Confirmatory Factor Analysis

Statsomat.com

08 January 2021

Warning: The automatic computation and interpretation delivered by the Statsomat should not completely replace the classical, made by humans graphical exploratory data analysis and statistical analysis. There may be data cases for which the Statsomat does not deliver the most optimal solution or output interpretation.

Basic Information

Automatic statistics for the file:

File case17.csv

Your selection for the encoding: Auto

Your selection for the decimal character: Auto

Observations (rows with at least one non-missing value): 250 Variables (columns with at least one non-missing value): 4

Variables considered continuous: 4

Variables considered continuous
01
02
O3
04

Warning: Based on an analysis of the variables considered continuous, we suspect outliers in the data. If observations are erroneous, you could drop them and restart the app. Outliers may affect negatively the execution or the results of the CFA. These are the suspected row numbers:

Table 3: Rows With Suspected Outliers

	01	02	03	04
124	-0.0198541	0.3633604	-0.008961	1.194713

Model Syntax

The following table describes the applied model equations in lavaan model syntax, either as entered by you in the text area (denoted by User=1) or established internally (User=0). The last column numbers the free parameters which are estimated.

Left hand side	Operator	Right hand side	User	Free parameter
factor	=~	01	1	0
factor	=~	02	1	1
factor	=~	03	1	2
factor	=~	04	1	3
01	~~	01	О	4
02	~~	02	О	5
03	~~	03	О	6
04	~~	04	О	7
factor	~~	factor	0	8

Assumptions

Open issue

Model Settings

Outputs

Model Fit Summary

lavaan 0.6-7 ended normally after 29 iterations

Estimator	ML
Optimization method	NLMINB
Number of free parameters	8
Number of observations	250
Model Test User Model:	
Test statistic	0.171
Degrees of freedom	2
P-value (Chi-square)	0.918
Model Test Baseline Model:	
Test statistic	511.420

Degrees of freed P-value	lom			6 0.000
User Model versus	Baseline M	Model:		
Comparative Fit Tucker-Lewis Ind		1.000 1.011		
Loglikelihood and	Informatio	on Criteri	.a:	
Loglikelihood us Loglikelihood un			I1)	-114.250 -114.164
Akaike (AIC) Bayesian (BIC) Sample-size adju	sted Bayes	sian (BIC)		244.500 272.672 247.311
Root Mean Square E	Error of Ap	proximati	on:	
RMSEA 90 Percent confi 90 Percent confi P-value RMSEA <=	dence inte			0.000 0.000 0.046 0.955
Standardized Root	Mean Squar	e Residua	ıl:	
SRMR				0.003
Parameter Estimate	es:			
Standard errors Information Information satu	rated (h1)	model	St	Standard Expected ructured
Latent Variables:	Patimata	C+ 4 F	z-value	D(> -)
factor =~	Estimate	Std.Err	z-varue	P(> z)
01	1.000			
02	0.947	0.065	14.626	0.000
03	1.151	0.085	13.581	0.000
04	1.255	0.098	12.820	0.000
Variances:	Patient.	O+ 1 F	1	D(> -)
Ω1	Estimate	Std.Err	z-value	
.01	0.031	0.004	7.744	0.000
.02	0.025	0.003	7.387	0.000
.03 .04	0.055 0.081	0.006	8.639 9.173	0.000
. U I	0.001	0.009	3.113	0.000

factor	0.067	0.009	7.637	0.000

Completely Standardized Parameter Estimates

Latent Variables:

	est.std	Std.Err	z-value	P(> z)	ci.lower	ci.upper
factor =~						
01	0.828	0.027	31.163	0.000	0.776	0.880
02	0.841	0.026	32.825	0.000	0.791	0.891
03	0.788	0.030	26.403	0.000	0.729	0.846
04	0.752	0.033	22.922	0.000	0.688	0.817
Variances:						
	est.std	Std.Err	z-value	P(> z)	ci.lower	ci.upper
.01	0.314	0.044	7.134	0.000	0.228	0.400
.02	0.293	0.043	6.798	0.000	0.208	0.377
.03	0.380	0.047	8.077	0.000	0.287	0.472
.04	0.434	0.049	8.791	0.000	0.337	0.531
factor	1.000				1.000	1.000

Communality

Table 5: Communality

Variable	Communality
01	0.69
02	0.71
03	0.62
04	0.57

Factor Reliability

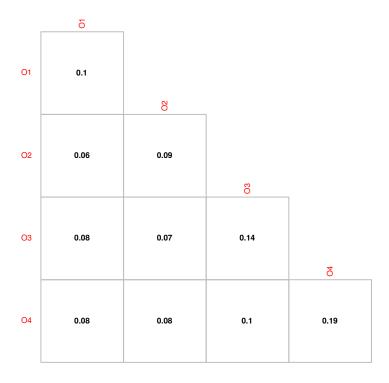
Table 6: Factor Reliability

	factor
Omega (Bentler)	0.87
Omega (McDonald)	0.87
AVE	0.63

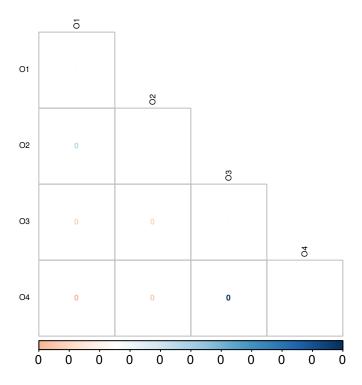
Observed Covariance Matrix

	5			
01	0.1	05		
O2	0.06	0.09	80	
О3	0.08	0.07	0.14	40
O4	0.08	0.08	0.1	0.19

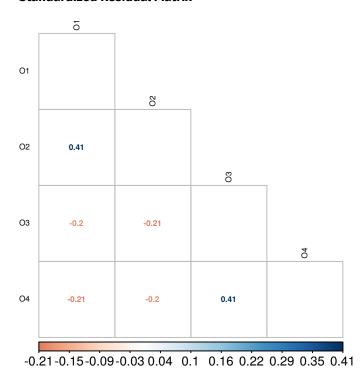
Model-Implied Covariance Matrix



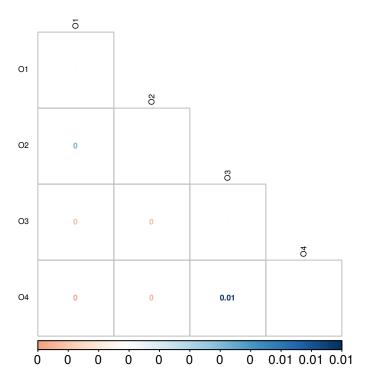
Residual Covariance Matrix



Standardized Residual Matrix



Residual Correlation Matrix



Modification Indices

Table 7: Modification Indices With Respect To Error Covariances

Left	Operator	Right	Modification Index	Expected Parameter Change	Delta	Power	Decision
03	~~	04	0.173	0.003	0.1	1	(nm)
01	~~	02	0.173	0.002	0.1	1	(nm)
02	~~	03	0.045	-0.001	0.1	1	(nm)
01	~~	04	0.045	-0.001	0.1	1	(nm)
01	~~	О3	0.038	-0.001	0.1	1	(nm)
02	~~	04	0.038	-0.001	0.1	1	(nm)

Note:

Maximum 10 modification indices in descending order of their magnitude are listed.

Interpretation

Goodness of Fit Indices

We consider some of the model fit indices from the Model Fit Summary section to check the goodness-of-fit of the model. To decide for an acceptable or non-acceptable model, we apply thresholds considered in the References: [@brown], [@kline].

Model Test User Model

The degrees of freedom are calculated as the number of known parameters minus the number of free parameters: 10

- 8 = 2 . The 2 degrees of freedom indicate an over-identified model, fact which basically enables further analysis and interpretation.

The test statistic with the value 0.171 is called the Chi-square model fit index and represents the difference between summaries of the model-implied covariance matrix and the observed covariance matrix which is hypothesized and desirable to be zero. In general, if the p-value is larger than 0.05 then the test is not statistically significant at 5 % error, the hypothesis cannot be rejected, which would be in favour of the model.

Model Test Baseline Model

The test statistic with the value 511.42 represents the difference between summaries of the baseline model (an alternative model-implied covariance matrix having zero covariances, i.e. a worst fitting model assuming independent variables) and the observed covariance matrix. The p-value of the test of a zero difference is <0.001 suggesting that the baseline model does not fit good to the data. This result is used indirectly in the construction of other model fit indices.

Root Mean Square Error of Approximation:

In our case, the p-value is 0.918 suggesting an acceptable model fit.

The Root Mean Square Error of Approximation (RMSEA) is a fit index based on the chi-square test statistic, which corrects for parsimony, i.e. overly complex models are penalized. RMSEA can be greater or equal than zero, with values close to zero suggesting an acceptable model fit.

In our case, the RMSEA is 0. The upper bound of the 90% confidence interval of the RMSEA is 0.046 and smaller than the threshold value 0.05, suggesting an excellent model fit.

Standardized Root Mean Square Residual:

The Standardized Root Mean Square Residual (SRMR) is a fit index derived from the residual correlation matrix with a range between zero and one with values close to zero suggesting an acceptable model fit.

In our case, the SRMR is 0 which is smaller than the threshold value 0.1 suggesting an acceptable model fit.

User Model versus Baseline Model

The Comparative Fit Index (CFI), evaluates the fit of the the model in relation to the worst-fitting baseline model described above. It ranges between zero and one, with values close to one suggesting good models (in the sense of departure from the baseline model).

In our case, the CFI is 1 which is greater or equal than the threshold value 0.95, suggesting a good model fit.

Similarly to the CFI, the Tucker-Lewis Index (TLI) evaluates the fit of the model in relation to the worst-fitting baseline model described above. Moreover, overly complex models are penalized. Values can range outside zero and one but the index is interpreted similarly to the CFI.

In our case, the TLI is 1.011 which is greater or equal than the threshold value 0.95, suggesting a good model fit.

Summary of the Goodness of Fit Indices

The TLI model fit index suggests an acceptable model fit. Moreover the Chi-square model fit index and the RMSEA suggest an acceptable model fit. We tentatively assume an acceptable model fit and verify this assertion by considering further metrics.

Residuals

We analyze the residual matrices from the Outputs chapter. The residual covariance matrix represents the difference between the observed covariance matrix and the fitted model-implied covariance matrix. Large absolute values indicate local areas of misfit. However, the residuals are affected by the raw metric and are difficult to interpret more precisely.

A better interpretation allows the standardized residual matrix (residuals divided by their estimated asymptotic standard error) and the residual correlation matrix.

There are no variable pairs with standardized residuals which are larger or equal than the considered threshold 2.58 [@brown] or correlation residuals which are larger or equal than the considered threshold 0.1 [@kline]. Therefore, no relationships among the variables are substantially underestimated by the model. There are no variable pairs with standardized residuals which are smaller or equal than the considered threshold -2.58 [@brown] or correlation residuals which are smaller or equal than the considered threshold -0.1 [@kline]. Therefore, no relationships among the variables are substantially overestimated by the model.

Modification Indices

In the interpretation of the modification indices table(s) we rely mostly on [@brown] and [@mi]. We cite from [@brown]: "The modification index reflects an approximation of how much the overall model Chi² will decrease if the fixed or constrained parameter is freely estimated." In other words, if adding a line with a high modification index to the model, i.e. if adding a parameter, the overall goodness-of-fit may be improved. Nevertheless, this should be done only under certain conditions, described in the sequel.

We consider only modification indices greater or equal than 3.84 (which are statistically significant at 5% type I error). Next, we search only for modification indices which achieve a power of minimum 75% in detecting a (relevant) misspecification of at least 0.1 for error or factor correlations, respectively 0.4 for factor loadings. These are characterized in the decision column by the label "epc:m". For more information with regard to the labels of the decision column, please consult the Appendix.

We remark that these conditions are not fulfilled for modification indices with respect to error covariances. Therefore, there exist no significant and relevant modification indices with respect to error covariances.

We remark that there exist no modification indices with respect to factor loadings.

We remark that there exist no modification indices with respect to factor covariances.

Parameter Estimates

Factor Loadings

We remark that the completely standardized factor loadings (section "Completely Standardized Parameter Estimates") are all statistically significant at 5% type I error. Moreover, in absolute value they are all greater than 0.4. This cutoff-value is considered in some CFA research areas a magnitude that is substantively meaningful [@brown]. Please consider also cutoff-values from your particular research area when interpreting the factor loadings. We summarize the interpretation of the completely standardized factor loadings in the next table:

Table 8: Check Completely Standardized Factor Loadings

Latent Variable	Observed Variable	Loading ¹	P-Value	Significant? ²	Relevance ³	Sign ⁴	Check
factor	01	0.83	<0.001	Yes	***	_	Ok
factor	02	0.84	<0.001	Yes	***	_	Ok
factor	03	0.79	<0.001	Yes	**	_	Ok
factor	04	0.75	<0.001	Yes	**	_	Ok

¹ The completely standardized factor loading can be interpreted as the correlation with the factor.

² Completely standardized factor loading significance at 5% type I error.

Table 8: Check Completely Standardized Factor Loadings (continued)

Latent Variable	Observed Variable	Loading ¹	P-Value	Significant? ²	Relevance ³	Sign ⁴	Check
Laterit variable	Observed variable	Louding	, value	Jigiiiiicaiic.	incic varies	7.5	CITCON

³ Stars correspond to factor loadings cutoff-values: 0.4, 0.6, 0.8.

Moreover, we remark that the significance test results for the completely standardized factor loadings from above coincide to those of the unstandardized factor loadings (within section "Model Fit Summary", for non-marker variables).

We proceed by interpreting the (unstandardized) factor loadings from the "Model Fit Summary" section:

Table 9: Interpretation of Unstandardized Factor Loadings

Interpretation of Unstandardized Factor Loadings					
A 1-unit increase in factor leads to a 1.00 -unit increase in the O1					
A 1-unit increase in factor leads to a 0.95 -unit increase in the O2					
A 1-unit increase in factor leads to a 1.15 -unit increase in the O3					
A 1-unit increase in factor leads to a 1.26 -unit increase in the O4					

Error Variances

We summarize the interpretation of the error variances and communalities in the next table:

Table 10: Completely Standardized Error Variances and Communality

Observed Variable	Error Variance ¹	Communality ²³	P-Value	Significant Error Variance? ⁴
01	0.31	0.69	<0.001	Yes
02	0.29	0.71	<0.001	Yes
03	0.38	0.62	<0.001	Yes
04	0.43	0.57	<0.001	Yes

¹ Can be interpreted as proportion of unexplained variance by the latent factor(s) (%).

Factor Reliability

The table "Factor Reliability" contains the omega measures of factor reliability given by Bentler (Bentler, 1972, 2009) and McDonald (McDonald, 1999) and the average variance extracted (AVE). The interpretatibility of the reliability measures depend on the specific research context. In some fields of research, omega values greater or equal than 0.6 and AVE values greater or equal than 0.5 (fulfilled by and large in your case) could be sufficient.

⁴ No (correct) information available. We assume the signs of the factor loadings correspond to your expectation.

⁵ _____

⁶ _____

⁷ ____

² Corresponds to the squared factor loading.

³ Can be interpreted as proportion of explained variance by the latent factor(s) (%).

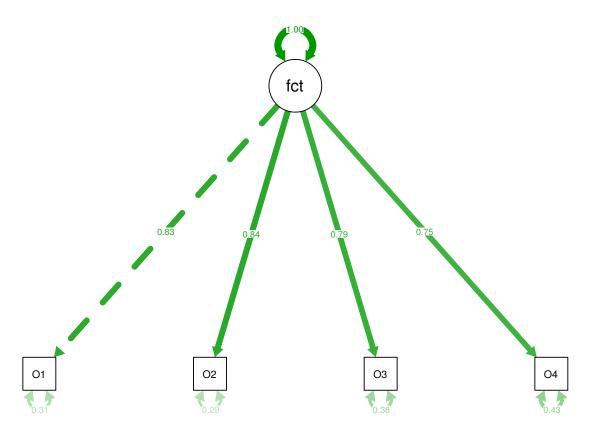
⁴ 5% type I error. Typically significant since a large portion of variance is not explained by the latent variable.

Final Summary

In our final evaluation, we distinguish between following model quality categories: acceptable, non-acceptable or uncertain.

Considering the goodness-of-fit indices, the model is acceptable. Moreover, we cannot identify localized areas of ill fit. We conclude that the model is acceptable.

Path Diagram



APPENDIX

Decision Column of the Modification Indices Table

```
not mi.significant & not high.power := "(i)"
mi.significant & not high.power := "**(m)**"
not mi.significant & high.power := "(nm)"
mi.significant & high.power & not epc.high := "epc:nm"
mi.significant & high.power & epc.high := "*epc:m*"
```

References