

Joachim_CFA_report

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First the analysis of GPTS

```
## This is lavaan 0.6-8
## lavaan is FREE software! Please report any bugs.

##

## #####
```

recoding variables in dataset 2 and 3

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

First model, based on previous studies (pdf):

##	chisq.scaled	df.scaled	cfi.scaled	tli.scaled	rmsea.scaled	srmr
##	425.135	134.000	0.989	0.987	0.050	0.033

```
## For constructs with categorical indicators, the alpha and the average variance extracted are calculated
```

```
##           reference persecutory
## alpha  0.9451981   0.9715748
## omega  0.9160101   0.9466453
## omega2 0.9160101   0.9466453
## omega3 0.9231342   0.9575649
## avevar 0.6910939   0.7828572
```

```
##           lhs op           rhs      mi      epc sepc.lv sepc.all sepc.nox
## 159   reference =~ partB_gpts10 19.104 -0.303 -0.303 -0.303 -0.303
## 310 partB_gptsb5 ~~ partB_gpts10 15.416 0.082 0.082 0.386 0.386
## 312 partB_gptsb6 ~~ partB_gptsb8 11.965 0.060 0.060 0.369 0.369
## 154   reference =~ partB_gptsb5 11.773 -0.229 -0.229 -0.229 -0.229
## 170 PartA_gptsa1 ~~ PartA_gptsa4 10.947 0.076 0.076 0.303 0.303
## 151   reference =~ partB_gptsb2 10.527 0.224 0.224 0.224 0.224
## 218 PartA_gptsa4 ~~ PartA_gptsa7 9.837 -0.096 -0.096 -0.359 -0.359
## 319 partB_gptsb8 ~~ partB_gpts10 9.546 0.064 0.064 0.314 0.314
## 303 partB_gptsb4 ~~ partB_gptsb8 8.973 -0.093 -0.093 -0.507 -0.507
## 292 partB_gptsb2 ~~ partB_gpts10 8.514 -0.104 -0.104 -0.474 -0.474
```

Second model. Item 8a was removed based on fit indices from previous step of the analysis

```
## chisq.scaled  df.scaled  cfi.scaled  tli.scaled rmsea.scaled      srmr
##      389.246      118.000      0.989      0.987      0.052      0.033
```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```
##           reference persecutory
## alpha  0.9337281   0.9715748
## omega  0.8979067   0.9466392
## omega2 0.8979067   0.9466392
## omega3 0.9049826   0.9574355
## avevar 0.6758938   0.7826914
```

```
##           lhs op           rhs      mi      epc sepc.lv sepc.all sepc.nox
## 151   reference =~ partB_gpts10 16.552 -0.302 -0.302 -0.302 -0.302
## 158 persecutory =~ PartA_gptsa7 13.958 0.271 0.271 0.271 0.271
## 284 partB_gptsb5 ~~ partB_gpts10 13.726 0.078 0.078 0.373 0.373
## 286 partB_gptsb6 ~~ partB_gptsb8 12.704 0.063 0.063 0.383 0.383
## 161 PartA_gptsa1 ~~ PartA_gptsa4 10.058 0.075 0.075 0.302 0.302
## 146   reference =~ partB_gptsb5 9.594 -0.221 -0.221 -0.221 -0.221
## 143   reference =~ partB_gptsb2 9.517 0.225 0.225 0.225 0.225
## 277 partB_gptsb4 ~~ partB_gptsb8 8.992 -0.093 -0.093 -0.509 -0.509
## 293 partB_gptsb8 ~~ partB_gpts10 8.750 0.062 0.062 0.306 0.306
## 266 partB_gptsb2 ~~ partB_gpts10 8.670 -0.106 -0.106 -0.479 -0.479
```

Difference in robust chi-square test (see Szczypiński et al., 2021 section 2.3.2) between models 1 i 2

```
## [1] "chisq.diff: " "38.732"      "p value: "      "0.001"
```

Summary of model 2 GPTS

```
## lavaan 0.6-8 ended normally after 23 iterations
##
## Estimator DWLS
## Optimization method NLMINB
## Number of model parameters 86
##
## Used Total
## Number of observations 865 1827
##
## Model Test User Model:
## Standard Robust
## Test Statistic 200.770 389.246
## Degrees of freedom 118 118
## P-value (Chi-square) 0.000 0.000
## Scaling correction factor 0.569
## Shift parameter 36.569
## simple second-order correction
##
## Parameter Estimates:
## Standard errors Robust.sem
## Information Expected
## Information saturated (h1) model Unstructured
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## reference =~
## PartA_gptsa1 0.822 0.018 44.913 0.000 0.822 0.822
## PartA_gptsa2 0.815 0.021 38.637 0.000 0.815 0.815
## PartA_gptsa3 0.740 0.021 34.996 0.000 0.740 0.740
## PartA_gptsa4 0.901 0.012 72.640 0.000 0.901 0.901
## PartA_gptsa5 0.825 0.017 47.197 0.000 0.825 0.825
## PartA_gptsa6 0.859 0.015 57.079 0.000 0.859 0.859
## PartA_gptsa7 0.783 0.020 39.242 0.000 0.783 0.783
## persecutory =~
## partB_gptsb1 0.830 0.018 45.789 0.000 0.830 0.830
## partB_gptsb2 0.899 0.016 56.431 0.000 0.899 0.899
## partB_gptsb3 0.850 0.015 55.634 0.000 0.850 0.850
## partB_gptsb4 0.890 0.016 55.277 0.000 0.890 0.890
## partB_gptsb5 0.910 0.012 75.933 0.000 0.910 0.910
## partB_gptsb6 0.913 0.011 86.556 0.000 0.913 0.913
## partB_gptsb7 0.883 0.018 48.185 0.000 0.883 0.883
## partB_gptsb8 0.916 0.011 86.265 0.000 0.916 0.916
## partB_gptsb9 0.888 0.015 60.438 0.000 0.888 0.888
## partB_gpts10 0.864 0.015 57.097 0.000 0.864 0.864
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## reference ~~
## persecutory 0.865 0.015 57.259 0.000 0.865 0.865
##
## Intercepts:
```

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.PartA_gpts1	0.000				0.000	0.000
##	.PartA_gpts2	0.000				0.000	0.000
##	.PartA_gpts3	0.000				0.000	0.000
##	.PartA_gpts4	0.000				0.000	0.000
##	.PartA_gpts5	0.000				0.000	0.000
##	.PartA_gpts6	0.000				0.000	0.000
##	.PartA_gpts7	0.000				0.000	0.000
##	.partB_gpts1	0.000				0.000	0.000
##	.partB_gpts2	0.000				0.000	0.000
##	.partB_gpts3	0.000				0.000	0.000
##	.partB_gpts4	0.000				0.000	0.000
##	.partB_gpts5	0.000				0.000	0.000
##	.partB_gpts6	0.000				0.000	0.000
##	.partB_gpts7	0.000				0.000	0.000
##	.partB_gpts8	0.000				0.000	0.000
##	.partB_gpts9	0.000				0.000	0.000
##	.partB_gpts10	0.000				0.000	0.000
##	reference	0.000				0.000	0.000
##	persecutory	0.000				0.000	0.000
##							
##	Thresholds:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	PartA_gpts1 t1	0.477	0.044	10.727	0.000	0.477	0.477
##	PartA_gpts1 t2	1.016	0.052	19.653	0.000	1.016	1.016
##	PartA_gpts1 t3	1.369	0.061	22.491	0.000	1.369	1.369
##	PartA_gpts1 t4	1.847	0.083	22.228	0.000	1.847	1.847
##	PartA_gpts2 t1	0.632	0.046	13.781	0.000	0.632	0.632
##	PartA_gpts2 t2	1.180	0.055	21.299	0.000	1.180	1.180
##	PartA_gpts2 t3	1.615	0.070	22.912	0.000	1.615	1.615
##	PartA_gpts2 t4	2.086	0.101	20.616	0.000	2.086	2.086
##	PartA_gpts3 t1	-0.025	0.043	-0.578	0.563	-0.025	-0.025
##	PartA_gpts3 t2	0.480	0.044	10.794	0.000	0.480	0.480
##	PartA_gpts3 t3	0.871	0.049	17.748	0.000	0.871	0.871
##	PartA_gpts3 t4	1.439	0.063	22.738	0.000	1.439	1.439
##	PartA_gpts4 t1	0.422	0.044	9.586	0.000	0.422	0.422
##	PartA_gpts4 t2	0.875	0.049	17.810	0.000	0.875	0.875
##	PartA_gpts4 t3	1.259	0.058	21.891	0.000	1.259	1.259
##	PartA_gpts4 t4	1.671	0.073	22.838	0.000	1.671	1.671
##	PartA_gpts5 t1	0.320	0.043	7.361	0.000	0.320	0.320
##	PartA_gpts5 t2	0.716	0.047	15.280	0.000	0.716	0.716
##	PartA_gpts5 t3	1.055	0.052	20.101	0.000	1.055	1.055
##	PartA_gpts5 t4	1.544	0.067	22.918	0.000	1.544	1.544
##	PartA_gpts6 t1	0.329	0.043	7.564	0.000	0.329	0.329
##	PartA_gpts6 t2	0.879	0.049	17.871	0.000	0.879	0.879
##	PartA_gpts6 t3	1.259	0.058	21.891	0.000	1.259	1.259
##	PartA_gpts6 t4	1.707	0.075	22.758	0.000	1.707	1.707
##	PartA_gpts7 t1	-0.221	0.043	-5.127	0.000	-0.221	-0.221
##	PartA_gpts7 t2	0.191	0.043	4.449	0.000	0.191	0.191
##	PartA_gpts7 t3	0.608	0.046	13.320	0.000	0.608	0.608
##	PartA_gpts7 t4	1.151	0.055	21.053	0.000	1.151	1.151
##	prtB_gpts1 t1	0.556	0.045	12.327	0.000	0.556	0.556
##	prtB_gpts1 t2	1.045	0.052	19.991	0.000	1.045	1.045
##	prtB_gpts1 t3	1.384	0.061	22.552	0.000	1.384	1.384

##	prtB_gptsb1 t4	1.787	0.079	22.498	0.000	1.787	1.787
##	prtB_gptsb2 t1	1.001	0.051	19.481	0.000	1.001	1.001
##	prtB_gptsb2 t2	1.414	0.062	22.664	0.000	1.414	1.414
##	prtB_gptsb2 t3	1.694	0.074	22.788	0.000	1.694	1.694
##	prtB_gptsb2 t4	2.061	0.099	20.820	0.000	2.061	2.061
##	prtB_gptsb3 t1	0.448	0.044	10.123	0.000	0.448	0.448
##	prtB_gptsb3 t2	0.809	0.048	16.813	0.000	0.809	0.809
##	prtB_gptsb3 t3	1.180	0.055	21.299	0.000	1.180	1.180
##	prtB_gptsb3 t4	1.682	0.074	22.815	0.000	1.682	1.682
##	prtB_gptsb4 t1	1.016	0.052	19.653	0.000	1.016	1.016
##	prtB_gptsb4 t2	1.326	0.059	22.286	0.000	1.326	1.326
##	prtB_gptsb4 t3	1.626	0.071	22.902	0.000	1.626	1.626
##	prtB_gptsb4 t4	1.915	0.088	21.853	0.000	1.915	1.915
##	prtB_gptsb5 t1	0.789	0.048	16.497	0.000	0.789	0.789
##	prtB_gptsb5 t2	1.086	0.053	20.429	0.000	1.086	1.086
##	prtB_gptsb5 t3	1.376	0.061	22.522	0.000	1.376	1.376
##	prtB_gptsb5 t4	1.773	0.079	22.552	0.000	1.773	1.773
##	prtB_gptsb6 t1	0.590	0.045	12.990	0.000	0.590	0.590
##	prtB_gptsb6 t2	0.945	0.050	18.779	0.000	0.945	0.945
##	prtB_gptsb6 t3	1.285	0.058	22.057	0.000	1.285	1.285
##	prtB_gptsb6 t4	1.682	0.074	22.815	0.000	1.682	1.682
##	prtB_gptsb7 t1	1.113	0.054	20.694	0.000	1.113	1.113
##	prtB_gptsb7 t2	1.422	0.063	22.690	0.000	1.422	1.422
##	prtB_gptsb7 t3	1.732	0.076	22.688	0.000	1.732	1.732
##	prtB_gptsb7 t4	2.140	0.106	20.149	0.000	2.140	2.140
##	prtB_gptsb8 t1	0.601	0.046	13.188	0.000	0.601	0.601
##	prtB_gptsb8 t2	0.897	0.049	18.116	0.000	0.897	0.897
##	prtB_gptsb8 t3	1.197	0.056	21.443	0.000	1.197	1.197
##	prtB_gptsb8 t4	1.564	0.068	22.926	0.000	1.564	1.564
##	prtB_gptsb9 t1	0.739	0.047	15.667	0.000	0.739	0.739
##	prtB_gptsb9 t2	1.140	0.054	20.952	0.000	1.140	1.140
##	prtB_gptsb9 t3	1.573	0.069	22.928	0.000	1.573	1.573
##	prtB_gptsb9 t4	1.915	0.088	21.853	0.000	1.915	1.915
##	prtB_gpts10 t1	0.675	0.046	14.566	0.000	0.675	0.675
##	prtB_gpts10 t2	0.973	0.051	19.133	0.000	0.973	0.973
##	prtB_gpts10 t3	1.333	0.060	22.322	0.000	1.333	1.333
##	prtB_gpts10 t4	1.648	0.072	22.876	0.000	1.648	1.648

##

Variances:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.PartA_gptsa1	0.324				0.324	0.324
##	.PartA_gptsa2	0.336				0.336	0.336
##	.PartA_gptsa3	0.452				0.452	0.452
##	.PartA_gptsa4	0.189				0.189	0.189
##	.PartA_gptsa5	0.319				0.319	0.319
##	.PartA_gptsa6	0.262				0.262	0.262
##	.PartA_gptsa7	0.387				0.387	0.387
##	.partB_gptsb1	0.311				0.311	0.311
##	.partB_gptsb2	0.191				0.191	0.191
##	.partB_gptsb3	0.278				0.278	0.278
##	.partB_gptsb4	0.208				0.208	0.208
##	.partB_gptsb5	0.172				0.172	0.172
##	.partB_gptsb6	0.166				0.166	0.166
##	.partB_gptsb7	0.220				0.220	0.220

```

##      .partB_gptsb8      0.160      0.160      0.160
##      .partB_gptsb9      0.212      0.212      0.212
##      .partB_gpts10     0.254      0.254      0.254
##      reference          1.000      1.000      1.000
##      persecutory        1.000      1.000      1.000
##
## Scales y*:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      PartA_gptsa1      1.000      1.000      1.000      1.000
##      PartA_gptsa2      1.000      1.000      1.000      1.000
##      PartA_gptsa3      1.000      1.000      1.000      1.000
##      PartA_gptsa4      1.000      1.000      1.000      1.000
##      PartA_gptsa5      1.000      1.000      1.000      1.000
##      PartA_gptsa6      1.000      1.000      1.000      1.000
##      PartA_gptsa7      1.000      1.000      1.000      1.000
##      partB_gptsb1      1.000      1.000      1.000      1.000
##      partB_gptsb2      1.000      1.000      1.000      1.000
##      partB_gptsb3      1.000      1.000      1.000      1.000
##      partB_gptsb4      1.000      1.000      1.000      1.000
##      partB_gptsb5      1.000      1.000      1.000      1.000
##      partB_gptsb6      1.000      1.000      1.000      1.000
##      partB_gptsb7      1.000      1.000      1.000      1.000
##      partB_gptsb8      1.000      1.000      1.000      1.000
##      partB_gptsb9      1.000      1.000      1.000      1.000
##      partB_gpts10      1.000      1.000      1.000      1.000

```

summary of fit measures for three models of GPTSA

deleted models 3 to 5 since they did not differ significantly from model 2

```

##      chisq.scaled      df.scaled chisq.scaling.factor
##      425.135      134.000      0.585
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.989      0.987      0.050
##      srmr
##      0.033

##      chisq.scaled      df.scaled chisq.scaling.factor
##      389.246      118.000      0.569
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.989      0.987      0.052
##      srmr
##      0.033

```

MUSEQ is next

Model 1

```

##      chisq.scaled      df.scaled chisq.scaling.factor
##      2930.988      804.000      1.128
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.899      0.892      0.062

```

```
##          srmr
##          0.063
```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```
##          auditory    visual olfactory gustatory    bodily    presence
## alpha  0.8344134 0.8650326 0.8954536 0.9077042 0.9057060 0.8694990
## omega  0.8033959 0.8321705 0.8673721 0.8772246 0.8841962 0.8351232
## omega2 0.8033959 0.8321705 0.8673721 0.8772246 0.8841962 0.8351232
## omega3 0.8236923 0.8596939 0.8992562 0.9185328 0.9074923 0.8521143
## avevar 0.4381763 0.4620095 0.5780030 0.5811473 0.5702036 0.6450147
```

```
##          lhs op      rhs      mi    epc sepc.lv sepc.all sepc.nox
## 1141  MUSEQ3_3 ~~ MUSEQ4_3 210.112 0.323  0.323    0.735    0.735
## 395   auditory == MUSEQ6_1  94.638 0.439  0.439    0.439    0.439
## 661   MUSEQ3_3 ~~ MUSEQ2_3  85.775 0.301  0.301    0.476    0.476
## 1204  MUSEQ3_6 ~~ MUSEQ3_7  83.931 0.217  0.217    0.613    0.613
## 477   gustatory == MUSEQ2_3  82.367 0.392  0.392    0.392    0.392
## 450   olfactory == MUSEQ4_3  80.065 0.612  0.612    0.612    0.612
## 429   visual == MUSEQ6_1  77.930 0.431  0.431    0.431    0.431
## 532   bodily == MUSEQ6_1  71.718 0.360  0.360    0.360    0.360
## 442   olfactory == MUSEQ2_3  68.029 0.319  0.319    0.319    0.319
## 1432  MUSEQ6_2 ~~ MUSEQ6_3  67.581 0.279  0.279    0.663    0.663
```

Model 2 without items 3.3 i 4.3

```
##          chisq.scaled      df.scaled chisq.scaling.factor
##          2374.362          725.000          1.082
##          cfi.scaled      tli.scaled      rmsea.scaled
##          0.916          0.910          0.057
##          srmr
##          0.060
```

```
##          lhs op      rhs      mi    epc sepc.lv sepc.all sepc.nox
## 631   MUSEQ3_3 ~~ MUSEQ2_3 100.753 0.326  0.326    0.492    0.492
## 377   auditory == MUSEQ6_1  99.648 0.456  0.456    0.456    0.456
## 409   visual == MUSEQ6_1  84.801 0.458  0.458    0.458    0.458
## 508   bodily == MUSEQ6_1  76.303 0.376  0.376    0.376    0.376
## 1325  MUSEQ6_2 ~~ MUSEQ6_3  67.317 0.279  0.279    0.666    0.666
## 1118  MUSEQ3_6 ~~ MUSEQ3_7  66.341 0.201  0.201    0.599    0.599
## 1313  MUSEQ5_7 ~~ MUSEQ5_8  55.788 0.190  0.190    0.502    0.502
## 662   MUSEQ4_4 ~~ MUSEQ5_5  52.986 0.211  0.211    0.331    0.331
## 547   presence == MUSEQ5_8  50.197 0.283  0.283    0.283    0.283
## 455   gustatory == MUSEQ2_2  49.646 0.296  0.296    0.296    0.296
```

Comparison between models 1 and 2 of MUSEQ

```
## [1] "chisq.diff: " "475.027"      "p value: "      "0"
```

Model 3 without item 6.1

```
## chisq.scaled    df.scaled    cfi.scaled    tli.scaled rmsea.scaled      srmr
##      2185.913      687.000        0.922        0.916        0.056      0.058
```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```
##      auditory    visual olfactory gustatory    bodily    presence
## alpha 0.8344134 0.8650326 0.8894462 0.9066283 0.9057060 0.8493407
## omega 0.8027678 0.8310620 0.8565348 0.8722275 0.8840501 0.7784855
## omega2 0.8027678 0.8310620 0.8565348 0.8722275 0.8840501 0.7784855
## omega3 0.8219198 0.8556483 0.8827242 0.8989991 0.9071091 0.7935248
## avevar 0.4378164 0.4619508 0.5986320 0.6068100 0.5702232 0.6753243
```

Comparison between models 2 and 3 of MUSEQ

```
## [1] "chisq.diff: " "166.467"      "p value: "      "0"
```

Summary of model 3 MUSEQ

```
## lavaan 0.6-8 ended normally after 31 iterations
##
##      Estimator                      DWLS
##      Optimization method          NLMINB
##      Number of model parameters      210
##
##      Number of observations          692
##
## Model Test User Model:
##
##      Test Statistic          Standard      Robust
##      Degrees of freedom          687          687
##      P-value (Chi-square)          0.000          0.000
##      Scaling correction factor          1.053
##      Shift parameter          322.349
##      simple second-order correction
##
## Parameter Estimates:
##
##      Standard errors          Robust.sem
##      Information          Expected
##      Information saturated (h1) model      Unstructured
##
## Latent Variables:
##
##      Estimate    Std.Err    z-value    P(>|z|)    Std.lv    Std.all
##      auditory =~
##      MUSEQ_1          0.668    0.027    24.872    0.000    0.668    0.668
##      MUSEQ_2          0.652    0.030    21.741    0.000    0.652    0.652
##      MUSEQ_3          0.590    0.036    16.223    0.000    0.590    0.590
##      MUSEQ_4          0.633    0.030    21.419    0.000    0.633    0.633
##      MUSEQ_5          0.567    0.031    18.257    0.000    0.567    0.567
##      MUSEQ_6          0.693    0.027    25.731    0.000    0.693    0.693
```


##	MUSEQ_7	0.801	0.022	35.991	0.000	0.801	0.801
##	visual =~						
##	MUSEQ2_1	0.724	0.023	31.987	0.000	0.724	0.724
##	MUSEQ2_2	0.716	0.027	26.998	0.000	0.716	0.716
##	MUSEQ2_3	0.578	0.031	18.895	0.000	0.578	0.578
##	MUSEQ2_4	0.570	0.030	18.766	0.000	0.570	0.570
##	MUSEQ2_5	0.726	0.026	28.266	0.000	0.726	0.726
##	MUSEQ2_6	0.664	0.035	18.895	0.000	0.664	0.664
##	MUSEQ2_7	0.711	0.024	29.741	0.000	0.711	0.711
##	MUSEQ2_8	0.726	0.025	29.608	0.000	0.726	0.726
##	olfactory =~						
##	MUSEQ3_1	0.758	0.022	33.828	0.000	0.758	0.758
##	MUSEQ3_2	0.746	0.024	30.618	0.000	0.746	0.746
##	MUSEQ3_4	0.768	0.028	26.995	0.000	0.768	0.768
##	MUSEQ3_5	0.735	0.024	31.144	0.000	0.735	0.735
##	MUSEQ3_6	0.822	0.020	41.110	0.000	0.822	0.822
##	MUSEQ3_7	0.810	0.021	38.924	0.000	0.810	0.810
##	gustatory =~						
##	MUSEQ4_1	0.751	0.021	35.901	0.000	0.751	0.751
##	MUSEQ4_2	0.789	0.020	39.805	0.000	0.789	0.789
##	MUSEQ4_4	0.706	0.025	27.818	0.000	0.706	0.706
##	MUSEQ4_5	0.790	0.023	34.995	0.000	0.790	0.790
##	MUSEQ4_6	0.800	0.019	42.303	0.000	0.800	0.800
##	MUSEQ4_7	0.778	0.024	32.816	0.000	0.778	0.778
##	MUSEQ4_8	0.832	0.025	32.835	0.000	0.832	0.832
##	bodily =~						
##	MUSEQ5_1	0.787	0.019	42.413	0.000	0.787	0.787
##	MUSEQ5_2	0.803	0.017	46.823	0.000	0.803	0.803
##	MUSEQ5_3	0.765	0.021	35.991	0.000	0.765	0.765
##	MUSEQ5_4	0.694	0.030	23.397	0.000	0.694	0.694
##	MUSEQ5_5	0.743	0.021	35.028	0.000	0.743	0.743
##	MUSEQ5_6	0.672	0.025	27.270	0.000	0.672	0.672
##	MUSEQ5_7	0.795	0.021	38.535	0.000	0.795	0.795
##	MUSEQ5_8	0.771	0.024	32.742	0.000	0.771	0.771
##	presence =~						
##	MUSEQ6_2	0.792	0.036	22.083	0.000	0.792	0.792
##	MUSEQ6_3	0.870	0.039	22.511	0.000	0.870	0.870
##	MUSEQ6_4	0.802	0.042	18.947	0.000	0.802	0.802
##							
##	Covariances:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	auditory ~~						
##	visual	0.811	0.022	37.405	0.000	0.811	0.811
##	olfactory	0.580	0.034	16.836	0.000	0.580	0.580
##	gustatory	0.660	0.029	22.626	0.000	0.660	0.660
##	bodily	0.740	0.024	30.968	0.000	0.740	0.740
##	presence	0.406	0.043	9.352	0.000	0.406	0.406
##	visual ~~						
##	olfactory	0.635	0.032	19.789	0.000	0.635	0.635
##	gustatory	0.675	0.029	23.132	0.000	0.675	0.675
##	bodily	0.761	0.025	30.880	0.000	0.761	0.761
##	presence	0.516	0.040	12.787	0.000	0.516	0.516
##	olfactory ~~						
##	gustatory	0.787	0.022	35.367	0.000	0.787	0.787

##	bodily	0.654	0.029	22.568	0.000	0.654	0.654
##	presence	0.401	0.046	8.646	0.000	0.401	0.401
##	gustatory ~~						
##	bodily	0.704	0.026	27.072	0.000	0.704	0.704
##	presence	0.426	0.046	9.163	0.000	0.426	0.426
##	bodily ~~						
##	presence	0.446	0.040	11.112	0.000	0.446	0.446
##							
##	Intercepts:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.MUSEQ_1	0.000				0.000	0.000
##	.MUSEQ_2	0.000				0.000	0.000
##	.MUSEQ_3	0.000				0.000	0.000
##	.MUSEQ_4	0.000				0.000	0.000
##	.MUSEQ_5	0.000				0.000	0.000
##	.MUSEQ_6	0.000				0.000	0.000
##	.MUSEQ_7	0.000				0.000	0.000
##	.MUSEQ2_1	0.000				0.000	0.000
##	.MUSEQ2_2	0.000				0.000	0.000
##	.MUSEQ2_3	0.000				0.000	0.000
##	.MUSEQ2_4	0.000				0.000	0.000
##	.MUSEQ2_5	0.000				0.000	0.000
##	.MUSEQ2_6	0.000				0.000	0.000
##	.MUSEQ2_7	0.000				0.000	0.000
##	.MUSEQ2_8	0.000				0.000	0.000
##	.MUSEQ3_1	0.000				0.000	0.000
##	.MUSEQ3_2	0.000				0.000	0.000
##	.MUSEQ3_4	0.000				0.000	0.000
##	.MUSEQ3_5	0.000				0.000	0.000
##	.MUSEQ3_6	0.000				0.000	0.000
##	.MUSEQ3_7	0.000				0.000	0.000
##	.MUSEQ4_1	0.000				0.000	0.000
##	.MUSEQ4_2	0.000				0.000	0.000
##	.MUSEQ4_4	0.000				0.000	0.000
##	.MUSEQ4_5	0.000				0.000	0.000
##	.MUSEQ4_6	0.000				0.000	0.000
##	.MUSEQ4_7	0.000				0.000	0.000
##	.MUSEQ4_8	0.000				0.000	0.000
##	.MUSEQ5_1	0.000				0.000	0.000
##	.MUSEQ5_2	0.000				0.000	0.000
##	.MUSEQ5_3	0.000				0.000	0.000
##	.MUSEQ5_4	0.000				0.000	0.000
##	.MUSEQ5_5	0.000				0.000	0.000
##	.MUSEQ5_6	0.000				0.000	0.000
##	.MUSEQ5_7	0.000				0.000	0.000
##	.MUSEQ5_8	0.000				0.000	0.000
##	.MUSEQ6_2	0.000				0.000	0.000
##	.MUSEQ6_3	0.000				0.000	0.000
##	.MUSEQ6_4	0.000				0.000	0.000
##	auditory	0.000				0.000	0.000
##	visual	0.000				0.000	0.000
##	olfactory	0.000				0.000	0.000
##	gustatory	0.000				0.000	0.000
##	bodily	0.000				0.000	0.000

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##      presence      0.000      0.000      0.000
##
## Thresholds:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      MUSEQ_1|t1 -0.788  0.053 -14.739  0.000 -0.788 -0.788
##      MUSEQ_1|t2 -0.051  0.048  -1.064  0.288 -0.051 -0.051
##      MUSEQ_1|t3  0.527  0.050  10.504  0.000  0.527  0.527
##      MUSEQ_1|t4  1.275  0.065  19.671  0.000  1.275  1.275
##      MUSEQ_2|t1 -0.268  0.048  -5.540  0.000 -0.268 -0.268
##      MUSEQ_2|t2  0.193  0.048   4.024  0.000  0.193  0.193
##      MUSEQ_2|t3  0.478  0.050   9.608  0.000  0.478  0.478
##      MUSEQ_2|t4  1.099  0.060  18.389  0.000  1.099  1.099
##      MUSEQ_3|t1 -1.023  0.058 -17.651  0.000 -1.023 -1.023
##      MUSEQ_3|t2 -0.674  0.052 -13.012  0.000 -0.674 -0.674
##      MUSEQ_3|t3 -0.398  0.049  -8.107  0.000 -0.398 -0.398
##      MUSEQ_3|t4  0.309  0.049   6.372  0.000  0.309  0.309
##      MUSEQ_4|t1 -0.854  0.055 -15.651  0.000 -0.854 -0.854
##      MUSEQ_4|t2  0.018  0.048   0.380  0.704  0.018  0.018
##      MUSEQ_4|t3  0.656  0.052  12.720  0.000  0.656  0.656
##      MUSEQ_4|t4  1.762  0.087  20.202  0.000  1.762  1.762
##      MUSEQ_5|t1 -0.730  0.053 -13.881  0.000 -0.730 -0.730
##      MUSEQ_5|t2 -0.011  0.048  -0.228  0.820 -0.011 -0.011
##      MUSEQ_5|t3  0.548  0.050  10.876  0.000  0.548  0.548
##      MUSEQ_5|t4  1.439  0.071  20.334  0.000  1.439  1.439
##      MUSEQ_6|t1 -0.355  0.049  -7.278  0.000 -0.355 -0.355
##      MUSEQ_6|t2  0.305  0.048   6.296  0.000  0.305  0.305
##      MUSEQ_6|t3  0.749  0.053  14.169  0.000  0.749  0.749
##      MUSEQ_6|t4  1.418  0.070  20.280  0.000  1.418  1.418
##      MUSEQ_7|t1 -0.087  0.048  -1.823  0.068 -0.087 -0.087
##      MUSEQ_7|t2  0.754  0.053  14.240  0.000  0.754  0.754
##      MUSEQ_7|t3  1.251  0.064  19.530  0.000  1.251  1.251
##      MUSEQ_7|t4  2.020  0.107  18.903  0.000  2.020  2.020
##      MUSEQ2_1|t1 -0.849  0.054 -15.582  0.000 -0.849 -0.849
##      MUSEQ2_1|t2 -0.040  0.048  -0.836  0.403 -0.040 -0.040
##      MUSEQ2_1|t3  0.552  0.050  10.950  0.000  0.552  0.552
##      MUSEQ2_1|t4  1.537  0.075  20.491  0.000  1.537  1.537
##      MUSEQ2_2|t1 -0.167  0.048  -3.493  0.000 -0.167 -0.167
##      MUSEQ2_2|t2  0.418  0.049   8.483  0.000  0.418  0.418
##      MUSEQ2_2|t3  0.870  0.055  15.859  0.000  0.870  0.870
##      MUSEQ2_2|t4  1.561  0.076  20.502  0.000  1.561  1.561
##      MUSEQ2_3|t1 -0.544  0.050 -10.801  0.000 -0.544 -0.544
##      MUSEQ2_3|t2  0.043  0.048   0.912  0.362  0.043  0.043
##      MUSEQ2_3|t3  0.378  0.049   7.730  0.000  0.378  0.378
##      MUSEQ2_3|t4  0.913  0.056  16.405  0.000  0.913  0.913
##      MUSEQ2_4|t1 -0.778  0.053 -14.597  0.000 -0.778 -0.778
##      MUSEQ2_4|t2 -0.123  0.048  -2.583  0.010 -0.123 -0.123
##      MUSEQ2_4|t3  0.283  0.048   5.842  0.000  0.283  0.283
##      MUSEQ2_4|t4  0.941  0.056  16.740  0.000  0.941  0.941
##      MUSEQ2_5|t1  0.069  0.048   1.443  0.149  0.069  0.069
##      MUSEQ2_5|t2  0.625  0.051  12.206  0.000  0.625  0.625
##      MUSEQ2_5|t3  1.011  0.058  17.524  0.000  1.011  1.011
##      MUSEQ2_5|t4  1.816  0.091  20.009  0.000  1.816  1.816
##      MUSEQ2_6|t1  0.612  0.051  11.986  0.000  0.612  0.612
##      MUSEQ2_6|t2  1.119  0.060  18.565  0.000  1.119  1.119

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##	MUSEQ2_6 t3	1.525	0.074	20.482	0.000	1.525	1.525
##	MUSEQ2_6 t4	2.147	0.120	17.961	0.000	2.147	2.147
##	MUSEQ2_7 t1	-0.946	0.056	-16.807	0.000	-0.946	-0.946
##	MUSEQ2_7 t2	-0.193	0.048	-4.024	0.000	-0.193	-0.193
##	MUSEQ2_7 t3	0.398	0.049	8.107	0.000	0.398	0.398
##	MUSEQ2_7 t4	1.106	0.060	18.448	0.000	1.106	1.106
##	MUSEQ2_8 t1	0.029	0.048	0.608	0.543	0.029	0.029
##	MUSEQ2_8 t2	0.716	0.052	13.665	0.000	0.716	0.716
##	MUSEQ2_8 t3	1.183	0.062	19.070	0.000	1.183	1.183
##	MUSEQ2_8 t4	1.897	0.097	19.637	0.000	1.897	1.897
##	MUSEQ3_1 t1	-0.227	0.048	-4.707	0.000	-0.227	-0.227
##	MUSEQ3_1 t2	0.519	0.050	10.355	0.000	0.519	0.519
##	MUSEQ3_1 t3	0.993	0.057	17.331	0.000	0.993	0.993
##	MUSEQ3_1 t4	1.855	0.093	19.842	0.000	1.855	1.855
##	MUSEQ3_2 t1	-0.080	0.048	-1.671	0.095	-0.080	-0.080
##	MUSEQ3_2 t2	0.433	0.049	8.783	0.000	0.433	0.433
##	MUSEQ3_2 t3	0.788	0.053	14.739	0.000	0.788	0.788
##	MUSEQ3_2 t4	1.418	0.070	20.280	0.000	1.418	1.418
##	MUSEQ3_4 t1	0.556	0.050	11.024	0.000	0.556	0.556
##	MUSEQ3_4 t2	1.113	0.060	18.507	0.000	1.113	1.113
##	MUSEQ3_4 t3	1.514	0.074	20.470	0.000	1.514	1.514
##	MUSEQ3_4 t4	2.226	0.129	17.296	0.000	2.226	2.226
##	MUSEQ3_5 t1	-0.054	0.048	-1.140	0.254	-0.054	-0.054
##	MUSEQ3_5 t2	0.674	0.052	13.012	0.000	0.674	0.674
##	MUSEQ3_5 t3	1.119	0.060	18.565	0.000	1.119	1.119
##	MUSEQ3_5 t4	2.049	0.110	18.705	0.000	2.049	2.049
##	MUSEQ3_6 t1	0.260	0.048	5.388	0.000	0.260	0.260
##	MUSEQ3_6 t2	0.854	0.055	15.651	0.000	0.854	0.854
##	MUSEQ3_6 t3	1.291	0.065	19.761	0.000	1.291	1.291
##	MUSEQ3_6 t4	1.967	0.102	19.242	0.000	1.967	1.967
##	MUSEQ3_7 t1	0.256	0.048	5.313	0.000	0.256	0.256
##	MUSEQ3_7 t2	0.854	0.055	15.651	0.000	0.854	0.854
##	MUSEQ3_7 t3	1.259	0.064	19.578	0.000	1.259	1.259
##	MUSEQ3_7 t4	2.049	0.110	18.705	0.000	2.049	2.049
##	MUSEQ4_1 t1	-0.036	0.048	-0.760	0.447	-0.036	-0.036
##	MUSEQ4_1 t2	0.886	0.055	16.065	0.000	0.886	0.886
##	MUSEQ4_1 t3	1.491	0.073	20.441	0.000	1.491	1.491
##	MUSEQ4_1 t4	2.379	0.150	15.871	0.000	2.379	2.379
##	MUSEQ4_2 t1	0.040	0.048	0.836	0.403	0.040	0.040
##	MUSEQ4_2 t2	0.688	0.052	13.230	0.000	0.688	0.688
##	MUSEQ4_2 t3	1.175	0.062	19.015	0.000	1.175	1.175
##	MUSEQ4_2 t4	1.993	0.104	19.081	0.000	1.993	1.993
##	MUSEQ4_4 t1	-0.215	0.048	-4.479	0.000	-0.215	-0.215
##	MUSEQ4_4 t2	0.347	0.049	7.127	0.000	0.347	0.347
##	MUSEQ4_4 t3	0.902	0.055	16.269	0.000	0.902	0.902
##	MUSEQ4_4 t4	1.780	0.088	20.145	0.000	1.780	1.780
##	MUSEQ4_5 t1	0.268	0.048	5.540	0.000	0.268	0.268
##	MUSEQ4_5 t2	0.969	0.057	17.071	0.000	0.969	0.969
##	MUSEQ4_5 t3	1.503	0.073	20.457	0.000	1.503	1.503
##	MUSEQ4_5 t4	2.322	0.141	16.421	0.000	2.322	2.322
##	MUSEQ4_6 t1	0.171	0.048	3.569	0.000	0.171	0.171
##	MUSEQ4_6 t2	0.778	0.053	14.597	0.000	0.778	0.778
##	MUSEQ4_6 t3	1.212	0.063	19.280	0.000	1.212	1.212
##	MUSEQ4_6 t4	2.185	0.124	17.650	0.000	2.185	2.185

##	MUSEQ4_7 t1	0.465	0.050	9.383	0.000	0.465	0.465
##	MUSEQ4_7 t2	1.023	0.058	17.651	0.000	1.023	1.023
##	MUSEQ4_7 t3	1.561	0.076	20.502	0.000	1.561	1.561
##	MUSEQ4_7 t4	2.322	0.141	16.421	0.000	2.322	2.322
##	MUSEQ4_8 t1	0.764	0.053	14.383	0.000	0.764	0.764
##	MUSEQ4_8 t2	1.389	0.069	20.188	0.000	1.389	1.389
##	MUSEQ4_8 t3	1.835	0.092	19.930	0.000	1.835	1.835
##	MUSEQ4_8 t4	2.525	0.175	14.404	0.000	2.525	2.525
##	MUSEQ5_1 t1	-0.716	0.052	-13.665	0.000	-0.716	-0.716
##	MUSEQ5_1 t2	0.091	0.048	1.899	0.058	0.091	0.091
##	MUSEQ5_1 t3	0.603	0.051	11.838	0.000	0.603	0.603
##	MUSEQ5_1 t4	1.380	0.068	20.155	0.000	1.380	1.380
##	MUSEQ5_2 t1	-0.730	0.053	-13.881	0.000	-0.730	-0.730
##	MUSEQ5_2 t2	-0.138	0.048	-2.886	0.004	-0.138	-0.138
##	MUSEQ5_2 t3	0.371	0.049	7.579	0.000	0.371	0.371
##	MUSEQ5_2 t4	1.126	0.060	18.623	0.000	1.126	1.126
##	MUSEQ5_3 t1	-0.351	0.049	-7.202	0.000	-0.351	-0.351
##	MUSEQ5_3 t2	0.175	0.048	3.645	0.000	0.175	0.175
##	MUSEQ5_3 t3	0.625	0.051	12.206	0.000	0.625	0.625
##	MUSEQ5_3 t4	1.243	0.064	19.481	0.000	1.243	1.243
##	MUSEQ5_4 t1	0.414	0.049	8.408	0.000	0.414	0.414
##	MUSEQ5_4 t2	0.833	0.054	15.373	0.000	0.833	0.833
##	MUSEQ5_4 t3	1.212	0.063	19.280	0.000	1.212	1.212
##	MUSEQ5_4 t4	1.612	0.079	20.492	0.000	1.612	1.612
##	MUSEQ5_5 t1	-0.539	0.050	-10.727	0.000	-0.539	-0.539
##	MUSEQ5_5 t2	-0.014	0.048	-0.304	0.761	-0.014	-0.014
##	MUSEQ5_5 t3	0.457	0.050	9.233	0.000	0.457	0.457
##	MUSEQ5_5 t4	1.113	0.060	18.507	0.000	1.113	1.113
##	MUSEQ5_6 t1	-0.076	0.048	-1.595	0.111	-0.076	-0.076
##	MUSEQ5_6 t2	0.453	0.050	9.158	0.000	0.453	0.453
##	MUSEQ5_6 t3	0.849	0.054	15.582	0.000	0.849	0.849
##	MUSEQ5_6 t4	1.459	0.072	20.382	0.000	1.459	1.459
##	MUSEQ5_7 t1	0.080	0.048	1.671	0.095	0.080	0.080
##	MUSEQ5_7 t2	0.702	0.052	13.448	0.000	0.702	0.702
##	MUSEQ5_7 t3	1.086	0.059	18.269	0.000	1.086	1.086
##	MUSEQ5_7 t4	1.682	0.082	20.403	0.000	1.682	1.682
##	MUSEQ5_8 t1	0.309	0.049	6.372	0.000	0.309	0.309
##	MUSEQ5_8 t2	1.029	0.058	17.715	0.000	1.029	1.029
##	MUSEQ5_8 t3	1.343	0.067	20.010	0.000	1.343	1.343
##	MUSEQ5_8 t4	1.993	0.104	19.081	0.000	1.993	1.993
##	MUSEQ6_2 t1	0.490	0.050	9.832	0.000	0.490	0.490
##	MUSEQ6_2 t2	1.183	0.062	19.070	0.000	1.183	1.183
##	MUSEQ6_2 t3	1.561	0.076	20.502	0.000	1.561	1.561
##	MUSEQ6_2 t4	2.112	0.116	18.238	0.000	2.112	2.112
##	MUSEQ6_3 t1	0.730	0.053	13.881	0.000	0.730	0.730
##	MUSEQ6_3 t2	1.418	0.070	20.280	0.000	1.418	1.418
##	MUSEQ6_3 t3	1.780	0.088	20.145	0.000	1.780	1.780
##	MUSEQ6_3 t4	2.079	0.113	18.484	0.000	2.079	2.079
##	MUSEQ6_4 t1	0.441	0.049	8.933	0.000	0.441	0.441
##	MUSEQ6_4 t2	1.283	0.065	19.716	0.000	1.283	1.283
##	MUSEQ6_4 t3	1.668	0.082	20.429	0.000	1.668	1.668
##	MUSEQ6_4 t4	2.185	0.124	17.650	0.000	2.185	2.185

##

Variances:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.MUSEQ_1	0.554				0.554	0.554
##	.MUSEQ_2	0.575				0.575	0.575
##	.MUSEQ_3	0.651				0.651	0.651
##	.MUSEQ_4	0.599				0.599	0.599
##	.MUSEQ_5	0.678				0.678	0.678
##	.MUSEQ_6	0.519				0.519	0.519
##	.MUSEQ_7	0.358				0.358	0.358
##	.MUSEQ2_1	0.476				0.476	0.476
##	.MUSEQ2_2	0.488				0.488	0.488
##	.MUSEQ2_3	0.666				0.666	0.666
##	.MUSEQ2_4	0.675				0.675	0.675
##	.MUSEQ2_5	0.473				0.473	0.473
##	.MUSEQ2_6	0.559				0.559	0.559
##	.MUSEQ2_7	0.495				0.495	0.495
##	.MUSEQ2_8	0.473				0.473	0.473
##	.MUSEQ3_1	0.425				0.425	0.425
##	.MUSEQ3_2	0.444				0.444	0.444
##	.MUSEQ3_4	0.410				0.410	0.410
##	.MUSEQ3_5	0.459				0.459	0.459
##	.MUSEQ3_6	0.325				0.325	0.325
##	.MUSEQ3_7	0.345				0.345	0.345
##	.MUSEQ4_1	0.436				0.436	0.436
##	.MUSEQ4_2	0.378				0.378	0.378
##	.MUSEQ4_4	0.502				0.502	0.502
##	.MUSEQ4_5	0.375				0.375	0.375
##	.MUSEQ4_6	0.360				0.360	0.360
##	.MUSEQ4_7	0.395				0.395	0.395
##	.MUSEQ4_8	0.307				0.307	0.307
##	.MUSEQ5_1	0.381				0.381	0.381
##	.MUSEQ5_2	0.354				0.354	0.354
##	.MUSEQ5_3	0.414				0.414	0.414
##	.MUSEQ5_4	0.519				0.519	0.519
##	.MUSEQ5_5	0.447				0.447	0.447
##	.MUSEQ5_6	0.548				0.548	0.548
##	.MUSEQ5_7	0.368				0.368	0.368
##	.MUSEQ5_8	0.406				0.406	0.406
##	.MUSEQ6_2	0.373				0.373	0.373
##	.MUSEQ6_3	0.243				0.243	0.243
##	.MUSEQ6_4	0.357				0.357	0.357
##	auditory	1.000				1.000	1.000
##	visual	1.000				1.000	1.000
##	olfactory	1.000				1.000	1.000
##	gustatory	1.000				1.000	1.000
##	bodily	1.000				1.000	1.000
##	presence	1.000				1.000	1.000
##							
##	Scales y*:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	MUSEQ_1	1.000				1.000	1.000
##	MUSEQ_2	1.000				1.000	1.000
##	MUSEQ_3	1.000				1.000	1.000
##	MUSEQ_4	1.000				1.000	1.000
##	MUSEQ_5	1.000				1.000	1.000

##	MUSEQ_6	1.000	1.000	1.000
##	MUSEQ_7	1.000	1.000	1.000
##	MUSEQ2_1	1.000	1.000	1.000
##	MUSEQ2_2	1.000	1.000	1.000
##	MUSEQ2_3	1.000	1.000	1.000
##	MUSEQ2_4	1.000	1.000	1.000
##	MUSEQ2_5	1.000	1.000	1.000
##	MUSEQ2_6	1.000	1.000	1.000
##	MUSEQ2_7	1.000	1.000	1.000
##	MUSEQ2_8	1.000	1.000	1.000
##	MUSEQ3_1	1.000	1.000	1.000
##	MUSEQ3_2	1.000	1.000	1.000
##	MUSEQ3_4	1.000	1.000	1.000
##	MUSEQ3_5	1.000	1.000	1.000
##	MUSEQ3_6	1.000	1.000	1.000
##	MUSEQ3_7	1.000	1.000	1.000
##	MUSEQ4_1	1.000	1.000	1.000
##	MUSEQ4_2	1.000	1.000	1.000
##	MUSEQ4_4	1.000	1.000	1.000
##	MUSEQ4_5	1.000	1.000	1.000
##	MUSEQ4_6	1.000	1.000	1.000
##	MUSEQ4_7	1.000	1.000	1.000
##	MUSEQ4_8	1.000	1.000	1.000
##	MUSEQ5_1	1.000	1.000	1.000
##	MUSEQ5_2	1.000	1.000	1.000
##	MUSEQ5_3	1.000	1.000	1.000
##	MUSEQ5_4	1.000	1.000	1.000
##	MUSEQ5_5	1.000	1.000	1.000
##	MUSEQ5_6	1.000	1.000	1.000
##	MUSEQ5_7	1.000	1.000	1.000
##	MUSEQ5_8	1.000	1.000	1.000
##	MUSEQ6_2	1.000	1.000	1.000
##	MUSEQ6_3	1.000	1.000	1.000
##	MUSEQ6_4	1.000	1.000	1.000

summary of fit measures for three models of MUSEQ

##	chisq.scaled	df.scaled	chisq.scaling.factor
##	2930.988	804.000	1.128
##	cfi.scaled	tli.scaled	rmsea.scaled
##	0.899	0.892	0.062
##	srmr		
##	0.063		

##	chisq.scaled	df.scaled	chisq.scaling.factor
##	2374.362	725.000	1.082
##	cfi.scaled	tli.scaled	rmsea.scaled
##	0.916	0.910	0.057
##	srmr		
##	0.060		

##	chisq.scaled	df.scaled	chisq.scaling.factor
##	2185.913	687.000	1.053

```
##          cfi.scaled          tli.scaled          rmsea.scaled
##          0.922              0.916              0.056
##          srmr
##          0.058
```

Analiza BAPS

1st model

```
## chisq.scaled  df.scaled  cfi.scaled  tli.scaled  rmsea.scaled      srmr
##    1130.643    132.000      0.968      0.962      0.085      0.082
```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```
##          survival_strategy  negative_beliefs  normalizing_beliefs
## alpha          0.9238154          0.9367321          0.9294762
## omega          0.6783573          0.9137507          0.9129554
## omega2         0.6783573          0.9137507          0.9129554
## omega3         0.7118750          0.9250568          0.9475383
## avevar         0.7049489          0.7278750          0.7247391
```

```
##          lhs op    rhs      mi    epc sepc.lv sepc.all sepc.nox
## 310          BAPS13 ~~ BAPS14 155.603 0.232 0.232 0.816 0.816
## 160 normalizing_beliefs =~ BAPS1 132.971 0.359 0.359 0.359 0.359
## 322          BAPS16 ~~ BAPS17 105.390 0.186 0.186 1.313 1.313
## 141 survival_strategy =~ BAPS12 62.149 -0.197 -0.197 -0.197 -0.197
## 171 normalizing_beliefs =~ BAPS12 53.847 -0.160 -0.160 -0.160 -0.160
## 220          BAPS4  ~~ BAPS5  53.497 0.185 0.185 0.872 0.872
## 186          BAPS1  ~~ BAPS16 53.061 0.248 0.248 0.902 0.902
## 155 negative_beliefs =~ BAPS14 51.477 0.147 0.147 0.147 0.147
## 290          BAPS10 ~~ BAPS12 44.995 0.132 0.132 1.015 1.015
## 317          BAPS14 ~~ BAPS17 38.219 -0.141 -0.141 -0.792 -0.792
```

Model 2 - without BAPS1

```
## chisq.scaled  df.scaled  cfi.scaled  tli.scaled  rmsea.scaled      srmr
##    903.224    116.000      0.974      0.970      0.081      0.070
```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```
##          survival_strategy  negative_beliefs  normalizing_beliefs
## alpha          0.9295837          0.9367321          0.9294762
## omega          0.8858424          0.9137498          0.9130740
## omega2         0.8858424          0.9137498          0.9130740
## omega3         0.9048820          0.9250671          0.9477461
## avevar         0.7414225          0.7279091          0.7247438
```

```
##          lhs op    rhs      mi    epc sepc.lv sepc.all sepc.nox
## 284          BAPS13 ~~ BAPS14 148.545 0.229 0.229 0.813 0.813
## 296          BAPS16 ~~ BAPS17 114.235 0.197 0.197 1.387 1.387
## 134 survival_strategy =~ BAPS12 58.813 -0.206 -0.206 -0.206 -0.206
```



```
## 147     negative_beliefs =~ BAPS14  52.865  0.148   0.148   0.148   0.148
## 162 normalizing_beliefs =~ BAPS12  50.260 -0.157  -0.157  -0.157  -0.157
## 264             BAPS10 ~~ BAPS12  44.350  0.131   0.131   1.008   1.008
## 291             BAPS14 ~~ BAPS17  41.524 -0.148  -0.148  -0.847  -0.847
## 290             BAPS14 ~~ BAPS16  37.687 -0.149  -0.149  -0.676  -0.676
## 149     negative_beliefs =~ BAPS16  32.384 -0.128  -0.128  -0.128  -0.128
## 287             BAPS13 ~~ BAPS17  30.854 -0.132  -0.132  -0.731  -0.731
```

Comparison between model 1 and 2 of BAPS

```
## [1] "chisq.diff: " "200.528"      "p value: "      "0"
```

MODEL 3 with covariance added between BAPS13 i BAPS14

```
## chisq.scaled    df.scaled    cfi.scaled    tli.scaled rmsea.scaled      srmr
##      742.108      115.000      0.979      0.976      0.072      0.069
```

```
## For constructs with categorical indicators, the alpha and the average variance extracted are calculated
```

```
##      survival_strategy negative_beliefs normalizing_beliefs
## alpha      0.9295837      0.9367321      0.9294762
## omega      0.8858410      0.9137363      0.8800942
## omega2     0.8858410      0.9137363      0.8800942
## omega3     0.9048735      0.9250082      0.8930653
## avevar     0.7414095      0.7278819      0.6919380
```

Comparison between model 2 and 3 of BAPS

```
## [1] "chisq.diff: " "74.144"      "p value: "      "0"
```

Summary of model 3 BAPS

```
## lavaan 0.6-8 ended normally after 28 iterations
```

```
##
```

```
## Estimator DWLS
```

```
## Optimization method NLMINB
```

```
## Number of model parameters 72
```

```
##
```

```
## Number of observations 1047
```

```
##
```

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

```
## Standard Robust
```

```
## Test Statistic 604.967 742.108
```

```
## Degrees of freedom 115 115
```

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

```
## Scaling correction factor 0.867
```

```
## Shift parameter 44.069
```

```
## simple second-order correction
```

```
##
```

```
## Parameter Estimates:
```

```
##
```

```

## Standard errors
## Information
## Information saturated (h1) model
##
## Latent Variables:
##
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## survival_strategy =~
## BAPS2 0.855 0.019 45.756 0.000 0.855 0.855
## BAPS3 0.807 0.020 40.554 0.000 0.807 0.807
## BAPS4 0.898 0.013 69.106 0.000 0.898 0.898
## BAPS5 0.896 0.013 68.219 0.000 0.896 0.896
## BAPS6 0.846 0.019 43.521 0.000 0.846 0.846
## negative_beliefs =~
## BAPS7 0.797 0.019 42.667 0.000 0.797 0.797
## BAPS8 0.782 0.017 46.898 0.000 0.782 0.782
## BAPS9 0.764 0.018 43.346 0.000 0.764 0.764
## BAPS10 0.926 0.008 113.813 0.000 0.926 0.926
## BAPS11 0.892 0.010 92.076 0.000 0.892 0.892
## BAPS12 0.940 0.007 125.622 0.000 0.940 0.940
## normalizing_beliefs =~
## BAPS13 0.758 0.017 44.572 0.000 0.758 0.758
## BAPS14 0.777 0.016 47.806 0.000 0.777 0.777
## BAPS15 0.754 0.018 41.798 0.000 0.754 0.754
## BAPS16 0.910 0.008 109.522 0.000 0.910 0.910
## BAPS17 0.953 0.006 155.467 0.000 0.953 0.953
## BAPS18 0.818 0.014 60.507 0.000 0.818 0.818
##
## Covariances:
##
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .BAPS13 ~~
## .BAPS14 0.220 0.019 11.637 0.000 0.220 0.535
## survival_strategy ~~
## negative_beliefs 0.414 0.033 12.470 0.000 0.414 0.414
## normalzng_blfs 0.493 0.031 15.656 0.000 0.493 0.493
## negative_beliefs ~~
## normalzng_blfs 0.385 0.030 12.725 0.000 0.385 0.385
##
## Intercepts:
##
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .BAPS2 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS3 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS4 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS5 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS6 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS7 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS8 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS9 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS10 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS11 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS12 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS13 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS14 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS15 0.000 0.000 0.000 0.000 0.000 0.000
## .BAPS16 0.000 0.000 0.000 0.000 0.000 0.000

```

##	.BAPS17	0.000			0.000	0.000
##	.BAPS18	0.000			0.000	0.000
##	survivl_strtgy	0.000			0.000	0.000
##	negative_belfs	0.000			0.000	0.000
##	normalzng_blfs	0.000			0.000	0.000
##						
##	Thresholds:					
##		Estimate	Std.Err	z-value	P(> z)	Std.lv Std.all
##	BAPS2 t1	0.826	0.044	18.790	0.000	0.826 0.826
##	BAPS2 t2	1.621	0.064	25.204	0.000	1.621 1.621
##	BAPS2 t3	2.344	0.117	19.953	0.000	2.344 2.344
##	BAPS3 t1	0.566	0.041	13.781	0.000	0.566 0.566
##	BAPS3 t2	1.561	0.062	25.225	0.000	1.561 1.561
##	BAPS3 t3	2.383	0.122	19.488	0.000	2.383 2.383
##	BAPS4 t1	0.511	0.041	12.571	0.000	0.511 0.511
##	BAPS4 t2	1.308	0.054	24.415	0.000	1.308 1.308
##	BAPS4 t3	1.931	0.081	23.937	0.000	1.931 1.931
##	BAPS5 t1	0.143	0.039	3.675	0.000	0.143 0.143
##	BAPS5 t2	1.008	0.047	21.530	0.000	1.008 1.008
##	BAPS5 t3	1.603	0.064	25.218	0.000	1.603 1.603
##	BAPS6 t1	0.895	0.045	19.914	0.000	0.895 0.895
##	BAPS6 t2	1.728	0.069	24.978	0.000	1.728 1.728
##	BAPS6 t3	2.474	0.135	18.371	0.000	2.474 2.474
##	BAPS7 t1	0.586	0.041	14.202	0.000	0.586 0.586
##	BAPS7 t2	1.164	0.050	23.288	0.000	1.164 1.164
##	BAPS7 t3	1.761	0.071	24.862	0.000	1.761 1.761
##	BAPS8 t1	-0.580	0.041	-14.082	0.000	-0.580 -0.580
##	BAPS8 t2	0.165	0.039	4.230	0.000	0.165 0.165
##	BAPS8 t3	0.764	0.043	17.697	0.000	0.764 0.764
##	BAPS9 t1	0.028	0.039	0.710	0.477	0.028 0.028
##	BAPS9 t2	0.592	0.041	14.322	0.000	0.592 0.592
##	BAPS9 t3	1.169	0.050	23.333	0.000	1.169 1.169
##	BAPS10 t1	-0.155	0.039	-3.983	0.000	-0.155 -0.155
##	BAPS10 t2	0.506	0.041	12.450	0.000	0.506 0.506
##	BAPS10 t3	1.033	0.047	21.839	0.000	1.033 1.033
##	BAPS11 t1	-0.128	0.039	-3.304	0.001	-0.128 -0.128
##	BAPS11 t2	0.449	0.040	11.172	0.000	0.449 0.449
##	BAPS11 t3	0.928	0.045	20.409	0.000	0.928 0.928
##	BAPS12 t1	0.061	0.039	1.575	0.115	0.061 0.061
##	BAPS12 t2	0.612	0.042	14.742	0.000	0.612 0.612
##	BAPS12 t3	1.091	0.048	22.538	0.000	1.091 1.091
##	BAPS13 t1	-0.809	0.044	-18.504	0.000	-0.809 -0.809
##	BAPS13 t2	0.153	0.039	3.922	0.000	0.153 0.153
##	BAPS13 t3	0.885	0.045	19.748	0.000	0.885 0.885
##	BAPS14 t1	-0.698	0.042	-16.467	0.000	-0.698 -0.698
##	BAPS14 t2	0.348	0.040	8.786	0.000	0.348 0.348
##	BAPS14 t3	1.169	0.050	23.333	0.000	1.169 1.169
##	BAPS15 t1	0.085	0.039	2.193	0.028	0.085 0.085
##	BAPS15 t2	1.079	0.048	22.392	0.000	1.079 1.079
##	BAPS15 t3	1.833	0.075	24.535	0.000	1.833 1.833
##	BAPS16 t1	-0.250	0.039	-6.388	0.000	-0.250 -0.250
##	BAPS16 t2	0.867	0.045	19.468	0.000	0.867 0.867
##	BAPS16 t3	1.630	0.065	25.194	0.000	1.630 1.630
##	BAPS17 t1	-0.303	0.039	-7.680	0.000	-0.303 -0.303

```

##      BAPS17|t2      0.823    0.044   18.733    0.000    0.823    0.823
##      BAPS17|t3      1.450    0.058   25.052    0.000    1.450    1.450
##      BAPS18|t1     -0.138    0.039   -3.551    0.000   -0.138   -0.138
##      BAPS18|t2      0.966    0.046   20.950    0.000    0.966    0.966
##      BAPS18|t3      1.749    0.070   24.904    0.000    1.749    1.749
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .BAPS2      0.269      0.269      0.269
##      .BAPS3      0.349      0.349      0.349
##      .BAPS4      0.193      0.193      0.193
##      .BAPS5      0.197      0.197      0.197
##      .BAPS6      0.285      0.285      0.285
##      .BAPS7      0.364      0.364      0.364
##      .BAPS8      0.388      0.388      0.388
##      .BAPS9      0.416      0.416      0.416
##      .BAPS10     0.143      0.143      0.143
##      .BAPS11     0.204      0.204      0.204
##      .BAPS12     0.117      0.117      0.117
##      .BAPS13     0.426      0.426      0.426
##      .BAPS14     0.396      0.396      0.396
##      .BAPS15     0.431      0.431      0.431
##      .BAPS16     0.172      0.172      0.172
##      .BAPS17     0.091      0.091      0.091
##      .BAPS18     0.331      0.331      0.331
##      survivl_strtgy 1.000      1.000      1.000
##      negative_belfs 1.000      1.000      1.000
##      normalzng_blfs 1.000      1.000      1.000
##
## Scales y*:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      BAPS2      1.000      1.000      1.000
##      BAPS3      1.000      1.000      1.000
##      BAPS4      1.000      1.000      1.000
##      BAPS5      1.000      1.000      1.000
##      BAPS6      1.000      1.000      1.000
##      BAPS7      1.000      1.000      1.000
##      BAPS8      1.000      1.000      1.000
##      BAPS9      1.000      1.000      1.000
##      BAPS10     1.000      1.000      1.000
##      BAPS11     1.000      1.000      1.000
##      BAPS12     1.000      1.000      1.000
##      BAPS13     1.000      1.000      1.000
##      BAPS14     1.000      1.000      1.000
##      BAPS15     1.000      1.000      1.000
##      BAPS16     1.000      1.000      1.000
##      BAPS17     1.000      1.000      1.000
##      BAPS18     1.000      1.000      1.000

```

summary of fit measures for three models of BAPS

```

##      chisq.scaled      df.scaled chisq.scaling.factor
##      1130.643      132.000      0.921

```

```
##          cfi.scaled          tli.scaled          rmsea.scaled
##          0.968              0.962              0.085
##          srmr
##          0.082

##          chisq.scaled          df.scaled chisq.scaling.factor
##          903.224              116.000              0.876
##          cfi.scaled          tli.scaled          rmsea.scaled
##          0.974              0.970              0.081
##          srmr
##          0.070

##          chisq.scaled          df.scaled chisq.scaling.factor
##          742.108              115.000              0.867
##          cfi.scaled          tli.scaled          rmsea.scaled
##          0.979              0.976              0.072
##          srmr
##          0.069
```

IVI analysis

```
## Warning in lav_model_vcov(lavmodel = lavmodel, lavsamplestats = lavsamplestats, : lavaan WARNING:
##   The variance-covariance matrix of the estimated parameters (vcov)
##   does not appear to be positive definite! The smallest eigenvalue
##   (= -1.661197e-17) is smaller than zero. This may be a symptom that
##   the model is not identified.

## lavaan 0.6-8 ended normally after 23 iterations
##
##   Estimator                      DWLS
##   Optimization method          NLMINB
##   Number of model parameters    106
##
##   Number of observations        433
##
## Model Test User Model:
##
##           Standard      Robust
##   Test Statistic      1912.552  1165.127
##   Degrees of freedom      296      296
##   P-value (Chi-square)    0.000      0.000
##   Scaling correction factor      1.937
##   Shift parameter      177.912
##   simple second-order correction
##
## Model Test Baseline Model:
##
##           Test statistic      17803.863  6278.569
##           Degrees of freedom      325      325
##           P-value      0.000      0.000
##           Scaling correction factor      2.936
##
## User Model versus Baseline Model:
```

```

##
## Comparative Fit Index (CFI)                0.908      0.854
## Tucker-Lewis Index (TLI)                  0.898      0.840
##
## Robust Comparative Fit Index (CFI)          NA
## Robust Tucker-Lewis Index (TLI)            NA
##
## Root Mean Square Error of Approximation:
##
## RMSEA                0.112      0.082
## 90 Percent confidence interval - lower      0.108      0.077
## 90 Percent confidence interval - upper      0.117      0.087
## P-value RMSEA <= 0.05      0.000      0.000
##
## Robust RMSEA          NA
## 90 Percent confidence interval - lower      NA
## 90 Percent confidence interval - upper      NA
##
## Standardized Root Mean Square Residual:
##
## SRMR                0.196      0.196
##
## Parameter Estimates:
##
## Standard errors      Robust.sem
## Information          Expected
## Information saturated (h1) model      Unstructured
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## metaphysical =~
## IVI_1          0.695  0.051  13.578  0.000  0.695  0.695
## IVI_3          0.700  0.042  16.600  0.000  0.700  0.700
## IVI_4          0.674  0.042  16.065  0.000  0.674  0.674
## IVI_5          0.624  0.045  13.761  0.000  0.624  0.624
## IVI_6          0.684  0.044  15.420  0.000  0.684  0.684
## IVI_8          0.880  0.030  29.431  0.000  0.880  0.880
## IVI_13         0.679  0.035  19.496  0.000  0.679  0.679
## IVI_14         0.841  0.040  21.206  0.000  0.841  0.841
## IVI_15         0.733  0.041  17.831  0.000  0.733  0.733
## IVI_16         0.499  0.063   7.956  0.000  0.499  0.499
## IVI_17         0.690  0.044  15.697  0.000  0.690  0.690
## IVI_25         0.740  0.046  16.197  0.000  0.740  0.740
## IVI_26         0.824  0.053  15.461  0.000  0.824  0.824
## positive =~
## IVI_2          0.693  0.040  17.126  0.000  0.693  0.693
## IVI_7          0.823  0.031  26.745  0.000  0.823  0.823
## IVI_9          0.961  0.024  39.307  0.000  0.961  0.961
## IVI_18         0.822  0.030  27.420  0.000  0.822  0.822
## IVI_19         0.781  0.032  24.458  0.000  0.781  0.781
## IVI_20         0.655  0.041  15.873  0.000  0.655  0.655
## IVI_21         0.656  0.078   8.405  0.000  0.656  0.656
## IVI_24         0.816  0.033  24.802  0.000  0.816  0.816
## loss_of_control =~

```

```

##      IVI_10      0.877    0.020   44.815    0.000    0.877    0.877
##      IVI_11      0.861    0.025   34.199    0.000    0.861    0.861
##      IVI_12      0.889    0.023   39.022    0.000    0.889    0.889
##      IVI_22      0.817    0.035   23.530    0.000    0.817    0.817
##      IVI_23      0.843    0.028   29.974    0.000    0.843    0.843
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      metaphysical ~~
##      positive      0.757    0.029   25.893    0.000    0.757    0.757
##      loss_of_contrl 0.599    0.034   17.743    0.000    0.599    0.599
##      positive ~~
##      loss_of_contrl 0.030    0.058    0.522    0.602    0.030    0.030
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .IVI_1      0.000
##      .IVI_3      0.000
##      .IVI_4      0.000
##      .IVI_5      0.000
##      .IVI_6      0.000
##      .IVI_8      0.000
##      .IVI_13     0.000
##      .IVI_14     0.000
##      .IVI_15     0.000
##      .IVI_16     0.000
##      .IVI_17     0.000
##      .IVI_25     0.000
##      .IVI_26     0.000
##      .IVI_2      0.000
##      .IVI_7      0.000
##      .IVI_9      0.000
##      .IVI_18     0.000
##      .IVI_19     0.000
##      .IVI_20     0.000
##      .IVI_21     0.000
##      .IVI_24     0.000
##      .IVI_10     0.000
##      .IVI_11     0.000
##      .IVI_12     0.000
##      .IVI_22     0.000
##      .IVI_23     0.000
##      metaphysical 0.000
##      positive      0.000
##      loss_of_contrl 0.000
##
## Thresholds:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      IVI_1|t1     1.683    0.104   16.132    0.000    1.683    1.683
##      IVI_1|t2     1.994    0.132   15.084    0.000    1.994    1.994
##      IVI_1|t3     2.832    0.320    8.860    0.000    2.832    2.832
##      IVI_3|t1     0.906    0.070   12.904    0.000    0.906    0.906
##      IVI_3|t2     1.464    0.091   16.124    0.000    1.464    1.464
##      IVI_3|t3     2.086    0.143   14.575    0.000    2.086    2.086

```

##	IVI_4 t1	1.326	0.084	15.761	0.000	1.326	1.326
##	IVI_4 t2	1.916	0.124	15.450	0.000	1.916	1.916
##	IVI_4 t3	2.603	0.242	10.751	0.000	2.603	2.603
##	IVI_5 t1	1.108	0.076	14.601	0.000	1.108	1.108
##	IVI_5 t2	1.660	0.103	16.163	0.000	1.660	1.660
##	IVI_5 t3	2.201	0.159	13.845	0.000	2.201	2.201
##	IVI_6 t1	1.198	0.079	15.166	0.000	1.198	1.198
##	IVI_6 t2	1.880	0.121	15.594	0.000	1.880	1.880
##	IVI_6 t3	2.603	0.242	10.751	0.000	2.603	2.603
##	IVI_8 t1	1.119	0.076	14.675	0.000	1.119	1.119
##	IVI_8 t2	1.683	0.104	16.132	0.000	1.683	1.683
##	IVI_8 t3	2.272	0.170	13.352	0.000	2.272	2.272
##	IVI_13 t1	0.578	0.064	9.011	0.000	0.578	0.578
##	IVI_13 t2	1.198	0.079	15.166	0.000	1.198	1.198
##	IVI_13 t3	1.733	0.108	16.042	0.000	1.733	1.733
##	IVI_14 t1	1.707	0.106	16.092	0.000	1.707	1.707
##	IVI_14 t2	2.461	0.207	11.908	0.000	2.461	2.461
##	IVI_14 t3	2.603	0.242	10.751	0.000	2.603	2.603
##	IVI_15 t1	1.186	0.079	15.099	0.000	1.186	1.186
##	IVI_15 t2	1.787	0.112	15.907	0.000	1.787	1.787
##	IVI_15 t3	2.201	0.159	13.845	0.000	2.201	2.201
##	IVI_16 t1	1.247	0.081	15.422	0.000	1.247	1.247
##	IVI_16 t2	1.916	0.124	15.450	0.000	1.916	1.916
##	IVI_16 t3	2.356	0.185	12.729	0.000	2.356	2.356
##	IVI_17 t1	1.415	0.088	16.028	0.000	1.415	1.415
##	IVI_17 t2	2.201	0.159	13.845	0.000	2.201	2.201
##	IVI_17 t3	2.832	0.320	8.860	0.000	2.832	2.832
##	IVI_25 t1	1.299	0.083	15.655	0.000	1.299	1.299
##	IVI_25 t2	1.916	0.124	15.450	0.000	1.916	1.916
##	IVI_25 t3	2.461	0.207	11.908	0.000	2.461	2.461
##	IVI_26 t1	1.817	0.115	15.819	0.000	1.817	1.817
##	IVI_26 t2	2.461	0.207	11.908	0.000	2.461	2.461
##	IVI_26 t3	2.603	0.242	10.751	0.000	2.603	2.603
##	IVI_2 t1	0.906	0.070	12.904	0.000	0.906	0.906
##	IVI_2 t2	1.637	0.101	16.185	0.000	1.637	1.637
##	IVI_2 t3	2.140	0.150	14.245	0.000	2.140	2.140
##	IVI_7 t1	0.830	0.068	12.120	0.000	0.830	0.830
##	IVI_7 t2	1.431	0.089	16.063	0.000	1.431	1.431
##	IVI_7 t3	2.086	0.143	14.575	0.000	2.086	2.086
##	IVI_9 t1	1.119	0.076	14.675	0.000	1.119	1.119
##	IVI_9 t2	1.817	0.115	15.819	0.000	1.817	1.817
##	IVI_9 t3	2.356	0.185	12.729	0.000	2.356	2.356
##	IVI_18 t1	0.978	0.072	13.577	0.000	0.978	0.978
##	IVI_18 t2	1.574	0.097	16.212	0.000	1.574	1.574
##	IVI_18 t3	2.201	0.159	13.845	0.000	2.201	2.201
##	IVI_19 t1	0.822	0.068	12.031	0.000	0.822	0.822
##	IVI_19 t2	1.615	0.100	16.201	0.000	1.615	1.615
##	IVI_19 t3	2.272	0.170	13.352	0.000	2.272	2.272
##	IVI_20 t1	0.774	0.067	11.495	0.000	0.774	0.774
##	IVI_20 t2	1.574	0.097	16.212	0.000	1.574	1.574
##	IVI_20 t3	2.201	0.159	13.845	0.000	2.201	2.201
##	IVI_21 t1	1.707	0.106	16.092	0.000	1.707	1.707
##	IVI_21 t2	2.461	0.207	11.908	0.000	2.461	2.461
##	IVI_24 t1	1.152	0.077	14.891	0.000	1.152	1.152

##	IVI_24 t2	1.994	0.132	15.084	0.000	1.994	1.994
##	IVI_24 t3	2.832	0.320	8.860	0.000	2.832	2.832
##	IVI_10 t1	-0.309	0.061	-5.032	0.000	-0.309	-0.309
##	IVI_10 t2	0.433	0.062	6.937	0.000	0.433	0.433
##	IVI_10 t3	1.175	0.078	15.031	0.000	1.175	1.175
##	IVI_11 t1	-0.101	0.060	-1.680	0.093	-0.101	-0.101
##	IVI_11 t2	0.598	0.064	9.291	0.000	0.598	0.598
##	IVI_11 t3	1.326	0.084	15.761	0.000	1.326	1.326
##	IVI_12 t1	0.497	0.063	7.883	0.000	0.497	0.497
##	IVI_12 t2	1.076	0.075	14.374	0.000	1.076	1.076
##	IVI_12 t3	1.637	0.101	16.185	0.000	1.637	1.637
##	IVI_22 t1	0.960	0.072	13.411	0.000	0.960	0.960
##	IVI_22 t2	1.660	0.103	16.163	0.000	1.660	1.660
##	IVI_22 t3	2.356	0.185	12.729	0.000	2.356	2.356
##	IVI_23 t1	0.736	0.067	11.043	0.000	0.736	0.736
##	IVI_23 t2	1.535	0.095	16.201	0.000	1.535	1.535
##	IVI_23 t3	1.994	0.132	15.084	0.000	1.994	1.994

##

Variances:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.IVI_1	0.517				0.517	0.517
##	.IVI_3	0.510				0.510	0.510
##	.IVI_4	0.545				0.545	0.545
##	.IVI_5	0.610				0.610	0.610
##	.IVI_6	0.533				0.533	0.533
##	.IVI_8	0.225				0.225	0.225
##	.IVI_13	0.540				0.540	0.540
##	.IVI_14	0.293				0.293	0.293
##	.IVI_15	0.463				0.463	0.463
##	.IVI_16	0.751				0.751	0.751
##	.IVI_17	0.523				0.523	0.523
##	.IVI_25	0.452				0.452	0.452
##	.IVI_26	0.321				0.321	0.321
##	.IVI_2	0.520				0.520	0.520
##	.IVI_7	0.322				0.322	0.322
##	.IVI_9	0.077				0.077	0.077
##	.IVI_18	0.324				0.324	0.324
##	.IVI_19	0.389				0.389	0.389
##	.IVI_20	0.571				0.571	0.571
##	.IVI_21	0.570				0.570	0.570
##	.IVI_24	0.334				0.334	0.334
##	.IVI_10	0.231				0.231	0.231
##	.IVI_11	0.258				0.258	0.258
##	.IVI_12	0.210				0.210	0.210
##	.IVI_22	0.332				0.332	0.332
##	.IVI_23	0.290				0.290	0.290
##	metaphysical	1.000				1.000	1.000
##	positive	1.000				1.000	1.000
##	loss_of_contr1	1.000				1.000	1.000

##

Scales y*:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	IVI_1	1.000				1.000	1.000
##	IVI_3	1.000				1.000	1.000

##	IVI_4	1.000	1.000	1.000
##	IVI_5	1.000	1.000	1.000
##	IVI_6	1.000	1.000	1.000
##	IVI_8	1.000	1.000	1.000
##	IVI_13	1.000	1.000	1.000
##	IVI_14	1.000	1.000	1.000
##	IVI_15	1.000	1.000	1.000
##	IVI_16	1.000	1.000	1.000
##	IVI_17	1.000	1.000	1.000
##	IVI_25	1.000	1.000	1.000
##	IVI_26	1.000	1.000	1.000
##	IVI_2	1.000	1.000	1.000
##	IVI_7	1.000	1.000	1.000
##	IVI_9	1.000	1.000	1.000
##	IVI_18	1.000	1.000	1.000
##	IVI_19	1.000	1.000	1.000
##	IVI_20	1.000	1.000	1.000
##	IVI_21	1.000	1.000	1.000
##	IVI_24	1.000	1.000	1.000
##	IVI_10	1.000	1.000	1.000
##	IVI_11	1.000	1.000	1.000
##	IVI_12	1.000	1.000	1.000
##	IVI_22	1.000	1.000	1.000
##	IVI_23	1.000	1.000	1.000

the model was not identified, so the the factor solution was not optimal. Thus we did run an iclust algorithm to find alternative factor model

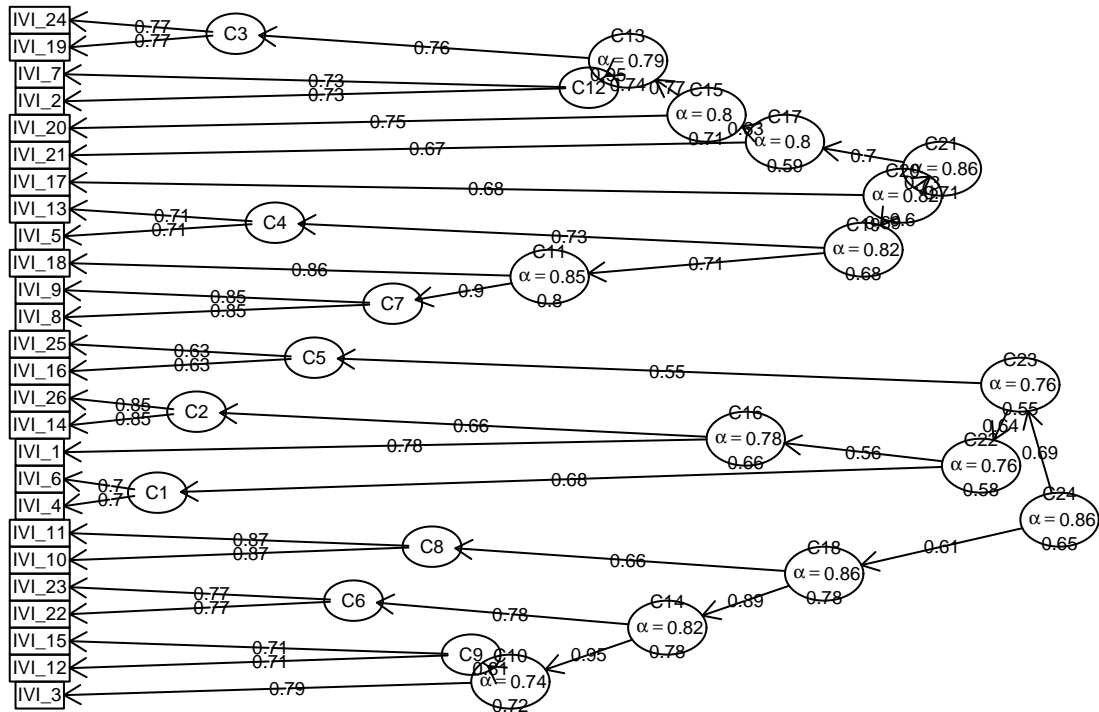
the iclust plot is kind of hard to read so I saved it in separate .pdf file “iclust_ivi.pdf”

```
##
## Attaching package: 'psych'

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

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

ICLUST



```
## ICLUST (Item Cluster Analysis)
## Call: iclust(r.mat = (POLYCHORIC_R(d, method = "Fox", verbose = F)))
##
## Purified Alpha:
## C24 C21
## 0.86 0.86
##
## G6* reliability:
## C24 C21
## 1 1
##
## Original Beta:
## C24 C21
## 0.65 0.71
##
## Cluster size:
## C24 C21
## 14 12
##
## Item by Cluster Structure matrix:
##      0 P C24 C21
## IVI_1 C24 C24 0.48 0.26
## IVI_2 C21 C21 0.11 0.56
## IVI_3 C24 C24 0.57 0.20
## IVI_4 C24 C24 0.43 0.21
## IVI_5 C21 C21 0.15 0.52
```

```

## IVI_6  C24 C24  0.56  0.12
## IVI_7  C21 C21  0.16  0.68
## IVI_8  C21 C21  0.22  0.66
## IVI_9  C21 C21  0.19  0.70
## IVI_10 C24 C24  0.60 -0.04
## IVI_11 C24 C24  0.57 -0.06
## IVI_12 C24 C24  0.71  0.13
## IVI_13 C21 C21  0.22  0.63
## IVI_14 C24 C24  0.58  0.20
## IVI_15 C24 C24  0.64  0.07
## IVI_16 C24 C24  0.38  0.17
## IVI_17 C21 C21  0.32  0.47
## IVI_18 C21 C21  0.17  0.67
## IVI_19 C21 C21  0.00  0.68
## IVI_20 C21 C21  0.17  0.52
## IVI_21 C21 C21  0.06  0.34
## IVI_22 C24 C24  0.63  0.04
## IVI_23 C24 C24  0.64  0.19
## IVI_24 C21 C21  0.05  0.68
## IVI_25 C24 C24  0.54  0.33
## IVI_26 C24 C24  0.55  0.18
##
## With eigenvalues of:
## C24 C21
## 4.6 4.5
##
## Purified scale intercorrelations
## reliabilities on diagonal
## correlations corrected for attenuation above diagonal:
##      C24  C21
## C24 0.86 0.27
## C21 0.23 0.86
##
## Cluster fit = 0.72  Pattern fit = 0.93  RMSR = 0.09

```

PQ16 analysis, starting with simple 1-factor model, only with data from 2nd study

```

##      chisq.scaled      df.scaled chisq.scaling.factor
##      2829.371         104.000         0.644
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.943           0.934           0.062
##      srmr
##      0.052

```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```

##      factor
## alpha 0.9179273
## omega 0.7640425
## omega2 0.7640425
## omega3 0.7699459
## avevar 0.4234952

```

```
##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 196 pq7 ~~ pq14 293.171  0.202   0.202   0.327   0.327
## 120 pq1 ~~ pq7  154.851  0.161   0.161   0.239   0.239
## 205 pq8 ~~ pq15 105.615  0.151   0.151   0.305   0.305
## 126 pq1 ~~ pq13  80.242 -0.169  -0.169  -0.326  -0.326
## 207 pq9 ~~ pq10  69.735  0.108   0.108   0.184   0.184
## 190 pq7 ~~ pq8   64.749 -0.150  -0.150  -0.257  -0.257
## 121 pq1 ~~ pq8   53.966 -0.140  -0.140  -0.250  -0.250
## 204 pq8 ~~ pq14  53.909 -0.127  -0.127  -0.248  -0.248
## 222 pq11 ~~ pq14  53.625  0.090   0.090   0.172   0.172
## 115 pq1 ~~ pq2   52.782  0.097   0.097   0.141   0.141
```

Next, we added covariance between pq7 and pq14

```
##      chisq.scaled      df.scaled chisq.scaling.factor
##      2389.095      103.000      0.640
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.952      0.944      0.057
##      srmr
##      0.049
```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```
##      factor
## alpha 0.9179273
## omega 0.7611770
## omega2 0.7611770
## omega3 0.7656690
## avevar 0.4215713
```

```
##      lhs op  rhs      mi      epc sepc.lv sepc.all sepc.nox
## 121 pq1 ~~ pq7 204.665  0.186   0.186   0.268   0.268
## 205 pq8 ~~ pq15 99.978  0.147   0.147   0.299   0.299
## 127 pq1 ~~ pq13  83.139 -0.172  -0.172  -0.334  -0.334
## 222 pq11 ~~ pq14  81.380  0.111   0.111   0.206   0.206
## 207 pq9 ~~ pq10  65.065  0.105   0.105   0.179   0.179
## 194 pq7 ~~ pq11  62.186  0.107   0.107   0.175   0.175
## 128 pq1 ~~ pq14  59.537  0.097   0.097   0.159   0.159
## 122 pq1 ~~ pq8   56.231 -0.143  -0.143  -0.257  -0.257
## 116 pq1 ~~ pq2   50.428  0.095   0.095   0.139   0.139
## 152 pq3 ~~ pq11  46.623 -0.107  -0.107  -0.177  -0.177
```

Comparison between model 1 and 2 of PQ16

```
## [1] "chisq.diff: " "274.242"      "p value: "      "0"
```

and in the next model, we added covariance between pq1 and pq7

```
##      chisq.scaled      df.scaled chisq.scaling.factor
##      2080.415      102.000      0.636
##      cfi.scaled      tli.scaled      rmsea.scaled
```

```
##          0.958          0.951          0.054
##          srmr
##          0.045
```

For constructs with categorical indicators, the alpha and the average variance extracted are calculated

```
##          factor
## alpha  0.9179273
## omega  0.7394947
## omega2 0.7394947
## omega3 0.7455761
## avevar 0.4200442
```

Comparison between model 1 and 2 of PQ16

```
## [1] "chisq.diff: " "203.971"      "p value: "      "0"
```

Summary of model 3 PQ16

```
## lavaan 0.6-8 ended normally after 20 iterations
##
## Estimator                      DWLS
## Optimization method           NLMINB
## Number of model parameters     66
##
## Number of observations         6772
##
## Model Test User Model:
##                               Standard    Robust
## Test Statistic               1318.434    2080.415
## Degrees of freedom           102         102
## P-value (Chi-square)         0.000         0.000
## Scaling correction factor     0.636
## Shift parameter              7.445
##      simple second-order correction
##
## Parameter Estimates:
##
## Standard errors               Robust.sem
## Information                   Expected
## Information saturated (h1) model Unstructured
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## factor =~
##   pq1         0.576   0.012  48.088   0.000   0.576   0.576
##   pq2         0.527   0.011  46.026   0.000   0.527   0.527
##   pq3         0.537   0.011  46.982   0.000   0.537   0.537
##   pq4         0.639   0.010  65.594   0.000   0.639   0.639
##   pq5         0.672   0.009  73.918   0.000   0.672   0.672
##   pq6         0.702   0.013  53.734   0.000   0.702   0.702
##   pq7         0.480   0.012  40.709   0.000   0.480   0.480
```

```

##      pq8          0.722    0.011   65.120    0.000    0.722    0.722
##      pq9          0.687    0.009   79.137    0.000    0.687    0.687
##      pq10         0.599    0.010   60.401    0.000    0.599    0.599
##      pq11         0.703    0.009   81.329    0.000    0.703    0.703
##      pq12         0.705    0.009   76.241    0.000    0.705    0.705
##      pq13         0.769    0.010   77.273    0.000    0.769    0.769
##      pq14         0.651    0.009   69.788    0.000    0.651    0.651
##      pq15         0.709    0.009   78.872    0.000    0.709    0.709
##      pq16         0.616    0.011   55.173    0.000    0.616    0.616
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .pq7 ~~
##      .pq14          0.218    0.011   20.316    0.000    0.218    0.327
##      .pq1 ~~
##      .pq7          0.185    0.012   15.985    0.000    0.185    0.258
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .pq1          0.000
##      .pq2          0.000
##      .pq3          0.000
##      .pq4          0.000
##      .pq5          0.000
##      .pq6          0.000
##      .pq7          0.000
##      .pq8          0.000
##      .pq9          0.000
##      .pq10         0.000
##      .pq11         0.000
##      .pq12         0.000
##      .pq13         0.000
##      .pq14         0.000
##      .pq15         0.000
##      .pq16         0.000
##      factor        0.000
##
## Thresholds:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      pq1|t1       -1.048    0.019  -56.044    0.000   -1.048   -1.048
##      pq1|t2         0.836    0.017   48.212    0.000    0.836    0.836
##      pq1|t3         2.014    0.034   59.302    0.000    2.014    2.014
##      pq2|t1       -0.884    0.018  -50.226    0.000   -0.884   -0.884
##      pq2|t2         0.808    0.017   47.023    0.000    0.808    0.808
##      pq2|t3         2.074    0.036   57.992    0.000    2.074    2.074
##      pq3|t1         0.130    0.015    8.479    0.000    0.130    0.130
##      pq3|t2         1.298    0.021   61.963    0.000    1.298    1.298
##      pq3|t3         2.142    0.038   56.351    0.000    2.142    2.142
##      pq4|t1         0.082    0.015    5.370    0.000    0.082    0.082
##      pq4|t2         1.376    0.022   63.043    0.000    1.376    1.376
##      pq4|t3         2.273    0.043   52.832    0.000    2.273    2.273
##      pq5|t1         0.063    0.015    4.155    0.000    0.063    0.063
##      pq5|t2         1.453    0.023   63.761    0.000    1.453    1.453
##      pq5|t3         2.342    0.046   50.820    0.000    2.342    2.342

```

##	pq6 t1	1.105	0.019	57.712	0.000	1.105	1.105
##	pq6 t2	1.873	0.030	61.856	0.000	1.873	1.873
##	pq6 t3	2.509	0.055	45.614	0.000	2.509	2.509
##	pq7 t1	-0.216	0.015	-14.036	0.000	-0.216	-0.216
##	pq7 t2	0.846	0.017	48.658	0.000	0.846	0.846
##	pq7 t3	1.410	0.022	63.403	0.000	1.410	1.410
##	pq8 t1	0.913	0.018	51.336	0.000	0.913	0.913
##	pq8 t2	1.866	0.030	61.952	0.000	1.866	1.866
##	pq8 t3	2.460	0.052	47.163	0.000	2.460	2.460
##	pq9 t1	0.028	0.015	1.823	0.068	0.028	0.028
##	pq9 t2	1.060	0.019	56.419	0.000	1.060	1.060
##	pq9 t3	1.763	0.028	63.235	0.000	1.763	1.763
##	pq10 t1	-0.037	0.015	-2.430	0.015	-0.037	-0.037
##	pq10 t2	0.972	0.018	53.530	0.000	0.972	0.972
##	pq10 t3	1.759	0.028	63.269	0.000	1.759	1.759
##	pq11 t1	0.124	0.015	8.091	0.000	0.124	0.124
##	pq11 t2	1.209	0.020	60.276	0.000	1.209	1.209
##	pq11 t3	2.058	0.035	58.347	0.000	2.058	2.058
##	pq12 t1	0.444	0.016	28.114	0.000	0.444	0.444
##	pq12 t2	1.505	0.023	64.046	0.000	1.505	1.505
##	pq12 t3	2.244	0.042	53.634	0.000	2.244	2.244
##	pq13 t1	0.913	0.018	51.336	0.000	0.913	0.913
##	pq13 t2	1.795	0.029	62.888	0.000	1.795	1.795
##	pq13 t3	2.397	0.049	49.147	0.000	2.397	2.397
##	pq14 t1	-0.333	0.016	-21.419	0.000	-0.333	-0.333
##	pq14 t2	1.037	0.019	55.704	0.000	1.037	1.037
##	pq14 t3	1.825	0.029	62.516	0.000	1.825	1.825
##	pq15 t1	0.423	0.016	26.865	0.000	0.423	0.423
##	pq15 t2	1.583	0.025	64.182	0.000	1.583	1.583
##	pq15 t3	2.309	0.045	51.800	0.000	2.309	2.309
##	pq16 t1	0.511	0.016	31.967	0.000	0.511	0.511
##	pq16 t2	1.457	0.023	63.791	0.000	1.457	1.457
##	pq16 t3	2.184	0.040	55.264	0.000	2.184	2.184
##							
##	Variances:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.pq1	0.668				0.668	0.668
##	.pq2	0.722				0.722	0.722
##	.pq3	0.711				0.711	0.711
##	.pq4	0.592				0.592	0.592
##	.pq5	0.548				0.548	0.548
##	.pq6	0.508				0.508	0.508
##	.pq7	0.769				0.769	0.769
##	.pq8	0.479				0.479	0.479
##	.pq9	0.529				0.529	0.529
##	.pq10	0.642				0.642	0.642
##	.pq11	0.506				0.506	0.506
##	.pq12	0.503				0.503	0.503
##	.pq13	0.408				0.408	0.408
##	.pq14	0.576				0.576	0.576
##	.pq15	0.498				0.498	0.498
##	.pq16	0.620				0.620	0.620
##	factor	1.000				1.000	1.000
##							


```
## Scales y*:
##           Estimate Std.Err  z-value  P(>|z|)   Std.lv  Std.all
##    pq1           1.000           1.000   1.000   1.000
##    pq2           1.000           1.000   1.000   1.000
##    pq3           1.000           1.000   1.000   1.000
##    pq4           1.000           1.000   1.000   1.000
##    pq5           1.000           1.000   1.000   1.000
##    pq6           1.000           1.000   1.000   1.000
##    pq7           1.000           1.000   1.000   1.000
##    pq8           1.000           1.000   1.000   1.000
##    pq9           1.000           1.000   1.000   1.000
##    pq10          1.000           1.000   1.000   1.000
##    pq11          1.000           1.000   1.000   1.000
##    pq12          1.000           1.000   1.000   1.000
##    pq13          1.000           1.000   1.000   1.000
##    pq14          1.000           1.000   1.000   1.000
##    pq15          1.000           1.000   1.000   1.000
##    pq16          1.000           1.000   1.000   1.000
```

summary of fit measures for three models of PQ16

```
##      chisq.scaled      df.scaled chisq.scaling.factor
##      2829.371         104.000         0.644
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.943           0.934           0.062
##      srmr
##      0.052
```

```
##      chisq.scaled      df.scaled chisq.scaling.factor
##      2389.095         103.000         0.640
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.952           0.944           0.057
##      srmr
##      0.049
```

```
##      chisq.scaled      df.scaled chisq.scaling.factor
##      2080.415         102.000         0.636
##      cfi.scaled      tli.scaled      rmsea.scaled
##      0.958           0.951           0.054
##      srmr
##      0.045
```