

ANOVA tests for multiple d-primes

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Outline

- 1 One-way anova for d-primes
- 2 Estimating and testing the common d-prime
- 3 Post-hoc tests and comparisons

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One-way ANOVA

A one-way anova model has the form

$$y_{ij} = \alpha_i + \varepsilon_{ij}$$

Tests of interest:

- Any difference between α_i ?
- Pairwise differences between α_i
- Comparing baseline α to the rest.

One-way ANOVA for d-primes

Conceptually:

$$y_{ij} = d'_i + \text{binomial-deviations}_{ij}$$

Formally:

$$y_i \sim \text{binom}(f_{m_i}(d'_i), n_i)$$

where

- m_i is the *method* (2-AFC, Triangle etc.)
- $f_{m_i}(d'_i)$ is the psychometric function for the i th method.
- y_i is the no. correct trials for the i th experiment.
- n_i is the total no. trials for the i th experiment.
- d'_i is d' for the i th experiment.

The any-differences hypothesis

$$H_0 : d'_1 = d'_2 = \dots = d'_n \quad \text{versus} \quad H_A : d'_i \neq d'_{i'}$$

for at least one pair of (i, i') .

Examples in R

The common d-prime

$$d'_i = d'_c + e'_i$$

where

- d'_i is the d' from the i th experiment
- d'_c is the *common* d'
- e'_i are deviations from the *common* d'

Challenge:

- Binomial data: $(y_i, n_i) \rightarrow d'_i$
- What if n_i is large for some experiments, but small for others?
- What if different protocols are used?

Solution: estimate d'_c with Maximum Likelihood.

Estimation of common d'

Simple solution — a weighted average:

$$d'_{wa,e} = \sum_i w'_i d'_i$$

where $w'_i = w_i / \sum_i w_i$ are normalized weights.

Better solution: The ML estimator of d'_c :

$$\hat{d}'_c = \arg \max_{d'_c} \ell_0(d'_c; \mathbf{x}, \mathbf{n}, \mathbf{m}).$$

- $\mathbf{x}, \mathbf{n}, \mathbf{m}$: data from all experiments.

Testing the common d'

$$H_0 : d'_c = d'_0 \quad \text{versus} \quad H_A : d'_c \neq d'_0$$

- where d'_0 is d' under the H_0 .

For difference testing we want:

$$H_0 : d'_c = 0 \quad \text{versus} \quad H_A : d'_c > 0$$

For similarity testing we want:

$$H_0 : d'_c \geq d'_0 \quad \text{versus} \quad H_A : d'_c < d'_0$$

Examples in R

All pairwise differences

Hypothesis test:

$$H_0 : d'_i - d'_{i'} = 0$$

$$H_A : d'_i - d'_{i'} \neq 0$$

for some pair (i, i') where $i, i' = 1, \dots, n$.

A *compact letter display* summarizes the pairwise tests:

- 1 A letter is assigned to all $i = 1, \dots, n$ groups.
- 2 Two groups sharing a letter are *not* significantly different.
- 3 Two groups *not* sharing a letter are significantly different.

p -values are usually adjusted for multiplicity.

Difference from common d'

$$H_0 : d'_i = d'_c \text{ for all } i$$

$$H_A : d'_i = d'_{c(i')}, \text{ for all } i \text{ except } i'$$

$d'_{c(i')}$ is the common d' considering all i except i'

Difference from specified value

$$H_0 : d'_i = d'_0 \quad \text{versus} \quad H_A : d'_i \neq d'_0$$

where d'_0 is the value of d' under H_0 .

Examples in R