

# Confirmatory Factor Analysis Part One<sup>1</sup>

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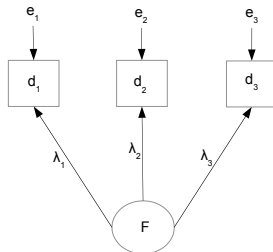
# A confirmatory factor analysis model

## One Factor: Starting simply

$$d_1 = \lambda_1 F + e_1$$

$$d_2 = \lambda_2 F + e_2$$

$$d_3 = \lambda_3 F + e_3$$



- $Var(F) = 1$
- $Var(e_j) = \omega_j$
- $F, e_1, e_2, e_3$  all independent.

# Calculate $\Sigma$

$$\begin{array}{rcl} d_1 & = & \lambda_1 F + e_1 \\ d_2 & = & \lambda_2 F + e_2 \\ d_3 & = & \lambda_3 F + e_3 \end{array} \quad \Sigma = \begin{array}{c|ccc} & d_1 & d_2 & d_3 \\ \hline d_1 & \lambda_1^2 + \omega_1 & \lambda_1 \lambda_2 & \lambda_1 \lambda_3 \\ d_2 & & \lambda_2^2 + \omega_2 & \lambda_2 \lambda_3 \\ d_3 & & & \lambda_3^2 + \omega_3 \end{array}$$

Are the parameters identifiable? What if just one  $\lambda$  is zero?

Suppose no factor loadings equal zero

$$\Sigma = \begin{pmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ & \sigma_{22} & \sigma_{23} \\ & & \sigma_{33} \end{pmatrix} = \begin{pmatrix} \lambda_1^2 + \omega_1 & \lambda_1 \lambda_2 & \lambda_1 \lambda_3 \\ & \lambda_2^2 + \omega_2 & \lambda_2 \lambda_3 \\ & & \lambda_3^2 + \omega_3 \end{pmatrix}$$

$$\lambda_1^2 = \frac{\sigma_{12}\sigma_{13}}{\sigma_{23}} = \frac{\lambda_1 \lambda_2 \lambda_1 \lambda_3}{\lambda_2 \lambda_3}$$

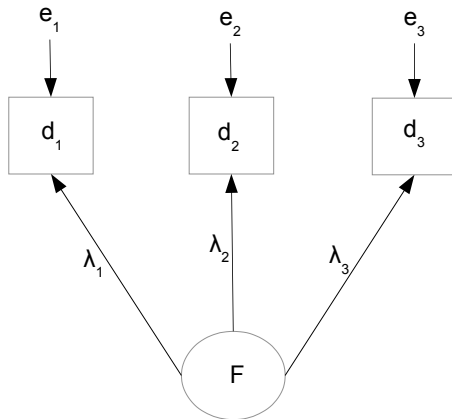
$$\lambda_2^2 = \frac{\sigma_{12}\sigma_{23}}{\sigma_{13}}$$

$$\lambda_3^2 = \frac{\sigma_{13}\sigma_{23}}{\sigma_{12}}$$

- Squared factor loadings are identifiable, but not the loadings.
- Replace all  $\lambda_j$  with  $-\lambda_j$ , get same  $\Sigma$
- Likelihood function will have two maxima, same height.
- Which one you find depends on where you start.

# Solution: Decide on the sign of one loading

Based on *meaning*



- Is  $F$  math ability or math *inability*? You decide.
- It's just a matter of naming the factors.

If  $\lambda_1 > 0$

$$\Sigma = \begin{pmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ & \sigma_{22} & \sigma_{23} \\ & & \sigma_{33} \end{pmatrix} = \begin{pmatrix} \lambda_1^2 + \omega_1 & \lambda_1 \lambda_2 & \lambda_1 \lambda_3 \\ & \lambda_2^2 + \omega_2 & \lambda_2 \lambda_3 \\ & & \lambda_3^2 + \omega_3 \end{pmatrix}$$

- Signs of  $\lambda_2$  and  $\lambda_3$  can be recovered right away from  $\Sigma$ .
- And all the parameters are identified.

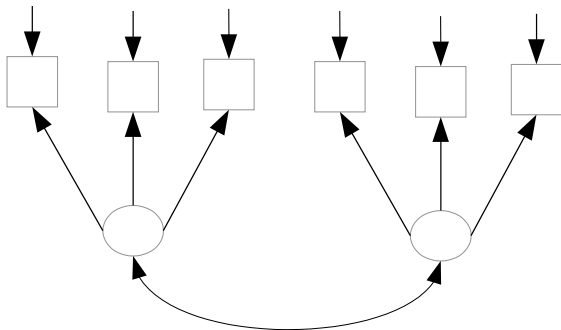
Add another variable:  $d_4 = \lambda_4 F + e_4$

$$\Sigma = \begin{pmatrix} \lambda_1^2 + \omega_1 & \lambda_1 \lambda_2 & \lambda_1 \lambda_3 & \lambda_1 \lambda_4 \\ & \lambda_2^2 + \omega_2 & \lambda_2 \lambda_3 & \lambda_2 \lambda_4 \\ & & \lambda_3^2 + \omega_3 & \lambda_3 \lambda_4 \\ & & & \lambda_4^2 + \omega_4 \end{pmatrix}$$

- Parameters will all be identifiable as long as 3 out of 4 loadings are non-zero, and one sign is known.
- For example, if only  $\lambda_1 = 0$  then the top row = 0, and you can get  $\lambda_2, \lambda_3, \lambda_4$  as before.
- For 5 variables, two loadings can be zero, etc.
- How many equality restrictions?  $6 - 4 = 2$ .

Now add another factor

$$\text{Var}(F_1) = \text{Var}(F_2) = 1$$



$$d_1 = \lambda_1 F_1 + e_1$$

$$\vdots$$

$$d_6 = \lambda_6 F_2 + e_6$$



## Covariance matrix of observable variables

$$\Sigma = \begin{pmatrix} \lambda_1^2 + \omega_1 & \lambda_1\lambda_2 & \lambda_1\lambda_3 & \lambda_1\lambda_4\phi_{12} & \lambda_1\lambda_5\phi_{12} & \lambda_1\lambda_6\phi_{12} \\ & \lambda_2^2 + \omega_2 & \lambda_2\lambda_3 & \lambda_2\lambda_4\phi_{12} & \lambda_2\lambda_5\phi_{12} & \lambda_2\lambda_6\phi_{12} \\ & & \lambda_3^2 + \omega_3 & \lambda_3\lambda_4\phi_{12} & \lambda_3\lambda_5\phi_{12} & \lambda_3\lambda_6\phi_{12} \\ & & & \lambda_4^2 + \omega_4 & \lambda_4\lambda_5 & \lambda_4\lambda_6 \\ & & & & \lambda_5^2 + \omega_5 & \lambda_5\lambda_6 \\ & & & & & \lambda_6^2 + \omega_6 \end{pmatrix}$$

- Identify  $\lambda_1, \lambda_2, \lambda_3$  from set One (assuming one sign is known).
- Identify  $\lambda_4, \lambda_5, \lambda_6$  from set Two (lower right).
- Identify  $\phi_{12}$  from any unused covariance.
- What if you added more variables?
- What if you added more factors?

# Three-variable identification rule

For standardized factors

For a factor analysis model, the parameters will be identifiable provided

- Errors are independent of one another and of the factors.
- Variances of all factors equal one.
- Each observed variable is a function of only one factor.
- There are at least three observable variables with non-zero loadings per factor.
- The sign of one non-zero loading is known for each factor.

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<http://www.utstat.toronto.edu/brunner/oldclass/2053f22>