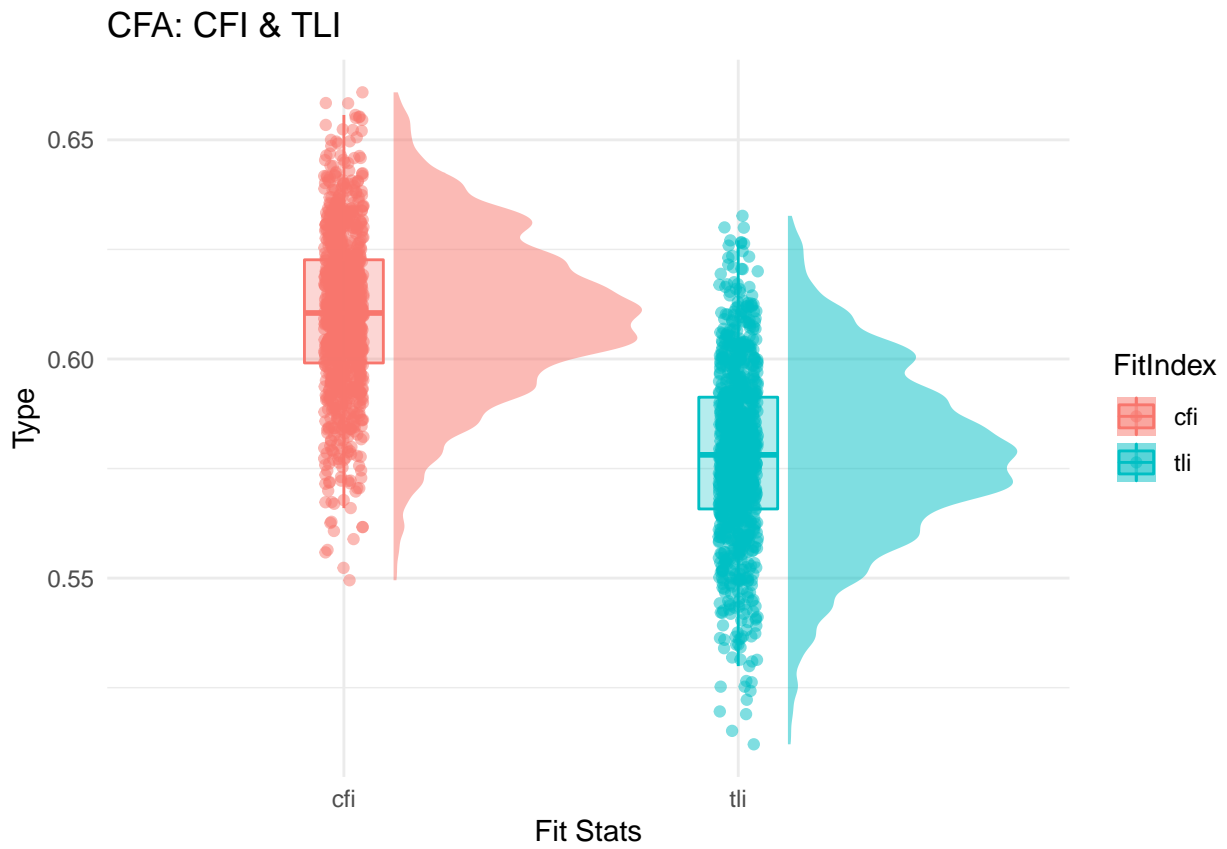
plt4 = resampled_res %>%
  ggplot(aes(x = "", y = Factors)) +
  ggdist::stat_halfeye(adjust = .5, width = .7, .width = 0, justification = -.2,
                      point_colour = NA, alpha = .5) +
  geom_jitter(width = .05, alpha = .5) +
  theme_minimal()+
  labs(
    title = 'EFA: Parallel Analysis Recommended Factors',
    subtitle = paste("Mean: ",avg," [Min: ", minimum, "Max:", maximum,"]"),
    x = 'Fit Stats',
    y = 'Type',
  )

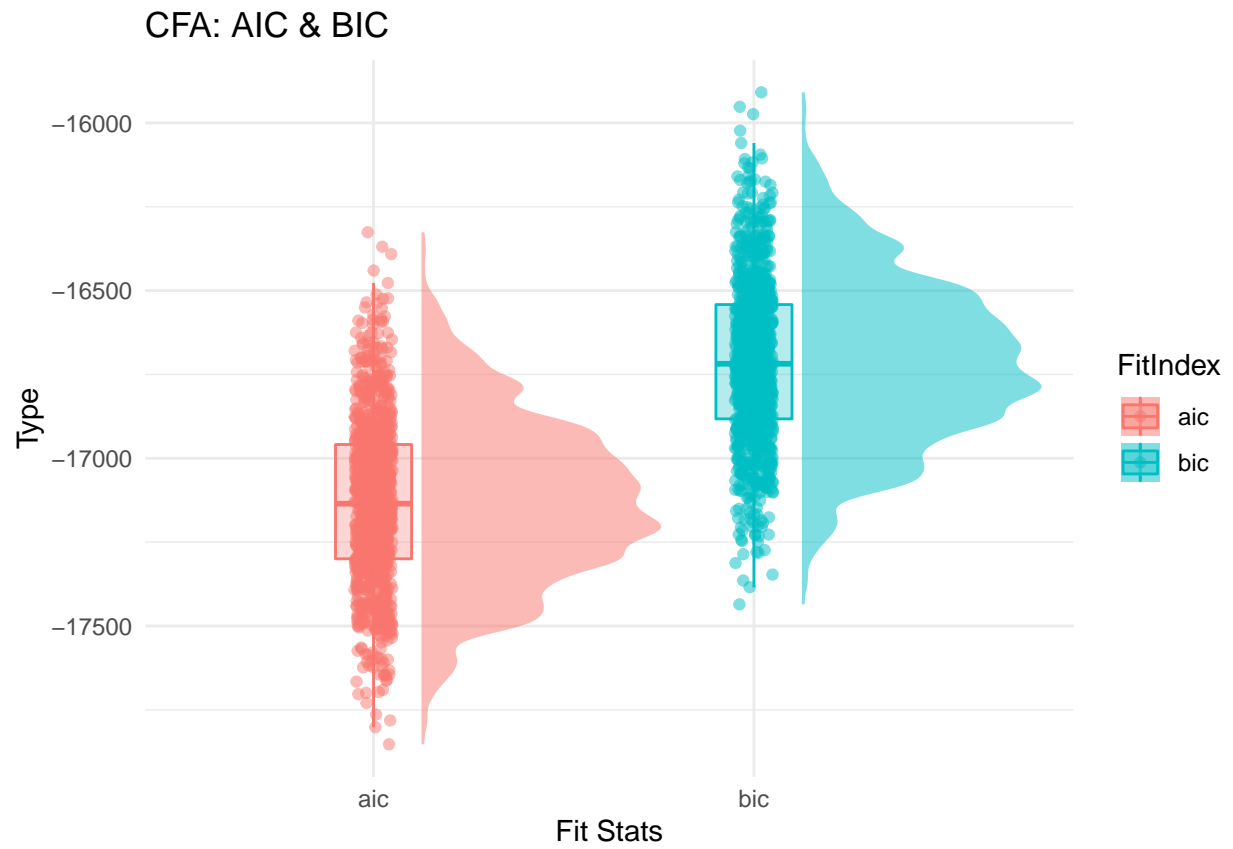
plt1;plt2;plt3;plt4

```


CFA: RMSEA & SRMR







EFA: Parallel Analysis Recommended Factors

Mean: 2 [Min: 1 Max: 2]

