

2019R2 STAT6108 Assignment 4

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```
# import data
teachers <- read.csv('hw4(2020).dat', header = FALSE, sep = '')
```

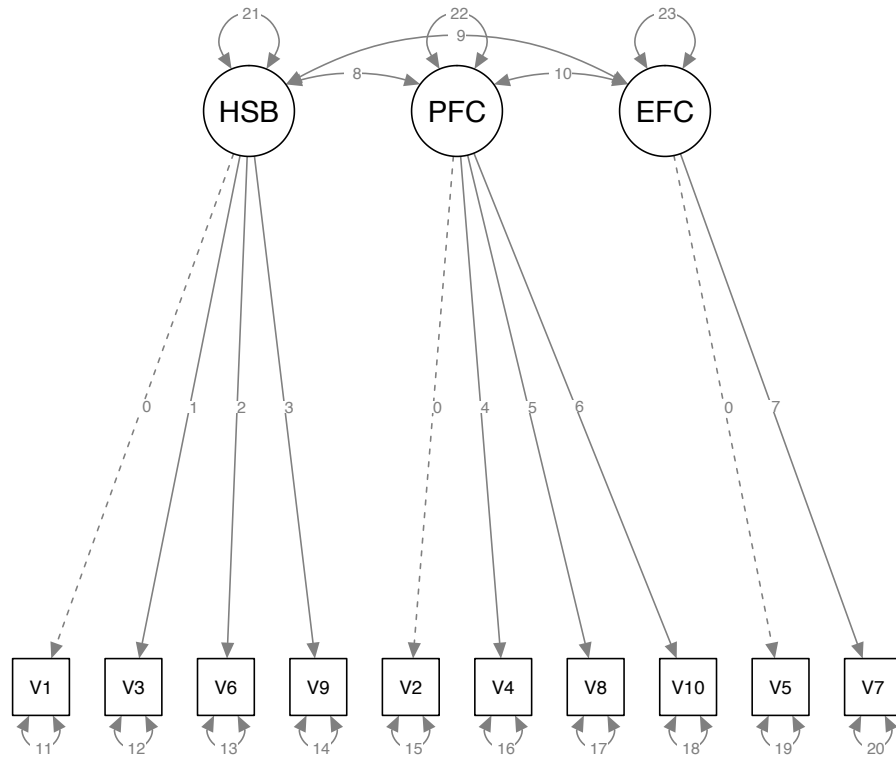
Model Setup

```
model <- 'HelpsSeekingBehavior =~ V1 + V3 + V6 + V9;
ProblemFocusedCoping =~ V2 + V4 + V8 + V10;
EmotionFocusedCoping =~ V5 + V7;
HelpsSeekingBehavior ~~ ProblemFocusedCoping;
HelpsSeekingBehavior ~~ EmotionFocusedCoping;
ProblemFocusedCoping ~~ EmotionFocusedCoping'
```

a)

Path diagram assuming unit loading identification

```
fit <- cfa(model, data = teachers)
semPlot::semPaths(object=fit, intercepts=FALSE, what="path", whatLabels="cons")
```



- Edge labeled 0 indicates the parameter is fixed, non-zero indicates free parameters.
- The factor loading of the first indicator is set to 1.0 for every latent variable to circumvent factor indeterminacy.
- The model satisfies the t-rule because the degrees of freedom is a positive integer.
- Latent factors are correlated, and degree of freedom is greater than zero. Hence the model is over-identified.

Two-indicator rule

- There are three factors
- Factor correlations are free
- At least 2 indicators per factor
- Each indicator loads on one factor
- Errors are uncorrelated

The model is identifiable.

b)

```
n_variable <- 10
n_factor <- 3

p_star <- (n_variable * (n_variable+1))/2
q <- n_variable + (n_variable - n_factor) + (n_factor * (n_factor+1))/2

df <- p_star - q
```

- There are 23.000 free parameters in the the proposed model.
- The degrees of freedom is 32.000.

c)

ULI

```
uli <- lavaan(model, data=teachers, auto.var=TRUE, auto.fix.first=TRUE, std.lv=FALSE)
uliSummary <- summary(uli, fit.measures=TRUE, standardized=TRUE, rsquare=TRUE)
```

UVI

```
uvi <- lavaan(model, data=teachers, auto.var=TRUE, auto.fix.first=FALSE, std.lv=TRUE)
uviSummary <- summary(uvi, fit.measures=TRUE, standardized=TRUE, rsquare=TRUE)
```

Prepare for comparison

```
# row numbers of fixed parameters in parameter table
uliFixedParameters <- c(1,5,9)
uviFixedParameters <- c(24,25,26)

# round numbers to 3 decimal places
uliSummary$PE[, -c(1:3)] <- round(uliSummary$PE[, -c(1:3)], 3)
uviSummary$PE[, -c(1:3)] <- round(uviSummary$PE[, -c(1:3)], 3)
```

Parameter estimations that are common in both methods

```
# compare all paremeters except fixed parameters
combined <- rbind(uliSummary$PE, uviSummary$PE)
interaction <- combined[duplicated(combined), , drop = FALSE]
interaction[, -which(names(interaction) == "std.no")]
```

##	lhs	op	rhs	exo	est	se	z	pvalue	std.lv	std.all
## 50	V1	~~	V1	0	0.211	0.026	8.190	0.000	0.211	0.426
## 51	V3	~~	V3	0	0.528	0.060	8.759	0.000	0.528	0.485
## 52	V6	~~	V6	0	0.404	0.043	9.433	0.000	0.404	0.592
## 53	V9	~~	V9	0	0.172	0.039	4.454	0.000	0.172	0.205
## 54	V2	~~	V2	0	0.391	0.052	7.479	0.000	0.391	0.469
## 55	V4	~~	V4	0	0.114	0.036	3.150	0.002	0.114	0.209
## 56	V8	~~	V8	0	0.321	0.033	9.837	0.000	0.321	0.747
## 57	V10	~~	V10	0	0.508	0.052	9.853	0.000	0.508	0.751
## 58	V5	~~	V5	0	0.225	0.073	3.072	0.002	0.225	0.314
## 59	V7	~~	V7	0	0.534	0.086	6.246	0.000	0.534	0.525
## 63	V1	r2	V1	0	0.574	NA	NA	NA	NA	NA
## 64	V3	r2	V3	0	0.515	NA	NA	NA	NA	NA
## 65	V6	r2	V6	0	0.408	NA	NA	NA	NA	NA
## 66	V9	r2	V9	0	0.795	NA	NA	NA	NA	NA
## 67	V2	r2	V2	0	0.531	NA	NA	NA	NA	NA
## 68	V4	r2	V4	0	0.791	NA	NA	NA	NA	NA
## 69	V8	r2	V8	0	0.253	NA	NA	NA	NA	NA
## 70	V10	r2	V10	0	0.249	NA	NA	NA	NA	NA
## 71	V5	r2	V5	0	0.686	NA	NA	NA	NA	NA
## 72	V7	r2	V7	0	0.475	NA	NA	NA	NA	NA

- Error variance and R^2 are the same across two identification methods
- All error variance are significant

Parameter estimations that are different

```
nRowTable <- dim(uviSummary$PE)[1]
tail <- tail(duplicated(combined), nRowTable)
uliSummary$PE[!tail, -which(colnames(uliSummary$PE) == "std.no")]
```

##	lhs	op	rhs	exo	est	se	z	pvalue	std.lv	std.all
## 1	HelpsSeekingBehavior	=~	V1	0	1.000	0.000	NA	NA	0.534	0.758
## 2	HelpsSeekingBehavior	=~	V3	0	1.402	0.134	10.434	0.000	0.749	0.718
## 3	HelpsSeekingBehavior	=~	V6	0	0.989	0.107	9.221	0.000	0.528	0.639
## 4	HelpsSeekingBehavior	=~	V9	0	1.533	0.126	12.196	0.000	0.818	0.892
## 5	ProblemFocusedCoping	=~	V2	0	1.000	0.000	NA	NA	0.665	0.729
## 6	ProblemFocusedCoping	=~	V4	0	0.989	0.103	9.607	0.000	0.658	0.889
## 7	ProblemFocusedCoping	=~	V8	0	0.496	0.072	6.881	0.000	0.330	0.503
## 8	ProblemFocusedCoping	=~	V10	0	0.617	0.090	6.829	0.000	0.410	0.499
## 9	EmotionFocusedCoping	=~	V5	0	1.000	0.000	NA	NA	0.702	0.828
## 10	EmotionFocusedCoping	=~	V7	0	0.991	0.157	6.318	0.000	0.695	0.689
## 11	HelpsSeekingBehavior	~~	ProblemFocusedCoping	0	0.029	0.028	1.027	0.305	0.081	0.081
## 12	HelpsSeekingBehavior	~~	EmotionFocusedCoping	0	0.144	0.034	4.204	0.000	0.384	0.384
## 13	ProblemFocusedCoping	~~	EmotionFocusedCoping	0	0.210	0.046	4.586	0.000	0.449	0.449
## 24	HelpsSeekingBehavior	~~	HelpsSeekingBehavior	0	0.285	0.046	6.250	0.000	1.000	1.000
## 25	ProblemFocusedCoping	~~	ProblemFocusedCoping	0	0.442	0.079	5.601	0.000	1.000	1.000
## 26	EmotionFocusedCoping	~~	EmotionFocusedCoping	0	0.492	0.095	5.161	0.000	1.000	1.000

```
uviSummary$PE[!tail, -which(names(uviSummary$PE) == "std.no")]
```

##	lhs	op	rhs	exo	est	se	z	pvalue	std.lv	std.all
## 1	HelpsSeekingBehavior	=~	V1	0	0.534	0.043	12.500	0.000	0.534	0.758
## 2	HelpsSeekingBehavior	=~	V3	0	0.749	0.064	11.636	0.000	0.749	0.718
## 3	HelpsSeekingBehavior	=~	V6	0	0.528	0.053	10.026	0.000	0.528	0.639
## 4	HelpsSeekingBehavior	=~	V9	0	0.818	0.052	15.619	0.000	0.818	0.892
## 5	ProblemFocusedCoping	=~	V2	0	0.665	0.059	11.203	0.000	0.665	0.729
## 6	ProblemFocusedCoping	=~	V4	0	0.658	0.047	14.078	0.000	0.658	0.889
## 7	ProblemFocusedCoping	=~	V8	0	0.330	0.045	7.356	0.000	0.330	0.503
## 8	ProblemFocusedCoping	=~	V10	0	0.410	0.056	7.293	0.000	0.410	0.499
## 9	EmotionFocusedCoping	=~	V5	0	0.702	0.068	10.322	0.000	0.702	0.828
## 10	EmotionFocusedCoping	=~	V7	0	0.695	0.077	9.016	0.000	0.695	0.689
## 11	HelpsSeekingBehavior	~~	ProblemFocusedCoping	0	0.081	0.078	1.040	0.298	0.081	0.081
## 12	HelpsSeekingBehavior	~~	EmotionFocusedCoping	0	0.384	0.074	5.182	0.000	0.384	0.384
## 13	ProblemFocusedCoping	~~	EmotionFocusedCoping	0	0.449	0.073	6.178	0.000	0.449	0.449
## 24	HelpsSeekingBehavior	~~	HelpsSeekingBehavior	0	1.000	0.000	NA	NA	1.000	1.000
## 25	ProblemFocusedCoping	~~	ProblemFocusedCoping	0	1.000	0.000	NA	NA	1.000	1.000
## 26	EmotionFocusedCoping	~~	EmotionFocusedCoping	0	1.000	0.000	NA	NA	1.000	1.000

- Since the scales are different, the estimates (and their respected standard error) produced by the two identification methods do not match.
- Parameters that are significant in uli are also significant in uvi; the result of Wald test for significance are the same.
- Standardised latent variables and complete standardised solutions are equal.

Compare goodness-of-fit

```
identical(round(uliSummary$FIT, 3), round(uviSummary$FIT, 3))
```

```
## [1] TRUE
```

- Both identification methods yields identical Chi-square goodness-of-fit test, residuals, and other goodness-of-fit indices.

d)

Goodness-of-fit evaluation

We evaluate the indices according to Hooper, Coughlan, & Mullen (2008).

```
uliSummary$FIT
```

##	npar	fmin	chisq	df	pvalue
##	23.000	0.523	232.378	32.000	0.000
##	baseline.chisq	baseline.df	baseline.pvalue	cfi	tli
##	956.825	45.000	0.000	0.780	0.691
##	logl	unrestricted.logl	aic	bic	ntotal
##	-2398.615	-2282.426	4843.230	4921.492	222.000
##	bic2	rmsea	rmsea.ci.lower	rmsea.ci.upper	rmsea.pvalue
##	4848.603	0.168	0.148	0.189	0.000
##	srmr				
##	0.092				

- $H_0: \Sigma = \Sigma(\theta)$
- Chi-square test statistics: 232.378; p-value: 0.000. H_0 is rejected at $\alpha = .95$.
- NNFI: $0.691 < 0.95$
- CFI: $0.780 < 0.95$
- RMSEA: $0.168 > 0.07$
- SRMR: $0.092 > 0.08$

Neither the goodness-of-fit test or the fit indices pass the acceptable threshold levels. The proposed model is of poor fit.

e)

Modification indices

```
mi1 <- modindices(uli, sort. = TRUE)
head(mi1, 2)[, -which(names(mi1) == "sepc.nox")]
```

##	lhs	op	rhs	mi	epc	sepc.lv	sepc.all
##	48	V1	~~	V6	56.721	0.195	0.195
##	57	V3	~~	V9	45.750	0.360	0.360

- Modification indices suggests there exist error covariance between V_1 and V_6 .
- This modification is justifiable as these indicators belong to the same latent factor according to the proposed model. Hence they are likely to subject to the same type of variance.

New model

```

newModel1 <- 'HelpsSeekingBehavior =~ V1 + V3 + V6 + V9;
ProblemFocusedCoping =~ V2 + V4 + V8 + V10;
EmotionFocusedCoping =~ V5 + V7;

HelpsSeekingBehavior ~~ ProblemFocusedCoping;
HelpsSeekingBehavior ~~ EmotionFocusedCoping;
ProblemFocusedCoping ~~ EmotionFocusedCoping;

V1~~V6'

uliNew1 <- lavaan(newModel1, data=teachers, auto.var=TRUE, auto.fix.first=TRUE, std.lv=FALSE)
uliNew1Summary <- summary(uliNew1, fit.measures=TRUE, standardized=TRUE, rsquare=TRUE)

```

New model parameter fitting

```
uliNew1Summary$PE[,~which(colnames(uliNew1Summary$PE) == "std.noxx")]
```

##	lhs op	rhs exo	est	se	z	pvalue
## 1	HelpsSeekingBehavior =~	V1	0 1.00000000	0.00000000	NA	NA
## 2	HelpsSeekingBehavior =~	V3	0 1.57245218	0.16304299	9.6444025	0.000000e+00
## 3	HelpsSeekingBehavior =~	V6	0 0.93688873	0.08989286	10.4222821	0.000000e+00
## 4	HelpsSeekingBehavior =~	V9	0 1.96588607	0.20307912	9.6803950	0.000000e+00
## 5	ProblemFocusedCoping =~	V2	0 1.00000000	0.00000000	NA	NA
## 6	ProblemFocusedCoping =~	V4	0 0.97877311	0.10125845	9.6660886	0.000000e+00
## 7	ProblemFocusedCoping =~	V8	0 0.49272481	0.07178734	6.8636726	6.711298e-12
## 8	ProblemFocusedCoping =~	V10	0 0.61839598	0.09003318	6.8685341	6.486589e-12
## 9	EmotionFocusedCoping =~	V5	0 1.00000000	0.00000000	NA	NA
## 10	EmotionFocusedCoping =~	V7	0 0.89173557	0.14562667	6.1234359	9.157870e-10
## 11	HelpsSeekingBehavior ~~ ProblemFocusedCoping		0 0.02480077	0.02304875	1.0760138	2.819211e-01
## 12	HelpsSeekingBehavior ~~ EmotionFocusedCoping		0 0.12244075	0.03015204	4.0607782	4.890942e-05
## 13	ProblemFocusedCoping ~~ EmotionFocusedCoping		0 0.22377420	0.04687993	4.7733477	1.811886e-06
## 14	V1 ~~	V6	0 0.19213158	0.03143827	6.1113924	9.876562e-10
## 15	V1 ~~	V1	0 0.28101558	0.03056538	9.1939163	0.000000e+00
## 16	V3 ~~	V3	0 0.55643982	0.06437812	8.6433064	0.000000e+00
## 17	V6 ~~	V6	0 0.49377595	0.04877951	10.1226105	0.000000e+00
## 18	V9 ~~	V9	0 0.01128643	0.05740288	0.1966178	8.441267e-01
## 19	V2 ~~	V2	0 0.38704382	0.05208436	7.4310948	1.076916e-13
## 20	V4 ~~	V4	0 0.11904211	0.03581781	3.3235451	8.888106e-04
## 21	V8 ~~	V8	0 0.32175470	0.03272460	9.8321970	0.000000e+00
## 22	V10 ~~	V10	0 0.50584806	0.05145592	9.8307070	0.000000e+00
## 23	V5 ~~	V5	0 0.17048374	0.08114393	2.1010043	3.564059e-02
## 24	V7 ~~	V7	0 0.58278054	0.08400991	6.9370450	4.003908e-12
## 25	HelpsSeekingBehavior ~~ HelpsSeekingBehavior		0 0.21490578	0.04155745	5.1712935	2.324791e-07
## 26	ProblemFocusedCoping ~~ ProblemFocusedCoping		0 0.44626959	0.07918072	5.6360893	1.739552e-08
## 27	EmotionFocusedCoping ~~ EmotionFocusedCoping		0 0.54723408	0.10344737	5.2899756	1.223327e-07
## 28	V1 r2	V1	0 0.43334649	NA	NA	NA
## 29	V3 r2	V3	0 0.48848036	NA	NA	NA
## 30	V6 r2	V6	0 0.27642519	NA	NA	NA
## 31	V9 r2	V9	0 0.98659306	NA	NA	NA
## 32	V2 r2	V2	0 0.53553631	NA	NA	NA
## 33	V4 r2	V4	0 0.78220032	NA	NA	NA

```

## 34          V8 r2          V8  0 0.25190553          NA          NA          NA
## 35          V10 r2         V10  0 0.25226554          NA          NA          NA
## 36          V5 r2          V5  0 0.76246411          NA          NA          NA
## 37          V7 r2          V7  0 0.42748853          NA          NA          NA
##      std.lv      std.all
## 1  0.46357931 0.65829058
## 2  0.72895630 0.69891370
## 3  0.43432223 0.52576154
## 4  0.91134411 0.99327391
## 5  0.66803413 0.73180347
## 6  0.65385384 0.88442089
## 7  0.32915699 0.50190191
## 8  0.41310962 0.50226043
## 9  0.73975272 0.87319191
## 10 0.65966381 0.65382607
## 11 0.08008340 0.08008340
## 12 0.35703875 0.35703875
## 13 0.45281916 0.45281916
## 14 0.19213158 0.51578486
## 15 0.28101558 0.56665351
## 16 0.55643982 0.51151964
## 17 0.49377595 0.72357481
## 18 0.01128643 0.01340694
## 19 0.38704382 0.46446369
## 20 0.11904211 0.21779968
## 21 0.32175470 0.74809447
## 22 0.50584806 0.74773446
## 23 0.17048374 0.23753589
## 24 0.58278054 0.57251147
## 25 1.00000000 1.00000000
## 26 1.00000000 1.00000000
## 27 1.00000000 1.00000000
## 28          NA          NA
## 29          NA          NA
## 30          NA          NA
## 31          NA          NA
## 32          NA          NA
## 33          NA          NA
## 34          NA          NA
## 35          NA          NA
## 36          NA          NA
## 37          NA          NA

```

- Result for Wald tests of significance for the parameters are the same as the old model.
- The parameter for the error covariance between V_1 and V_6 is also significant.

Likelihood ratio test

```
lavTestLRT(uli, uliNew1)
```

```
## Chi-Squared Difference Test
##
```



```
##           Df      AIC      BIC  Chisq Chisq diff Df diff Pr(>Chisq)
## uliNew1 31 4789.9 4871.6 177.10
## uli      32 4843.2 4921.5 232.38      55.282      1 1.044e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

- Likelihood ratio test reveals significant difference between models with and without error co-variance between V_1 and V_6 .
- Hence the new model is a better model.

Goodness-of-fit of the new model

- $H_0: \Sigma = \Sigma(\theta)$
- Chi-square test statistics: 177.096; p-value: 0.000. H_0 is rejected at $\alpha = .95$.
- NNFI: $0.767 < 0.95$
- CFI: $0.840 < 0.95$
- RMSEA: $0.146 > 0.07$
- SRMR: $0.087 > 0.08$

The new model has better commonly used goodness-of-fit indices (closer to recommended cutoff) across all measures. However, non of which passes the recommended cutoff. Chi-square goodness-of-fit test has a lower test statistic but still rejects H_0 .

Further modification indices

```
mi2 <- modindices(uliNew1, sort. = TRUE)
head(mi2, 3)[, -which(colnames(mi2) == "sepc.no")]
```

```
##      lhs op rhs      mi      epc sepc.lv sepc.all
## 89 V10 ~~ V5 26.827 0.168 0.168 0.572
## 65 V6  ~~ V2 24.755 0.138 0.138 0.315
## 69 V6  ~~ V5 23.344 -0.128 -0.128 -0.440
```

The top expected reduction in chi-square statistic is 26.827. The current chi-square statistic is 177.096. Hence, even if another parameter is added into the model, the chi-square statistic is expected to be around: $177.096 - 26.827 = 150.269$, which is still far from the required 43.773 critical value evaluated by `qchisq(p=.95, df=30.000)`.

One could endlessly add new parameter according to the modification indices until a non-significant Chi-square test statistic is reached. This is highly data driven and defeats much the purpose of the analysis.

In summary, the proposed model may not be a good fit.

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Official Statistics and Structural
Equation Modeling

Course title