

Lab 1 - Correlations, reliability and consistency

Tony Tan & Jarl Kristensen

University of Oslo

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Today

Estimate and interpret:

- Correlations between variables
- Test-retest reliability
- Alternate/parallel test-forms
- Internal consistency
 - Cronbach's alpha
 - Single factor model
 - McDonald's omega

Correlation

$$\rho_{XY} = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

- Correlation is a measure of the strength of the **linear relationship** between two variables.
- The correlation coefficient is a number between -1 and 1 , where 0 indicates no linear relationship, and 1 or -1 indicates perfect a linear relationship.
- Correlations are superior to **covariances** for interpreting interrelationship between two variables thanks to the standardisation procedure.

Task 1: Correlation

Estimate the correlation between a pair of variables in `likert_data.rds` and comment on the relationship between the variables you chose.

Test reliability

- Test-retest reliability
- Alternate test forms

Use the correlation of the **sum scores** of the two test forms to estimate the reliability of the test.

Cronbach's α

$$\alpha = \frac{m}{m-1} \left[1 - \frac{\sum_{j=1}^m \text{Var}(X_j)}{\sigma_Y^2} \right]$$


Task 2: Test reliability

- Estimate coefficient alpha for one of the scales in the `dich_data.rds` dataset.
- How would you describe the estimated value?

Cronbach's α : Pitfalls

- Dependent on the number of items
- Assumes a **single factor model** with **equal loadings** for all items
- A **lower bound** estimate of reliability as long as assumptions are met

Single factor model

- lavaan and other  packages
- This lab focuses on lavaan
- Lecture 6: single factor model

lavaan: Syntax

Operator	Reading	Meaning
$=\sim$	is measured by	define a latent variable
\sim	is regressed on	define a regression model
$\sim\sim$	is correlated with	specify covariances

lavaan: Example code

```
# Load lavaan  
library(lavaan)  
  
# Define latent variable  
lat_var = " y =~ x1 + x2 + x3 "  
  
# Run a confirmatory factor analysis  
cfa <- cfa(lat_var, data = mydata)  
  
# Model evaluation  
summary(cfa, fit.measures = TRUE)  
  
# Extract model coefficients  
coef(cfa)
```

Task 3: Single factor model

- Estimate the single factor model using the scale you chose in Task 2
- Evaluate the model fit.
- Evaluate if the α you calculated in Task 2 violates the assumptions α relies on.

Coefficient omega

$$\omega = \frac{\sigma_C^2}{\sigma_C^2 + \sigma_U^2} = \frac{\sigma_Y^2}{\sigma_{T_Y}^2 + \sigma_{E_Y}^2} = \frac{\left(\sum_{j=1}^m \lambda_j\right)^2}{\left(\sum_{j=1}^m \lambda_j\right)^2 + \sum_{j=1}^m \psi_i^2}$$

Coefficient omega is the ratio of the true score variance (common) to total score variance (common + unique).

Task 4: Standardised model

- Use the coefficients from Task 3 to calculate ω .
- Evaluate the reliability of the scale.
- Compare ω (Task 4) and α (Task 3).