

EFA vs. CFA revisited

FACTOR ANALYSIS IN R



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Review of the differences between EFA & CFA

EFA:

- Estimates all possible variable/factor relationships
- Looking for patterns in the data
- Use when you don't have a well-developed theory

CFA:

- Only specified variable/factor relationships
- Testing a theory that you know in advance
- This is the right thing to publish!

```
# View the first five rows of the EFA loadings
```

```
EFA_model$loadings[1:5,]
```

```
      MR2      MR1      MR3      MR5      MR4
A1  0.24282081 -0.15387946  0.0780303740 -0.3897470 -0.08461786
A2 -0.02320759  0.01798410  0.0679900414  0.6584172 -0.01095097
A3 -0.05917275 -0.12693134  0.0238309309  0.6154942  0.05036830
A4 -0.03852599 -0.08709392  0.1936172346  0.4005924 -0.17361760
A5 -0.13355262 -0.23429925  0.0001429341  0.5075002  0.07523716
```

```
# View the first five loadings from the CFA estimated from the EFA results
```

```
summary(EFA_CFA)$coeff[1:5,]
```

```
      Estimate Std Error  z value      Pr(>|z|)
F4A1 -0.5038817  0.04497739 -11.20300  3.941591e-29 A1 <--- MR5
F4A2  0.8207622  0.03465055  23.68684  4.927422e-124 A2 <--- MR5
F4A3  1.0360812  0.03700471  27.99863  1.688392e-172 A3 <--- MR5
F4A4  0.8264718  0.04471746  18.48208  2.878650e-76  A4 <--- MR5
F4A5  0.9012645  0.03688629  24.43359  7.520155e-132 A5 <--- MR5
```

Comparing factor loadings

```
# View the first five loadings from the CFA estimated from the EFA results  
summary(EFA_CFA)$coeff[1:5, ]
```

```
      Estimate Std Error  z value      Pr(>|z|)  
F4A1 -0.5038817 0.04497739 -11.20300 3.941591e-29 A1 <--- MR5
```

```
# View the first five rows of the EFA loadings  
EFA_model$loadings[1:5, ]
```

```
      MR2      MR1      MR3      MR5      MR4  
A1      -0.3897470
```

Differences in factor scores

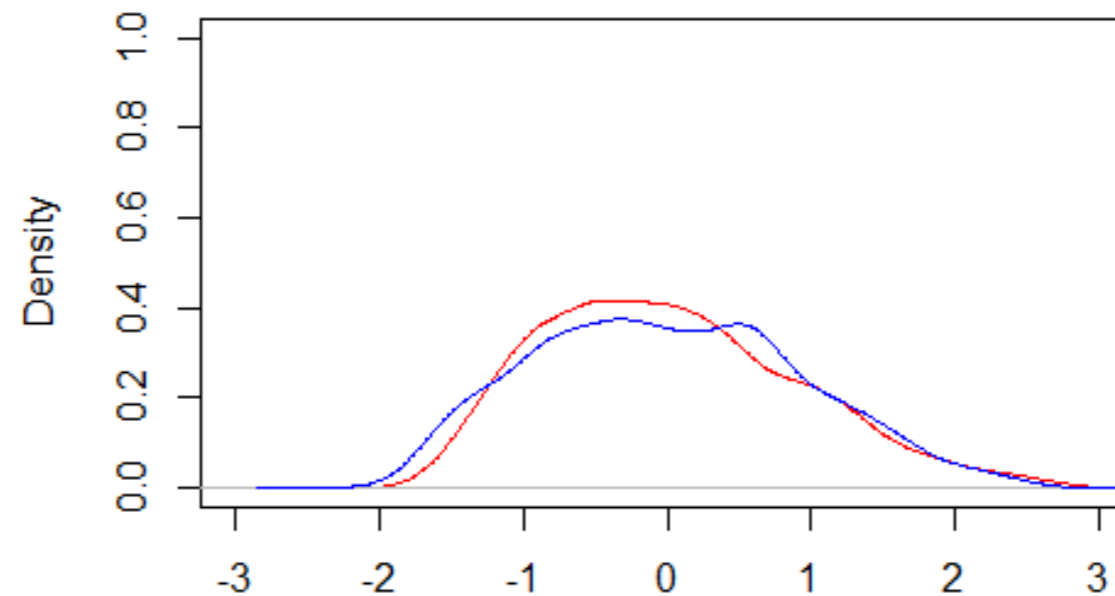
```
# Extracting factor scores from the EFA model  
EFA_scores <- EFA_model$scores
```

```
# Calculate factor scores for the EFA dataset  
CFA_scores <- fscores(EFA_CFA, data = bfi_EFA)
```

Differences in factor scores, visualized

```
plot(density(EFA_scores[,1], na.rm = TRUE),  
     xlim = c(-3, 3), ylim = c(0, 1), col = "blue")  
lines(density(CFA_scores[,1], na.rm = TRUE),  
      xlim = c(-3, 3), ylim = c(0, 1), col = "red")
```

`density.default(x = CFA_scores[, 1], na.rm = TRUE)`



Let's practice!

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Adding loadings to improve model fit

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When to make adjustments

Remember:

- EFAs estimate all item/factor loadings
- CFAs only estimate specified loadings
- Poor model fit could be due to excluded loadings

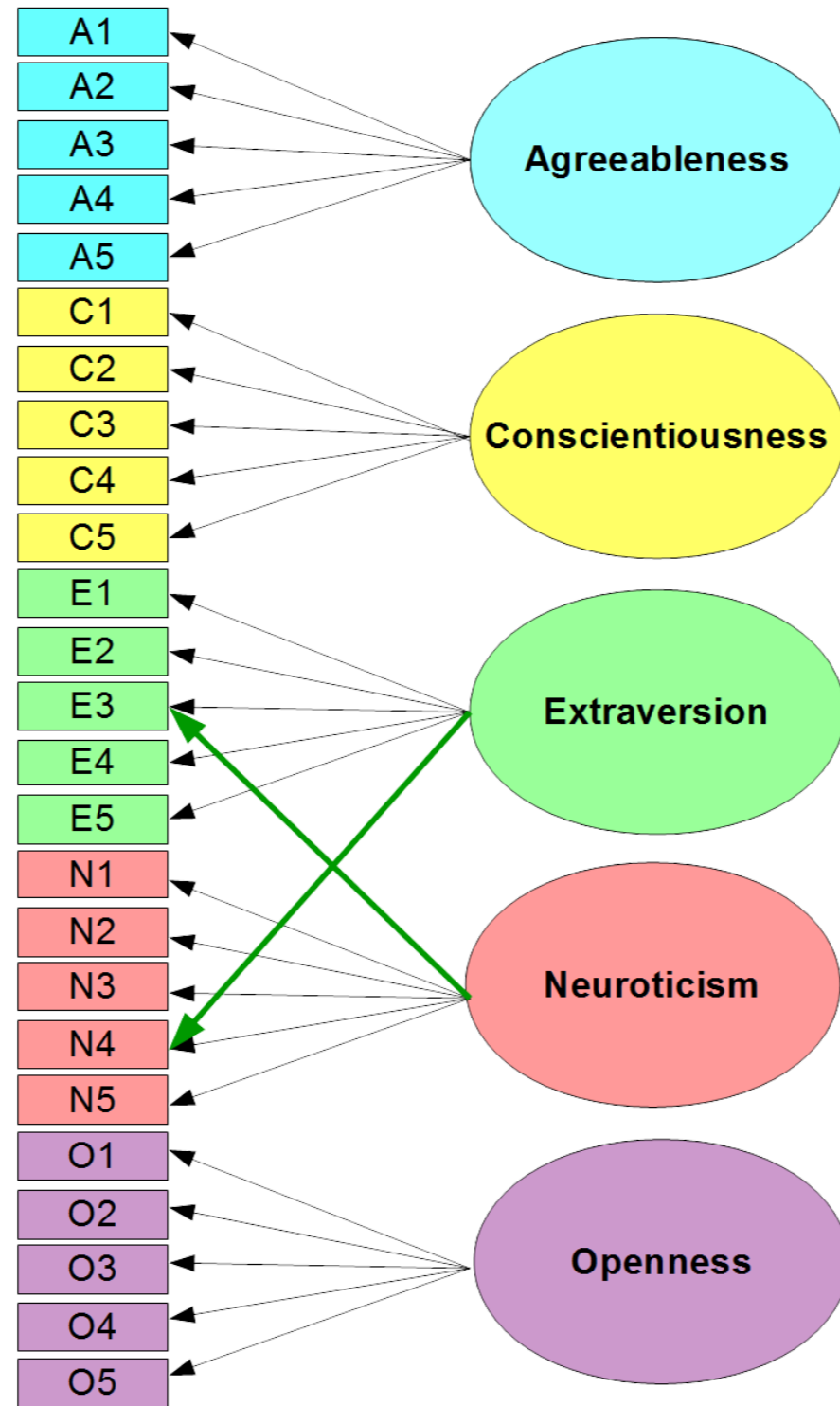
Adding loadings to the syntax

Two promising item/factor relationships to add:

- Extraversion ? Item N4
- Neuroticism ? Item E3

```
summary(theory_CFA)
```

```
...  
  
Parameter Estimates  
      Estimate  Std Error  z value  Pr(>|z|)  
C[EXT,NEU]  0.2362614  0.03364096  7.023029  2.171093e-12 NEU <--> EXT  
  
...
```



Adding new loadings to the syntax

```
# Add some plausible item/factor loadings to the syntax
theory_syn_add <- "
AGE: A1, A2, A3, A4, A5
CON: C1, C2, C3, C4, C5
EXT: E1, E2, E3, E4, E5, N4
NEU: N1, N2, N3, N4, N5, E3
OPE: O1, O2, O3, O4, O5 "
# As before, convert your equations to sem-compatible syntax
theory_syn2 <- cfa(text = theory_syn_add, reference.indicators = FALSE)
# Run a CFA with the revised syntax
theory_CFA_add <- sem(model = theory_syn2, data = bfi_CFA)
```

Comparing the original and revised models

```
# Conduct a likelihood ratio test  
anova(theory_CFA, theory_CFA_add)
```

```
LR Test for Difference Between Models
```

	Model	Df	Model	Chisq	Df	LR	Chisq	Pr(>Chisq)
	theory_CFA	265		2212.0				
	theory_CFA_rev	263		2097.8	2	114.28	< 2.2e-16	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Comparing the original and revised models

```
# Compare the comparative fit indices - higher is better!  
summary(theory_CFA)$CFI
```

```
0.785075
```

```
summary(theory_CFA_add)$CFI
```

```
0.7974694
```

Comparing the original and revised models

```
# Compare the RMSEA values - lower is better!
```

```
summary(theory_CFA)$RMSEA
```

```
0.07731925      NA      NA 0.90
```

```
summary(theory_CFA_add)$RMSEA
```

```
0.07534156      NA      NA 0.90
```

- More information about fit indices: <http://davidakenny.net/cm/fit.htm>

Let's practice!

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Removing loadings to improve fit

FACTOR ANALYSIS IN R

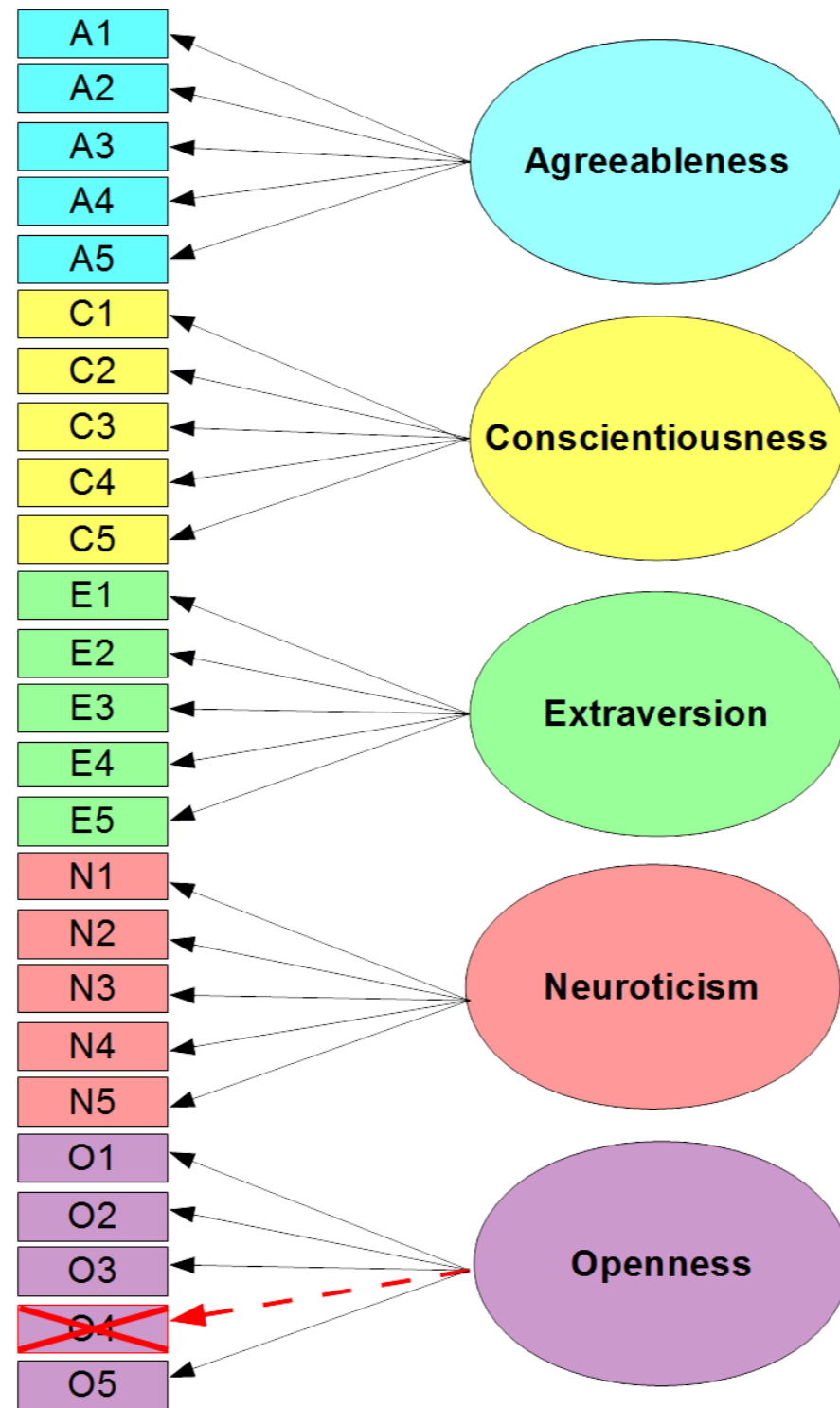


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What does it mean to remove a loading?

```
summary(theory_CFA)
```

```
Parameter Estimates
      Estimate Std Error z value Pr(>|z|)
lam[A1:AGE] -0.501  0.0449  -11.17  5.79e-29 A1 <--- AGE
...
lam[01:OPE]  0.636  0.0379   16.79  2.88e-63 01 <--- OPE
lam[02:OPE] -0.731  0.0532  -13.75  5.39e-43 02 <--- OPE
lam[03:OPE]  0.809  0.0399   20.25  3.24e-91 03 <--- OPE
lam[04:OPE]  0.287  0.0413    6.95  3.69e-12 04 <--- OPE
lam[05:OPE] -0.624  0.0444  -14.06  7.07e-45 05 <--- OPE
```



Removing a loading in the syntax

```
# The original syntax
theory_syn_eq <- "
AGE: A1, A2, A3, A4, A5
CON: C1, C2, C3, C4, C5
EXT: E1, E2, E3, E4, E5
NEU: N1, N2, N3, N4, N5
OPE: O1, O2, O3, O4, O5 "

# Remove the worst item/factor loadings from the syntax
theory_syn_del <- "
AGE: A1, A2, A3, A4, A5
CON: C1, C2, C3, C4, C5
EXT: E1, E2, E3, E4, E5
NEU: N1, N2, N3, N4, N5
OPE: O1, O2, O3, O5 "

# As before, convert your equations to sem-compatible syntax
theory_syn3 <- cfa(text = theory_syn_del, reference.indicators = FALSE)
```

Running the revised CFA

```
# Run a CFA with the revised syntax
theory_CFA_de1 <- sem(model = theory_syn3, data = bfi_CFA)
```

Warning messages:

```
1: In sem.semmod(model = theory_syn3, data = bfi_CFA) :
  -170 observations removed due to missingness
2: In sem.semmod(model = theory_syn3, data = bfi_CFA) :
  The following observed variables are in the input covariance
  or raw-moment matrix but do not appear in the model:
```

04

Comparing the original and revised models

```
anova(theory_CFA, theory_CFA_de1)
```

```
Error in anova.objectiveML(theory_CFA, theory_CFA_de1) :  
  the models are fit to different moment matrices
```

```
# Compare the comparative fit indices - higher is better!  
summary(theory_CFA)$CFI
```

```
0.785075
```

```
summary(theory_CFA_de1)$CFI
```

```
0.7995587
```

Comparing the original and revised models

```
# Compare the RMSEA values - lower is better!  
summary(theory_CFA)$RMSEA
```

```
0.07731925      NA      NA 0.900000000
```

```
summary(theory_CFA_de1)$RMSEA
```

```
0.07718057      NA      NA 0.900000000
```

More information about fit indices: <http://davidakenny.net/cm/fit.htm>

Let's practice!

FACTOR ANALYSIS IN R

Now you can conduct and interpret EFAs and CFAs!

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Things you can do

- Conduct a unidimensional EFA
- Conduct a multidimensional EFA
- Conduct a CFA based on EFA results
- Conduct a CFA based on theory
- Interpret fit statistics
- Compare and refine models

More information

- `psych` and `sem` package documentation
- Books on multivariate analysis
- [Applied Multivariate Statistical Analysis by Johnson & Wichern](#)
- [Structural Equation Modeling with lavaan in R](#)

Congrats!
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