

# Within Cases

## The Humble $t$ -test

# Overview

- 1 The Issue
- 2 Univariate
- 3 Multivariate

# Independent Observations

- Most statistical models assume independent observations.
- Sometimes the assumption of independence is unreasonable.
- For example, times series and within cases designs.

## Within Cases

- A case contributes a value of the response variable for every value of a categorical explanatory variable.
- As opposed to explanatory variables that are *Between Cases*: Explanatory variables partition the sample.
- It is natural to expect data from the same case to be correlated, *not* independent.
- For example, the same subject appears in several treatment conditions
- Hearing study: How does pitch affect our ability to hear faint sounds? Subjects are presented with tones at a variety of different pitch and volume levels (in a random order). They press a key when they think they hear something.
- A study can have both within and between cases factors.

## You may hear terms like

- **Longitudinal:** The same variables are measured repeatedly over time. Usually lots of variables, including categorical ones, and large samples. If there's an experimental treatment, its usually once at the beginning, like a surgery. Basically its *tracking* what happens over time.
- **Repeated measures:** Usually, same subjects experience two or more experimental treatments. Usually quantitative explanatory variables and small samples.

# Student's Sleep Study (*Biometrika*, 1908)

First Published Example of a  $t$ -test

- Patients take two sleeping medicines several days apart.
- Half get  $A$  first, half get  $B$  first.
- Reported hours of sleep are recorded.
- It's natural to subtract, and test whether the mean *difference* equals zero.
- That's what Gossett did.
- But some might do an independent  $t$ -test with  $n_1 = n_2$ .
- Is it harmful?

## Conclusions from an earlier discussion

- When covariance is positive, matched  $t$ -test has better power
- Each case serves as its own control.
- A huge number of unknown influences are removed by subtraction.
- This makes the analysis more precise.

# Hotelling's $t^2$

## Multivariate Matched $t$ -test

- $\mathbf{X}_1, \dots, \mathbf{X}_n \stackrel{i.i.d.}{\sim} N_k(\boldsymbol{\mu}, \boldsymbol{\Sigma})$
- $\bar{\mathbf{X}}_n = \frac{1}{n} \sum_{i=1}^n \mathbf{X}_i$  and  $\mathbf{S} = \frac{1}{n-1} \sum_{i=1}^n (\mathbf{X}_i - \bar{\mathbf{X}}_n) (\mathbf{X}_i - \bar{\mathbf{X}}_n)'$
- $t^2 = n (\bar{\mathbf{X}}_n - \boldsymbol{\mu})' \mathbf{S}^{-1} (\bar{\mathbf{X}}_n - \boldsymbol{\mu}) \sim T^2(k, n-1)$
- That is,  $\frac{n-k}{k(n-1)} t^2 \sim F(k, n-k)$
- When  $k = 1$ , reduces to the familiar  $t^2 = F(1, n-1)$
- Test  $H_0 : \boldsymbol{\mu} = \boldsymbol{\mu}_0$



## Test *Collections* of Contrasts

$H_0 : \mathbf{L}\boldsymbol{\mu} = \mathbf{h}$ , where  $\mathbf{L}$  is  $r \times k$

- $t^2 = n (\bar{\mathbf{X}}_n - \boldsymbol{\mu})' \mathbf{S}^{-1} (\bar{\mathbf{X}}_n - \boldsymbol{\mu}) \sim T^2(k, n - 1)$ ,  
so if  $H_0$  is true
- $t^2 = n (\mathbf{L}\bar{\mathbf{X}}_n - \mathbf{h})' (\mathbf{L}\mathbf{S}\mathbf{L}')^{-1} (\mathbf{L}\bar{\mathbf{X}}_n - \mathbf{h}) \sim T^2(r, n - 1)$
- Could also calculate contrast variables, like differences.
  - Expected value of the contrast is the contrast of expected values.
  - Just test (simultaneously) whether the means of the contrast variables are zero, using the first formula.
- For 2 or more within-cases factors, use contrasts to test for main effects, interactions

# Compare Wald-like tests

Recall

- If  $\mathbf{Y}_n = \sqrt{n}(\mathbf{T}_n - \boldsymbol{\theta}) \xrightarrow{d} \mathbf{Y} \sim N_k(\mathbf{0}, \boldsymbol{\Sigma})$ , then

$$W_n = n(\mathbf{L}\mathbf{T}_n - \mathbf{h})' (\mathbf{L}\widehat{\boldsymbol{\Sigma}}_n\mathbf{L}')^{-1} (\mathbf{L}\mathbf{T}_n - \mathbf{h}) \xrightarrow{d} W \sim \chi^2(r)$$

$$t^2 = n(\mathbf{L}\bar{\mathbf{X}}_n - \mathbf{h})' (\mathbf{L}\mathbf{S}\mathbf{L}')^{-1} (\mathbf{L}\bar{\mathbf{X}}_n - \mathbf{h}) \sim T^2(r, n-1)$$

- And

$$F = \frac{n-r}{r(n-1)} t^2 \sim F(r, n-r) \Rightarrow t^2 = \frac{n-1}{n-r} rF \xrightarrow{d} Y \sim \chi^2(r)$$

- So the Hotelling  $t$ -squared test is robust with respect to normality.