

# CFA\_SimDesign

```
library(SimDesign)
library(lavaan)
library(tidyr)
```

## Step 1 — Defining conditions (sample size and model)

```
Design <- createDesign(sample_size = c(50, 100, 500, 1e3, 1e4, 1e5),
                        model_type = c("true", "misspecified"))
```

## Step 2 — Define generate, analyse, and summarise functions

```
# Generate function: Simulate data based on the true model
Generate <- function(condition, fixed_objects) {
  N <- condition$sample_size

  # True model: Two factors, 8 indicators each
  true_model <- '
    f1 =~ 0.7*y1 +
    0.7*y2 +
    0.7*y3 +
    0.7*y4 +
    0.7*y5 +
    0.7*y6 +
    0.7*y7 +
    0.7*y8

    f2 =~ 0.7*y9 +
```

```

    0.7*y10 +
    0.7*y11 +
    0.7*y12 +
    0.7*y13 +
    0.7*y14 +
    0.7*y15 +
    0.7*y16
  ,

# Misspecification: Add cross-loadings
misspecified_model <- '
  f1 =~ 0.7*y1 +
  0.7*y2 +
  0.7*y3 +
  0.7*y4 +
  0.7*y5 +
  0.7*y6 +
  0.7*y7 +
  0.7*y8 +
  0.3*y9 # added cross-loading

  f2 =~ 0.7*y9 +
  0.7*y10 +
  0.7*y11 +
  0.7*y12 +
  0.7*y13 +
  0.7*y14 +
  0.7*y15 +
  0.7*y16 +
  0.3*y1 # added cross-loading
,

# Generate data based on the condition
if(condition$model_type == "true") {
  data <- simulateData(true_model, sample.nobs = N)
} else {
  data <- simulateData(misspecified_model, sample.nobs = N)
}

return(data)

```

```

}

# Analyse function: Fit the true model and extract fit measures
Analyse <- function(condition, dat, fixed_objects) {
  true_model <- '
    f1 =~ y1 + y2 + y3 + y4 + y5 + y6 + y7 + y8
    f2 =~ y9 + y10 + y11 + y12 + y13 + y14 + y15 + y16
  '

  fit <- cfa(true_model, data = dat)

  # Extract fit measures
  fit_measures <- fitMeasures(fit, c("cfi", "tli", "rmsea", "srmr", "chisq"))
  return(fit_measures)
}

# Summarise function: Calculate the average fit measures from n replications
Summarise <- function(condition, results, fixed_objects) {
  summary_measures <- colMeans(results)
  return(summary_measures)
}

# Make sure it doesn't try resuming a simulation
SimClean()

# Step 3 --- Collect results by looping over the rows in design
library(SimDesign)
library(lavaan)

res <- runSimulation(design=Design, replications=100,
  generate=Generate, analyse=Analyse, summarise=Summarise)

```

Design: 1/12; RAM Used: 62.3 Mb; Replications: 100; Total Time: 0.00s  
 Conditions: sample\_size=50, model\_type=true

Design: 2/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 4.09s

Conditions: sample\_size=100, model\_type=true

Design: 3/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 8.09s  
Conditions: sample\_size=500, model\_type=true

Design: 4/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 12.08s  
Conditions: sample\_size=1000, model\_type=true

Design: 5/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 16.38s  
Conditions: sample\_size=10000, model\_type=true

Design: 6/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 22.12s  
Conditions: sample\_size=1e+05, model\_type=true

Design: 7/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 45.44s  
Conditions: sample\_size=50, model\_type=misspecified

Design: 8/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 52.94s  
Conditions: sample\_size=100, model\_type=misspecified

Design: 9/12; RAM Used: 78.8 Mb; Replications: 100; Total Time: 01m 0.23s  
Conditions: sample\_size=500, model\_type=misspecified

Design: 10/12; RAM Used: 78.9 Mb; Replications: 100; Total Time: 01m 7.60s  
Conditions: sample\_size=1000, model\_type=misspecified

Design: 11/12; RAM Used: 78.9 Mb; Replications: 100; Total Time: 01m 15.24s  
Conditions: sample\_size=10000, model\_type=misspecified

Design: 12/12; RAM Used: 78.9 Mb; Replications: 100; Total Time: 01m 20.64s  
Conditions: sample\_size=1e+05, model\_type=misspecified

Simulation complete. Total execution time: 01m 51.22s

```
res <- res %>% dplyr::select(-COMPLETED, -SEED)
```

```
# -----  
# -----  
print(res)
```

# A tibble: 12 x 10

	sample_size	model_type	cfi	tli	rmsea	srmr	chisq
	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	50	true	0.89669	0.88426	0.055521	0.10270	123.07
2	100	true	0.96744	0.97400	0.025384	0.072316	111.04
3	500	true	0.99557	0.99826	0.0085409	0.032497	105.55
4	1000	true	0.99833	0.99997	0.0045251	0.022673	103.05
5	10000	true	0.99976	0.99989	0.0018654	0.0073577	106.12
6	100000	true	0.99998	1.0000	0.00052441	0.0023223	104.07
7	50	misspecified	0.86934	0.85193	0.063645	0.10913	127.54
8	100	misspecified	0.94071	0.93262	0.040156	0.083298	122.90
9	500	misspecified	0.96078	0.95430	0.034933	0.054521	168.11
10	1000	misspecified	0.96418	0.95827	0.033450	0.048682	219.91
11	10000	misspecified	0.96309	0.95700	0.034296	0.044508	1315.6
12	100000	misspecified	0.96344	0.95741	0.034130	0.043800	12102.

# i 3 more variables: REPLICATIONS <dbl>, SIM\_TIME <chr>, RAM\_USED <chr>

```
true_model <- '  
  f1 =~ y1 + y2 + y3 + y4 + y5 + y6 + y7 + y8  
  f2 =~ y9 + y10 + y11 + y12 + y13 + y14 + y15 + y16  
,  
  
simulated_data <- simulateData(true_model, sample.nobs = 1000)  
  
fit <- cfa(true_model, data = simulated_data)  
  
summary(fit, fit.measures = TRUE)
```

lavaan 0.6-18 ended normally after 26 iterations

Estimator	ML
Optimization method	NLMINB

Number of model parameters	33
Number of observations	1000

Model Test User Model:

Test statistic	100.847
Degrees of freedom	103
P-value (Chi-square)	0.542

Model Test Baseline Model:

Test statistic	6951.272
Degrees of freedom	120
P-value	0.000

User Model versus Baseline Model:

Comparative Fit Index (CFI)	1.000
Tucker-Lewis Index (TLI)	1.000

Loglikelihood and Information Criteria:

Loglikelihood user model (H0)	-24950.843
Loglikelihood unrestricted model (H1)	-24900.419
Akaike (AIC)	49967.685
Bayesian (BIC)	50129.641
Sample-size adjusted Bayesian (SABIC)	50024.831

Root Mean Square Error of Approximation:

RMSEA	0.000
90 Percent confidence interval - lower	0.000
90 Percent confidence interval - upper	0.016
P-value H <sub>0</sub> : RMSEA ≤ 0.050	1.000
P-value H <sub>0</sub> : RMSEA ≥ 0.080	0.000

Standardized Root Mean Square Residual:

SRMR	0.020
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Parameter Estimates:

Standard errors	Standard
Information	Expected
Information saturated (h1) model	Structured

Latent Variables:

	Estimate	Std.Err	z-value	P(> z )
f1 =~				
y1	1.000			
y2	0.896	0.042	21.329	0.000
y3	0.975	0.043	22.443	0.000
y4	0.986	0.044	22.449	0.000
y5	0.989	0.044	22.638	0.000
y6	0.941	0.043	21.733	0.000
y7	0.912	0.043	21.339	0.000
y8	0.911	0.043	21.427	0.000
f2 =~				
y9	1.000			
y10	1.015	0.050	20.281	0.000
y11	1.003	0.050	20.141	0.000
y12	1.059	0.051	20.604	0.000
y13	1.045	0.050	20.947	0.000
y14	1.066	0.051	20.812	0.000
y15	1.013	0.049	20.775	0.000
y16	1.062	0.049	21.649	0.000

Covariances:

	Estimate	Std.Err	z-value	P(> z )
f1 ~~				
f2	-0.014	0.037	-0.379	0.705

Variances:

	Estimate	Std.Err	z-value	P(> z )
.y1	0.909	0.048	18.944	0.000
.y2	1.010	0.051	19.898	0.000
.y3	0.986	0.051	19.371	0.000
.y4	1.008	0.052	19.368	0.000
.y5	0.982	0.051	19.265	0.000
.y6	1.040	0.053	19.721	0.000
.y7	1.044	0.052	19.893	0.000
.y8	1.027	0.052	19.856	0.000
.y9	0.966	0.049	19.615	0.000
.y10	1.066	0.054	19.800	0.000

.y11	1.070	0.054	19.870	0.000
.y12	1.088	0.055	19.626	0.000
.y13	0.987	0.051	19.425	0.000
.y14	1.057	0.054	19.507	0.000
.y15	0.961	0.049	19.528	0.000
.y16	0.878	0.046	18.947	0.000
f1	1.145	0.086	13.277	0.000
f2	0.956	0.078	12.257	0.000