

Psychology 454: Latent Variable Modeling

Using the lavaan package for latent variable modeling

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Outline

- 1 lavaan: LAtent VArIable ANalysis
- 2 Confirmatory models
 - Comparisons with EFA and sem
 - Graphic output
- 3 Confirmatory cfa for multiple groups
 - Two groups from covariance matrices
 - 9 cognitive variables from Holzinger-Swineford, 1939
- 4 References

3 major structural modeling programs in R

- sem (by John Fox)
 - Uses ram notation for parameters
 - *psych* will work as a front end for developing parameters
 - Development work seems to have switched to OpenMx
 - Will not do multiple groups
- lavaan (by Yves Rosseel)
 - Uses a more compact notation than sem
 - Will work on multiple groups
 - Still under development
- OpenMx (by Michael Neal, Steve Boker and the OpenMx group)
 - Very powerful structural equation package
 - Based upon Mx (developed for behavioral geneticists)
 - Somewhat idiosyncratic syntax

Getting lavaan

- Beta version (0.4-5) may be downloaded from lavaan website.
 - Will handle covariance matrices – objects to but will run correlations matrices

```
install.packages("lavaan", repos="http://www.da.ugent.be", type="source")  
library(lavaan)
```

- R version on CRAN is 0.3-3.

```
install.packages("lavaan")  
library(lavaan)
```

- Will not handle covariance or correlation matrices
- Documentation is also available at http://users.ugent.be/~yrosseel/lavaan/lavaan_usersguide_0.3-1.pdf
- For more information about *lavaan* go to <http://lavaan.ugent.be/>

Using lavaan

- Confirmatory factoring models with cfa
 - Single group
 - Multiple group (factor invariance issues)
- Structural Equation Models with sem
 - Single group models
 - Regression models
 - Complex regression models
 - latent variable models

Confirmatory models for a Thurstone data set – Bechtoldt.1 and then Becholdt.2

- Bechtoldt (1961) split a data set from Thurstone & Thurstone (1941) into two equal parts (N=212, 213) to examine factor stability.
 - One set has become known as the “Thurstone” data set in SAS and in McDonald (1999).
 - Both are available in the *psych* package and can be analyzed using *cfa* in *lavaan*
- The following script forms two subsets (b2 is equivalent to “Thurstone”) and then does a *cfa*

```
data(bifactor)}
b1 <- Bechtoldt.1[c(3:8,15:17),c(3:8,15:17)]
b2 <- Bechtoldt.2[c(3:8,15:17),c(3:8,15:17)]
Thurstone.mod <- ' F1 =~ Sentences + Vocabulary + Completion
                  F2 =~ First_Letters + Four_letter_words + Suffixes
                  F3 =~ Letter_Series + Pedigrees + Letter_Grouping'
t.cfa.2 <- cfa(Thurstone.mod,sample.cov=b2,sample.nobs=213,std.lv=TRUE)
summary(t.cfa.2)
```

lavaan output for a cfa – first a warning

```
> t.cfa.2 <- cfa(Thurstone.mod,sample.cov=b2,  
                sample.nobs=213,std.lv=TRUE)
```

Warning message:

```
In Sample(data = data, group = group, sample.cov = sample.cov,  
          sample.mean = sample.mean,  :
```

sample covariance matrix looks like a correlation matrix!

lavaan currently does not support the analysis of correlation matrices; the standard errors in the summary output will be most likely wrong; see the following reference:

Cudeck, R. (1989). Analysis of correlation matrices using covariance structure models. *Psychological Bulletin*, 105, 317-327.

Limited output unless requested

```
> summary(t.cfa.2)
```

```
Lavaan (0.4-5) converged normally after 28 iterations
```

Number of observations	213
Estimator	ML
Minimum Function Chi-square	38.376
Degrees of freedom	24
P-value	0.032

More complete output

```
> summary(t.cfa.2,fit.measures=TRUE)
```

Lavaan (0.4-5) converged normally after 28 iterations

Loglikelihood and Information Criteria:

Number of observations	213	Loglikelihood user model (H0)	-2181.238
Estimator	ML	Loglikelihood unrestricted model (H1)	-2162.050
Minimum Function Chi-square	38.376	Number of free parameters	21
Degrees of freedom	24	Akaike (AIC)	4404.476
P-value	0.032	Bayesian (BIC)	4475.063
		Sample-size adjusted Bayesian (BIC)	4408.520

Chi-square test baseline model:

Root Mean Square Error of Approximation:

Minimum Function Chi-square	1107.090	RMSEA	0.053
Degrees of freedom	36	90 Percent Confidence Interval	0.016 0.083
P-value	0.000	P-value RMSEA <= 0.05	0.404

Full model versus baseline model:

Standardized Root Mean Square Residual:

Comparative Fit Index (CFI)	0.987	SRMR	0.044
Tucker-Lewis Index (TLI)	0.980		

With parameter estimates - notice that we fixed latent variances to 1

Parameter estimates:

Information					Expected				
Standard Errors					Standard				
	Estimate	Std.err	Z-value	P(> z)					
Latent variables:					Variances:				
F1 =~					Sentences	0.181	0.028	6.388	0.000
Sentences	0.903	0.054	16.727	0.000	Vocabulary	0.164	0.028	5.953	0.000
Vocabulary	0.912	0.054	17.005	0.000	Completion	0.266	0.033	8.026	0.000
Completion	0.854	0.056	15.317	0.000	First_Letters	0.300	0.051	5.923	0.000
F2 =~					Four_letter_w	0.363	0.052	6.941	0.000
First_Letters	0.834	0.060	13.783	0.000	Suffixes	0.504	0.059	8.513	0.000
Four_letter_w	0.795	0.061	12.937	0.000	Letter_Series	0.388	0.059	6.594	0.000
Suffixes	0.701	0.064	10.960	0.000	Pedigrees	0.479	0.062	7.751	0.000
F3 =~					Letter_Groupi	0.503	0.063	7.995	0.000
Letter_Series	0.779	0.064	12.173	0.000	F1	1.000			
Pedigrees	0.718	0.065	10.998	0.000	F2	1.000			
Letter_Groupi	0.702	0.066	10.679	0.000	F3	1.000			
Covariances:									
F1 ~~									
F2	0.643	0.050	12.755	0.000					
F3	0.670	0.051	13.153	0.000					
F2 ~~									
F3	0.637	0.058	10.900	0.000					

Alternative parameterization one variable path per latent set to 1

```
summary(t.cfa.2,fit.measures=TRUE)
```

Latent variables:

F1 =~

	Estimate	Std.err	Z-value	P(> z)
Sentences	1.000			
Vocabulary	1.010	0.051	19.938	0.000
Completion	0.946	0.054	17.644	0.000

F2 =~

First_Letters	1.000			
Four_letter_w	0.954	0.082	11.668	0.000
Suffixes	0.841	0.081	10.326	0.000

F3 =~

Letter_Series	1.000			
Pedigrees	0.922	0.097	9.469	0.000
Letter_Groupi	0.901	0.097	9.288	0.000

Covariances:

F1 ~~

F2	0.484	0.072	6.751	0.000
F3	0.471	0.071	6.653	0.000

F2 ~~

F3	0.414	0.068	6.118	0.000
----	-------	-------	-------	-------

Variances:

Sentences	0.181	0.028	6.388	0.000
Vocabulary	0.164	0.028	5.953	0.000
Completion	0.266	0.033	8.026	0.000
First_Letters	0.300	0.051	5.923	0.000
Four_letter_w	0.363	0.052	6.941	0.000
Suffixes	0.504	0.059	8.513	0.000
Letter_Series	0.388	0.059	6.594	0.000
Pedigrees	0.479	0.062	7.751	0.000
Letter_Groupi	0.503	0.063	7.995	0.000
F1	0.815	0.097	8.363	0.000
F2	0.695	0.101	6.891	0.000
F3	0.607	0.100	6.087	0.000

Compare to the efa from psych and sem from sem

- This data set has been discussed before (many times, see e.g., Week 4)
 - We compared methods of factor extraction (minres and mle) and rotation (varimax and oblimin)
 - We compared EFA and SEM solutions
- Now compare those solutions to the *lavaan* solutions
- Both in ease of set up and in statistical modeling

create the sem commands by using psych

```
f3 <- fa(Thurstone,3,fm='mle')
mod3 <- structure.diagram(f3,cut=.45,errors=TRUE)
mod3
```

	Path	Parameter	Value
[1,]	"ML1->V1"	"F1V1"	NA
[2,]	"ML1->V2"	"F1V2"	NA
[3,]	"ML1->V3"	"F1V3"	NA
[4,]	"ML2->V4"	"F2V4"	NA
[5,]	"ML2->V5"	"F2V5"	NA
[6,]	"ML2->V6"	"F2V6"	NA
[7,]	"ML3->V7"	"F3V7"	NA
[8,]	"ML3->V8"	"F3V8"	NA
[9,]	"ML3->V9"	"F3V9"	NA
[10,]	"V1<->V1"	"x1e"	NA
[11,]	"V2<->V2"	"x2e"	NA
...			
[18,]	"V9<->V9"	"x9e"	NA
[19,]	"ML2<->ML1"	"rF2F1"	NA
[20,]	"ML3<->ML1"	"rF3F1"	NA
[21,]	"ML3<->ML2"	"rF3F2"	NA
[22,]	"ML1<->ML1"	NA	"1"
[23,]	"ML2<->ML2"	NA	"1"
[24,]	"ML3<->ML3"	NA	"1"

Running sem

```
> rownames(Thurstone) <- colnames(Thurstone) #to get the names to match the modl
> sem3 <- sem(mod3,Thurstone,N=213)
> summary(sem3,digits=2)
```

```
Model Chisquare = 38   Df = 24 Pr(>Chisq) = 0.033
Chisquare (null model) = 1102   Df = 36
Goodness-of-fit index = 0.96
Adjusted goodness-of-fit index = 0.92
RMSEA index = 0.053   90% CI: (0.015, 0.083)
Bentler-Bonnett NFI = 0.97
Tucker-Lewis NNFI = 0.98
Bentler CFI = 0.99
SRMR = 0.044
BIC = -90
```

Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.97	-0.42	0.00	0.04	0.09	1.63

With parameter estimates

Parameter Estimates

	Estimate	Std Error	z value	Pr(> z)	
F1V1	0.90	0.054	16.7	0.0e+00	V1 <--- ML1
F1V2	0.91	0.054	17.0	0.0e+00	V2 <--- ML1
F1V3	0.86	0.056	15.3	0.0e+00	V3 <--- ML1
F2V4	0.84	0.061	13.8	0.0e+00	V4 <--- ML2
F2V5	0.80	0.062	12.9	0.0e+00	V5 <--- ML2
F2V6	0.70	0.064	10.9	0.0e+00	V6 <--- ML2
F3V7	0.78	0.065	12.0	0.0e+00	V7 <--- ML3
F3V8	0.72	0.067	10.7	0.0e+00	V8 <--- ML3
F3V9	0.70	0.067	10.5	0.0e+00	V9 <--- ML3
x1e	0.18	0.028	6.4	1.7e-10	V1 <--> V1
x2e	0.16	0.028	5.9	3.0e-09	V2 <--> V2
x3e	0.27	0.033	8.0	1.6e-15	V3 <--> V3
x4e	0.30	0.051	5.9	2.7e-09	V4 <--> V4
x5e	0.36	0.052	7.0	3.4e-12	V5 <--> V5
x6e	0.51	0.060	8.4	0.0e+00	V6 <--> V6
x7e	0.39	0.062	6.3	2.3e-10	V7 <--> V7
x8e	0.48	0.065	7.4	1.8e-13	V8 <--> V8
x9e	0.51	0.065	7.7	9.5e-15	V9 <--> V9
rF2F1	0.64	0.051	12.6	0.0e+00	ML1 <--> ML2
rF3F1	0.67	0.054	12.5	0.0e+00	ML1 <--> ML3
rF3F2	0.64	0.059	10.7	0.0e+00	ML2 <--> ML3

A direct comparison of statistical estimates

Model Chisquare = 38 Df = 24 Pr(>Chisq) = 0.033
 Chisquare (null model) = 1102 Df = 36
 Goodness-of-fit index = 0.96
 Adjusted goodness-of-fit index = 0.92
 RMSEA index = 0.053 90% CI: (0.015, 0.083)
 Bentler-Bonnett NFI = 0.97
 Tucker-Lewis NNFI = 0.98
 Bentler CFI = 0.99
 SRMR = 0.044
 BIC = -90

Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.97	-0.42	0.00	0.04	0.09	1.63

Number of observations	213
Estimator	ML
Minimum Function Chi-square	38.376
Degrees of freedom	24
P-value	0.032
Chi-square test baseline model:	
Minimum Function Chi-square	1107.090
Degrees of freedom	36
P-value	0.000
Full model versus baseline model:	
Comparative Fit Index (CFI)	0.987
Tucker-Lewis Index (TLI)	0.980
Root Mean Square Error of Approximation:	
RMSEA	0.053
90 Percent Confidence Interval	0.016 0.083
P-value RMSEA <= 0.05	0.404
Standardized Root Mean Square Residual:	
SRMR	0.044

A direct comparison of parameter estimates

sem

Parameter Estimates					
	Estimate	Std Error	z value	Pr(> z)	
F1V1	0.90	0.054	16.7	0.0e+00	V1 <--- ML1
F1V2	0.91	0.054	17.0	0.0e+00	V2 <--- ML1
F1V3	0.86	0.056	15.3	0.0e+00	V3 <--- ML1
F2V4	0.84	0.061	13.8	0.0e+00	V4 <--- ML2
F2V5	0.80	0.062	12.9	0.0e+00	V5 <--- ML2
F2V6	0.70	0.064	10.9	0.0e+00	V6 <--- ML2
F3V7	0.78	0.065	12.0	0.0e+00	V7 <--- ML3
F3V8	0.72	0.067	10.7	0.0e+00	V8 <--- ML3
F3V9	0.70	0.067	10.5	0.0e+00	V9 <--- ML3
x1e	0.18	0.028	6.4	1.7e-10	V1 <--> V1
x2e	0.16	0.028	5.9	3.0e-09	V2 <--> V2
x3e	0.27	0.033	8.0	1.6e-15	V3 <--> V3
x4e	0.30	0.051	5.9	2.7e-09	V4 <--> V4
x5e	0.36	0.052	7.0	3.4e-12	V5 <--> V5
x6e	0.51	0.060	8.4	0.0e+00	V6 <--> V6
x7e	0.39	0.062	6.3	2.3e-10	V7 <--> V7
x8e	0.48	0.065	7.4	1.8e-13	V8 <--> V8
x9e	0.51	0.065	7.7	9.5e-15	V9 <--> V9
rF2F1	0.64	0.051	12.6	0.0e+00	ML1 <--> ML2
rF3F1	0.67	0.054	12.5	0.0e+00	ML1 <--> ML3
rF3F2	0.64	0.059	10.7	0.0e+00	ML2 <--> ML3

lavaan

Latent variables:

F1 =~

Sentences	0.903	0.054	16.727	0.000
Vocabulary	0.912	0.054	17.005	0.000
Completion	0.854	0.056	15.317	0.000

F2 =~

First_Letters	0.834	0.060	13.783	0.000
Four_letter_w	0.795	0.061	12.937	0.000
Suffixes	0.701	0.064	10.960	0.000

F3 =~

Letter_Series	0.779	0.064	12.173	0.000
Pedigrees	0.718	0.065	10.998	0.000
Letter_Groupi	0.702	0.066	10.679	0.000

Covariances:

F1 ~~

F2	0.643	0.050	12.755	0.000
F3	0.670	0.051	13.153	0.000

F2 ~~

F3	0.637	0.058	10.900	0.000
----	-------	-------	--------	-------

lavaan.diagram

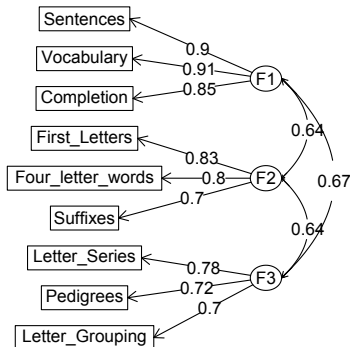
Currently, *lavaan* does not draw structural diagrams. But, it is not hard to form a simple function to draw lavaan diagrams from lavaan output using tools from *psych*.

```
"lavaan.diagram" <-
function(fit,model="cfa",...) {
#if (is.null(fit@Model@GLIST[[1]]$beta)) {model <- "cfa"} else {model <- "sem"}
if(model=="cfa") {fx=fit@Model@GLIST$lambda
                  colnames(fx) <- fit@Model@dimNames$lambda[[2]]
                  Phi <- fit@Model@GLIST$psi
                  Rx <- fit@Model@GLIST$theta
                  v.labels <- fit@Model@dimNames$lambda[[1]]
structure.diagram(fx=fx,Phi=Phi,Rx=Rx,labels=v.labels,...)}
else {structure.diagram(fx=fit@Model@GLIST$lambda,Phi=fit@Model@GLIST$beta,
                        Rx=fit@Model@GLIST$theta,...) }
```

This function is not ready for prime time because it does not yet draw sem (just cfa) diagrams.

lavaan diagram for the Thurstone (Bechtoldt.2) data set

Structural model



Confirmatory factor structures across groups

- When comparing measures across age or across genders, it is important to make sure that the factor structures are in fact the same.
 - When measuring change, we want to make sure that our measure is the same for different ages.
 - When comparing ethnic groups, gender, genetic relationships, want to make sure that the measures are invariant across the groups
- This can be done by doing multiple group cfa.
- Possible to do in OpenMx and lavaan, but not in sem

Comparing Bechtoldt1 and Bechtoldt2

```
two.mod <- cfa(Thurstone.mod,sample.cov=list(b1,b2),
              sample.nobs=list(212,213),std.lv=TRUE)
> summary(two.mod,fit.measures=TRUE)
```

Model converged normally after 26 iterations using ML

Minimum Function Chi-square	74.045
Degrees of freedom	48
P-value	0.0093

Chi-square for each group:

Group 1	35.669
Group 2	38.376

Chi-square test baseline model:

Minimum Function Chi-square	2205.154
Degrees of freedom	63
P-value	0.0000

Does not seem to work with lavaan beta– need to use the old

Two groups from covariance matrices

Loadings for two groups

Group 1 [Group 1]:

	Estimate	Std.err	Z-value	P(> z)
Latent variables:				
F1 =~				
Sentences	0.907	0.054	16.800	0.000
Vocabulary	0.913	0.054	16.992	0.000
Completion	0.840	0.056	14.890	0.000
F2 =~				
First_Letters	0.829	0.064	12.939	0.000
Four_letter_words	0.731	0.066	11.126	0.000
Suffixes	0.650	0.067	9.668	0.000
F3 =~				
Letter_Series	0.847	0.060	14.206	0.000
Pedigrees	0.788	0.061	12.872	0.000
Letter_Grouping	0.711	0.063	11.202	0.000

Latent covariances:

F1 ~~				
F2	0.565	0.058	9.668	0.000
F3	0.700	0.045	15.528	0.000
F2 ~~				
F3	0.570	0.062	9.137	0.000

Latent variances:

F1	1.000
F2	1.000
F3	1.000

Residual variances:

Sentences	0.173	0.028	6.137	0.000
-----------	-------	-------	-------	-------

Group 2 [Group 2]:

	Estimate	Std.err	Z-value	P(> z)
Latent variables:				
F1 =~				
Sentences	0.903	0.054	16.727	0.000
Vocabulary	0.912	0.054	17.005	0.000
Completion	0.854	0.056	15.317	0.000
F2 =~				
First_Letters	0.834	0.060	13.783	0.000
Four_letter_words	0.795	0.061	12.937	0.000
Suffixes	0.701	0.064	10.960	0.000
F3 =~				
Letter_Series	0.779	0.064	12.173	0.000
Pedigrees	0.718	0.065	10.998	0.000
Letter_Grouping	0.702	0.066	10.679	0.000

Latent covariances:

F1 ~~				
F2	0.643	0.050	12.755	0.000
F3	0.670	0.051	13.153	0.000
F2 ~~				
F3	0.637	0.058	10.900	0.000

Latent variances:

F1	1.000
F2	1.000
F3	1.000

Residual variances:

Sentences	0.173	0.028	6.137	0.000
-----------	-------	-------	-------	-------

Two groups from covariance matrices

Constrain the two groups to be equal

```
two.mod <- cfa(Thurstone.mod,sample.cov=list(b1,b2),
               sample.nobs=list(212,213),std.lv=TRUE,
               group.constraints=c("loadings"))
summary(two.mod,fit.measures=TRUE)
```

Model converged normally after 25 iterations using ML

Minimum Function Chi-square	76.128
Degrees of freedom	57
P-value	0.0461

Chi-square for each group:

Group 1	36.700
Group 2	39.428

Chi-square test baseline model:

Minimum Function Chi-square	2205.154
Degrees of freedom	63
P-value	0.0000

Full model versus baseline model:

Comparative Fit Index (CFI)	0.991
Tucker-Lewis Index (TLI)	0.990

Two groups from covariance matrices

Parameter estimates

Model estimates:

Group 1 [Group 1]:

	Estimate	Std.err	Z-value	P(> z)
Latent variables:				
F1 =~				
Sentences	0.903	0.038	23.705	0.000
Vocabulary	0.911	0.038	24.046	0.000
Completion	0.846	0.040	21.362	0.000
F2 =~				
First_Letters	0.831	0.044	18.943	0.000
Four_letter_words	0.767	0.045	17.120	0.000
Suffixes	0.679	0.046	14.674	0.000
F3 =~				
Letter_Series	0.816	0.044	18.752	0.000
Pedigrees	0.756	0.045	16.976	0.000
Letter_Grouping	0.705	0.046	15.465	0.000

Latent covariances:

F1 ~~				
F2	0.565	0.056	10.044	0.000
F3	0.697	0.044	15.746	0.000
F2 ~~				
F3	0.569	0.061	9.304	0.000

Latent variances:

F1	1.000
F2	1.000
F3	1.000

Group 2 [Group 2]:

Latent variables:

	Estimate	Std.err	Z-value	P(> z)
Latent variables:				
F1 =~				
Sentences	0.903			
Vocabulary	0.911			
Completion	0.846			
F2 =~				
First_Letters	0.831			
Four_letter_words	0.767			
Suffixes	0.679			
F3 =~				
Letter_Series	0.816			
Pedigrees	0.756			
Letter_Grouping	0.705			

Latent covariances:

F1 ~~				
F2	0.641	0.049	13.108	0.000
F3	0.672	0.048	13.939	0.000
F2 ~~				
F3	0.633	0.057	11.145	0.000

Latent variances:

F1	1.000
F2	1.000
F3	1.000

Compare goodness of fits

Because the models are in fact samples from the same data, they should agree.

Model converged normally after 26 iterations using ML

Minimum Function Chi-square	74.045
Degrees of freedom	48
P-value	0.0093

Chi-square for each group:

Group 1	35.669
Group 2	38.376

Chi-square test baseline model:

Minimum Function Chi-square	2205.154
Degrees of freedom	63
P-value	0.0000

Full model versus baseline model:

Comparative Fit Index (CFI)	0.988
Tucker-Lewis Index (TLI)	0.984

Model converged normally after 25 iterations using ML

Minimum Function Chi-square	76.128
Degrees of freedom	57
P-value	0.0461

Chi-square for each group:

Group 1	36.700
Group 2	39.428

Chi-square test baseline model:

Minimum Function Chi-square	2205.154
Degrees of freedom	63
P-value	0.0000

Full model versus baseline model:

Comparative Fit Index (CFI)	0.991
Tucker-Lewis Index (TLI)	0.990

9 cognitive variables from Holzinger-Swineford, 1939

Descriptive statistics of their data set

```
> describe(HolzingerSwineford1939)
```

	var	n	mean	sd	median	trimmed	mad	min	max	range	skew	kurtosis	se
id	1	301	176.55	105.94	163.00	176.78	140.85	1.00	351.00	350.00	-0.01	-1.35	6.11
sex	2	301	1.51	0.50	2.00	1.52	0.00	1.00	2.00	1.00	-0.06	-2.01	0.03
ageyr	3	301	13.00	1.05	13.00	12.89	1.48	11.00	16.00	5.00	0.69	0.25	0.06
agemo	4	301	5.38	3.45	5.00	5.32	4.45	0.00	11.00	11.00	0.09	-1.21	0.20
school*	5	301	1.52	0.50	2.00	1.52	0.00	1.00	2.00	1.00	-0.07	-2.01	0.03
grade	6	300	7.48	0.50	7.00	7.47	0.00	7.00	8.00	1.00	0.09	-2.00	0.03
x1	7	301	4.94	1.17	5.00	4.96	1.24	0.67	8.50	7.83	-0.25	0.36	0.07
x2	8	301	6.09	1.18	6.00	6.02	1.11	2.25	9.25	7.00	0.47	0.38	0.07
x3	9	301	2.25	1.13	2.12	2.20	1.30	0.25	4.50	4.25	0.38	-0.89	0.07
x4	10	301	3.06	1.16	3.00	3.02	0.99	0.00	6.33	6.33	0.27	0.12	0.07
x5	11	301	4.34	1.29	4.50	4.40	1.48	1.00	7.00	6.00	-0.35	-0.53	0.07
x6	12	301	2.19	1.10	2.00	2.09	1.06	0.14	6.14	6.00	0.86	0.88	0.06
x7	13	301	4.19	1.09	4.09	4.16	1.10	1.30	7.43	6.13	0.25	-0.27	0.06
x8	14	301	5.53	1.01	5.50	5.49	0.96	3.05	10.00	6.95	0.53	1.24	0.06
x9	15	301	5.37	1.01	5.42	5.37	0.99	2.78	9.25	6.47	0.20	0.34	0.06

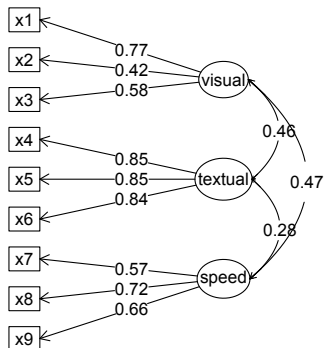
cfa syntax

Because we are using a covariance analysis, we need to standardize the observed variables to express the loadings as correlations.

```
HS.model <- '  
  visual =~ x1 + x2 + x3  
  textual =~ x4 + x5 + x6  
  speed =~ x7 + x8 + x9  
'  
  
fit <- cfa(HS.model, data = HolzingerSwineford1939, std.lv=TRUE, std.ov=TRUE)  
summary(fit)  
lavaan.diagram(fit, cut=.2, digits=2)
```

Lavaan diagram of Holzinger-Swineford 1939 cfa

Structural model



Now do multiple groups

```
fit.2g <- cfa(HS.model, data=HolzingerSwineford1939, group="school",
              std.lv=TRUE, std.ov=TRUE)
```

```
summary(fit.2g)
```

Number of observations per group

Pasteur	156
Grant-White	145

Estimator	ML
Minimum Function Chi-square	115.851
Degrees of freedom	48
P-value	0.000

Chi-square for each group:

Pasteur	64.309
Grant-White	51.542

Parameter estimates:

Information	Expected
Standard Errors	Standard

3 cognitive variables from Holzinger-Swineford, 1939

Compare the parameters for both schools

Group 1 [Pasteur]:

	Estimate	Std.err	Z-value	P(> z)
Latent variables:				
visual =~				
x1	0.884	0.111	7.934	0.000
x2	0.335	0.089	3.753	0.000
x3	0.513	0.093	5.525	0.000
textual =~				
x4	0.821	0.069	11.927	0.000
x5	0.854	0.068	12.604	0.000
x6	0.836	0.068	12.230	0.000
speed =~				
x7	0.545	0.098	5.557	0.000
x8	0.679	0.104	6.531	0.000
x9	0.550	0.098	5.596	0.000

Covariances:

visual ~~				
textual	0.484	0.086	5.600	0.000
speed	0.299	0.109	2.755	0.006
textual ~~				
speed	0.325	0.100	3.256	0.001

Variances:

x1	0.212	0.165	1.286	0.198
x2	0.881	0.104	8.464	0.000
x3	0.731	0.100	7.271	0.000
x4	0.320	0.052	6.138	0.000
x5	0.265	0.050	5.292	0.000

Group 2 [Grant-White]:

	Estimate	Std.err	Z-value	P(> z)
Latent variables:				
visual =~				
x1	0.674	0.090	7.525	0.000
x2	0.515	0.091	5.642	0.000
x3	0.691	0.090	7.711	0.000
textual =~				
x4	0.863	0.070	12.355	0.000
x5	0.826	0.071	11.630	0.000
x6	0.823	0.071	11.572	0.000
speed =~				
x7	0.657	0.084	7.819	0.000
x8	0.793	0.083	9.568	0.000
x9	0.698	0.084	8.357	0.000

Covariances:

visual ~~				
textual	0.541	0.085	6.355	0.000
speed	0.523	0.094	5.562	0.000
textual ~~				
speed	0.336	0.091	3.674	0.000

Variances:

x1	0.538	0.095	5.675	0.000
x2	0.728	0.099	7.339	0.000
x3	0.515	0.095	5.409	0.000
x4	0.249	0.051	4.870	0.000
x5	0.310	0.053	5.812	0.000

Constrain the two schools to have equal loadings

(This works on lavaan 0.3.3 but not the beta version 0.4-5)

```
fit.2g <- cfa(HS.model, data=HolzingerSwineford1939, group="school",
              std.lv=TRUE, std.ov=TRUE, group.constraints=c("loadings"))
summary(fit.2g)
```

Model converged normally after 27 iterations using ML

Minimum Function Chi-square	122.862
Degrees of freedom	57
P-value	0.0000

Chi-square for each group:

Grant-White	54.264
Pasteur	68.598

Show more fit statistics

```
> summary(fit.2g,fit.measures=TRUE)
```

```

                                Full model versus baseline model:
Model converged normally after 27 iterations using ML

    Minimum Function Chi-square      122.862    Comparative Fit Index (CFI)      0.926
    Degrees of freedom                57         Tucker-Lewis Index (TLI)      0.919
    P-value                          0.0000    Loglikelihood and Information Criteria:

Chi-square for each group:

    Grant-White      54.264    Loglikelihood user model (H0)      -3417.421
    Pasteur          68.598    Loglikelihood unrestricted model (H1) -3355.990

                                Akaike (AIC)      6900.841
                                Bayesian (BIC)     7023.176

Chi-square test baseline model:

    Minimum Function Chi-square      957.769    Root Mean Square Error of Approximation:
    Degrees of freedom                63         RMSEA                                0.088
    P-value                          0.0000    90 Percent Confidence Interval      0.066 0.109

                                Standardized Root Mean Square Residual:

                                SRMR                                0.084

```


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Thurstone, L. L. & Thurstone, T. G. (1941). *Factorial studies of intelligence*. Chicago, Ill.: The University of Chicago press.