# GESIS Spring Seminar 23

## Comparative Social Research with Multi-Group SEM $\,$

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1

## **Table of Contents**

Packages

Data	1
Confirmatory measurement analysis (CFA)  Procedure	2 2 3
Packages	
<pre># basic tools library(tidyverse)</pre>	
<pre># checking data library(MVN) library(correlation) library(performance) library(parameters) library(insight)</pre>	
<pre># building the models library(lavaan)</pre>	
<pre># cheking the models library(semTools)</pre>	
<pre># model visualization library(semPlot) library(semptools)</pre>	

## Data

```
ESS07 <- read.csv("../Data/ESS07.csv")
```

### Confirmatory measurement analysis (CFA)

#### Procedure

Step 1: Translation of the theoretical model into a measurement model

• We need a model that is able to conceptualize the relationship between manifest indicators and latent variables

Step 2: Checking whether it is a reflexive or formative construct

- Reflective measurement: indicators are "effects" of the latent variable
- Formative measurement: latent variable is "caused" or composed by the manifest indicators

Step 3: Specification of the model (Choose a method to assign a scale to the latent variable)

- Reference indicator method: factor loading of one indicator is fixed to 1.0
- Fixed factor method: the factor variance is fixed to 1.0

Step 4: Checking the model assumptions (ML estimation requirements)

- Data are continuous and multivariate normal (case of non-normal data standard errors and chi-squre-statistics should be adjusted e.g., robust ML = MLR)
- Sample size is sufficiently large (ratio of number of cases and number of parameters N:q rule => 20:1 or 10:1)

Step 5: Checking wether the model is identified

- Complexity of a CFA models is limited by the total number of observations (p) and the number of unknown model parameters (q)
- Degrees of freedom must be at least zero: df = p q >= 0
- j = number of indicators
- p = j \* (j + 1) / 2
- q = counting of the parameters to be calculated (factor loading, residual error, factor variance)
- df = 0 = model is identified and saturated / df > 0 = model is overidentified

Step 6: Checking the fit measurements

Index	Тур	Theoretical range	Cut-off	N sensitive	Penalty for complexity
$\overline{\mathrm{X}2/\mathrm{df}}$	badness	>= 0	< 5	yes	no
CFI	godness	0.0 - 1.0	>= 0.95 (>= 0.90)	no	yes
TLI	godness	0.0 - 1.0*		no	yes
SRMR	badness	>= 0	< 0.08	yes	no
RMSEA	badness	>= 0	<=0.05 (<=0.08)	yes to small N	yes
PCLOSE	badness	0.0 - 1.0	>=0.95	yes	/

<sup>\*</sup>negative values indicate extremely misspecified model; when exceeds 1, model is extremely well-fitting

Step 7: Checking for misspecification

- What causes model misfit? => Indicator choice, Factor choice, Violations of assumptions (e.g., multivariate normality) and Causal structure (e.g., restrictions)
- How to diagnose model misfit? => Parameter estimates (Heywood cases?), Residual matrices (i.e., differences between observed and estimated covariances), Modification indices (approximation of the reduction of chi-square if a single constrained parameter is freely estimated)

Step 8: Identifying the mean structure (optional)

- Fixing the intercept of the reference indicator to zero
- Fixing the latent mean to zero

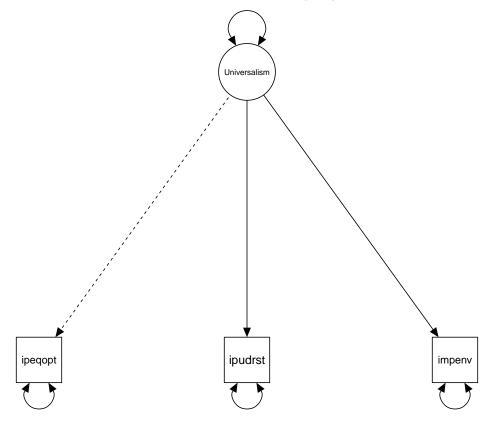
#### CFA model of Universalism

#### Step 1

?!

#### Step 2

Indicators are "effects" of the latent variable and therefore we specify a reflective measurement model

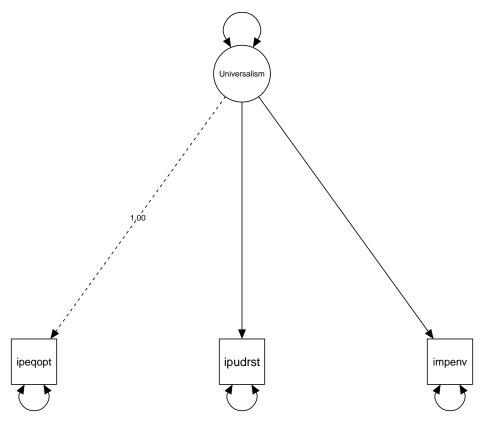


Step 3 The syntax commands for specifying a lavaan model: https://lavaan.ugent.be/tutorial/syntax1.html

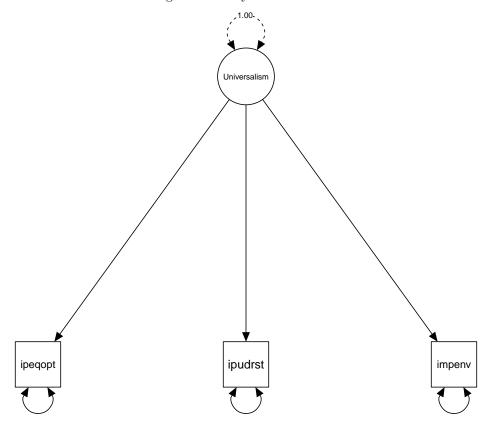
Formula type	Operator	Mnemonic
latent variable definition	=~	is measured by
regression	~	is regressed on
(residual) (co)variance	~~	is correlated with
intercept	~1	intercept

```
universalism <- "uni =~ ipeqopt + ipudrst + impenv"
```

Lavaan automatically uses the Reference indicator method, so one indicator's factual loading is fixed at 1. Accordingly, the variance of the latent factor is estimated.



In Lavaan, however, it is also very easy to use the fixed factor method, i.e. to fix the variance of the latent factor to 1. This allows all factor loadings to be freely estimated.



This requires the use of labels, i.e. the labeling of individual parameters in order to be able to control them directly. The syntax commands for labeling a lavaan model can be found here: https://lavaan.ugent.be/tutorial/syntax2.html. We just need to understand how parameters are fixed in Lavaan and how they can be calculated freely.

Reference indicator method:

```
universalism <- "uni =~ 1*ipeqopt + ipudrst + impenv"
```

Fixed factor method:

```
universalism <- "
# specification of the measurement model
uni =~ NA*ipeqopt + ipudrst + impenv

# specification of the variance
uni ~~ 1*uni
"</pre>
```

Freely estimate the factor loading of the first indicator by NA\*ipeqopt and fix the variance of the latent factor by 1\*uni. The variance of a variable is specified as covariance with itself, hence uni  $\sim$  uni. In the end we get uni  $\sim$  1\*uni

#### Step 4

Now let's get an overview of the properties of our variables. First, let's look at the descriptive statistics of our items to gain knowledge of missing values, central tendency properties, variability, and distribution.

```
ESS07 %>%
  select(ipeqopt, ipudrst, impenv) %>%
  mvn() %>%
  # optional for better display
  export_table(table_width = 1, digits = 3)
```

Test	 	Variable	 	Statistic	 	p value	 	Normality
Anderson-Darling Anderson-Darling		ipeqopt	   	362.6880 332.9049	•	<0.001 <0.001	   	NO NO
Anderson-Darling		impenv	İ	322.7889	•		i	NO

n	1	Mean	I	Std.Dev		Median		Min	I	Max	I	25th	75th	I	Skew	I	Kurtosis
5844		4.945	1	1.040		 5		1		6	1	 4	 6	 	-1.142	 	1.295
5844	1	4.717	1	1.027	1	5		1		6	1	4	5	-	-0.937		0.892
5844	1	4.878	1	1.025	Τ	5	Ι	1	Ι	6	Ι	4	6	- 1	-0.933	I	0.732