

Lecture 18: t-tests

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Previously: Wald tests for a population mean

Suppose X_1, \dots, X_n are an iid sample from a population with mean μ and variance σ^2 . We wish to test

$$H_0 : \mu = \mu_0 \qquad H_A : \mu \neq \mu_0$$

► If σ^2 is known:

$$Z_n = \frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma}$$

► If σ^2 is unknown:

$$Z_n = \frac{\sqrt{n}(\bar{X} - \mu_0)}{S}$$

Issue: Wald tests with small n

The Wald test for a population mean μ relies on

$$Z_n = \frac{\sqrt{n}(\bar{X}_n - \mu)}{s} \approx N(0, 1)$$

- ▶ $Z_n \xrightarrow{d} N(0, 1)$ as $n \rightarrow \infty$
- ▶ But for small n , Z_n is not normal, even if $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$

What is the exact distribution of $\frac{\sqrt{n}(\bar{X}_n - \mu)}{s}$?

t -tests

If $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$, then

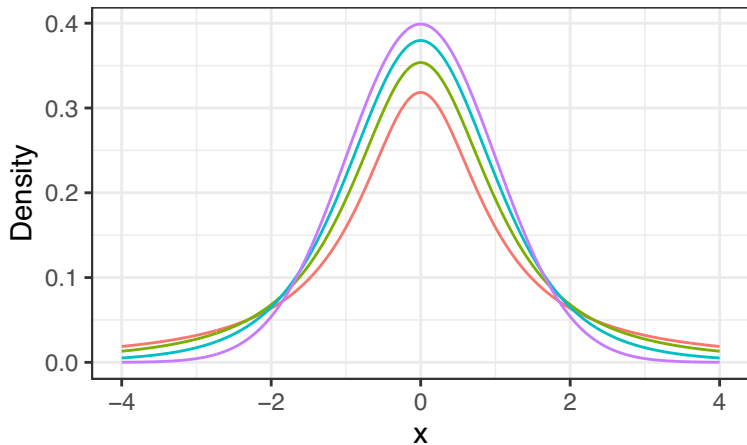
$$\frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma} \sim N(0, 1)$$

$$\frac{\sqrt{n}(\bar{X}_n - \mu)}{s} \sim t_{n-1}$$

t distribution
with $n-1$ df

t-distribution

Formal: X_1, X_2, X_3, \dots $X_i \sim t_i$
 $X_n \xrightarrow{d} N(0, 1)$



— df = 1 — df = 2 — df = 5 — N(0, 1)

As $df \uparrow$, $t \rightarrow N(0, 1)$

t distribution

Definition: Let $Z \sim N(0, 1)$ and $V \sim \chi_d^2$ be independent. Then

$$T = \frac{Z}{\sqrt{V/d}} \sim t_d$$

$$Z = \frac{\sqrt{n}(\bar{X} - \mu)}{\sigma} \sim N(0, 1)$$

Claim: If $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$, then

$$\frac{\sqrt{n}(\bar{X}_n - \mu)}{s} \sim t_{n-1}$$

$$\frac{\sqrt{n}(\bar{X} - \mu)}{s} = \frac{\sqrt{n}(\bar{X} - \mu)/\sigma}{s/\sigma} = \frac{Z}{s/\sigma} = \frac{Z}{\sqrt{((n-1)s^2/\sigma^2)/(n-1)}}$$

$$s^2 = \frac{1}{n-1} \sum_i (X_i - \bar{X})^2 \Rightarrow \frac{s}{\sigma} = \sqrt{\frac{(n-1)s^2}{(n-1)\sigma^2}} = \sqrt{((n-1)s^2/\sigma^2)/(n-1)}$$

$$\Rightarrow \text{WTS: } (n-1) \frac{s^2}{\sigma^2} \sim \chi_{n-1}^2 \quad \text{and} \quad \frac{(n-1)s^2}{\sigma^2} \perp \frac{\sqrt{n}(\bar{X} - \mu)}{\sigma}$$

↑
(independent)

What we want to show

$$(n-1) \frac{s^2}{\sigma^2} \sim \chi_{n-1}^2$$

$$(n-1) \frac{s^2}{\sigma^2} \stackrel{d}{=} \frac{\sqrt{n}(\bar{X} - \mu)}{\sigma}$$

$$(n-1) \frac{s^2}{\sigma^2} = \frac{1}{\sigma^2} \sum_i (x_i - \bar{x})^2 = \sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)^2$$

Know: $\frac{x_i - \mu}{\sigma} \stackrel{iid}{\sim} N(0,1) \Rightarrow \sum_i \left(\frac{x_i - \mu}{\sigma} \right)^2 \sim \chi_n^2$

Know: if $v_1 \sim \chi_{d_1}^2$, $v_2 \sim \chi_{d_2}^2$ independent, then
 $v_1 + v_2 \sim \chi_{d_1 + d_2}^2$

Idea: $\underbrace{\sum_i \left(\frac{x_i - \mu}{\sigma} \right)^2}_{\chi_n^2} = \underbrace{\sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)^2}_{\text{hope is } \chi_{n-1}^2} + \underbrace{\text{Something}}_{\chi_1^2}$

Decomposing the sum of squares

$$\begin{aligned}\sum_i (x_i - \bar{x}) &= \sum x_i - n\bar{x} \\ &= n\bar{x} - n\bar{x} = 0\end{aligned}$$

$$\sum_{i=1}^n \left(\frac{X_i - \mu}{\sigma} \right)^2 = \sum_{i=1}^n \left(\frac{X_i - \bar{X}}{\sigma} \right)^2 + n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2$$

$$\begin{aligned}\sum_i \left(\frac{x_i - \mu}{\sigma} \right)^2 &= \sum_i \left(\frac{x_i - \bar{x} + \bar{x} - \mu}{\sigma} \right)^2 \\ &= \sum_i \left[\left(\frac{x_i - \bar{x}}{\sigma} \right)^2 + \left(\frac{\bar{x} - \mu}{\sigma} \right)^2 + 2 \left(\frac{x_i - \bar{x}}{\sigma} \right) \left(\frac{\bar{x} - \mu}{\sigma} \right) \right] \\ &= \sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)^2 + \underbrace{\sum_i \left(\frac{\bar{x} - \mu}{\sigma} \right)^2}_{n \left(\frac{\bar{x} - \mu}{\sigma} \right)^2} + 2 \underbrace{\sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right) \left(\frac{\bar{x} - \mu}{\sigma} \right)}_{\left(\frac{\bar{x} - \mu}{\sigma} \right) \sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)} \\ &= \sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)^2 + n \left(\frac{\bar{x} - \mu}{\sigma} \right)^2\end{aligned}$$

Cochran's theorem

$$Z^T Z = \sum_i Z_i^2$$

Let $Z_1, \dots, Z_n \stackrel{iid}{\sim} N(0, 1)$, and let $Z = [Z_1, \dots, Z_n]^T$. Let

$A_1, \dots, A_k \in \mathbb{R}^{n \times n}$ be symmetric matrices such that

$Z^T Z = \sum_{i=1}^k Z^T A_i Z$, and let $r_i = \text{rank}(A_i)$. Then the following are equivalent:

- ▶ $r_1 + \dots + r_k = n$
- ▶ The $Z^T A_i Z$ are independent
- ▶ Each $Z^T A_i Z \sim \chi_{r_i}^2$

Application to t-tests

$$\text{Let } Z_i = \frac{x_i - \mu}{\sigma} \stackrel{iid}{\sim} N(0,1)$$

$$Z = \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}$$

$$\underbrace{\sum_i \left(\frac{x_i - \mu}{\sigma} \right)^2}_{Z^T Z} = \sum_i \underbrace{\left(\frac{x_i - \bar{x}}{\sigma} \right)^2}_{\uparrow} + \underbrace{n \left(\frac{\bar{x} - \mu}{\sigma} \right)^2}_{\uparrow}$$

want to find A_1, A_2 st $Z^T A_1 Z$ $Z^T A_2 Z$

$$n \left(\frac{\bar{x} - \mu}{\sigma} \right)^2 = n (\bar{Z})^2 = \frac{1}{n} \left(\sum_j z_j \right)^2$$

$$= \frac{1}{n} [z_1 \dots z_n] \underbrace{\begin{bmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{bmatrix}}_{J_n} \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}$$

$J_n = n \times n$ matrix of all 1s

$$= Z^T \left(\frac{1}{n} J_n \right) Z$$

$$A_2 = \frac{1}{n} J_n$$

Application to t-tests

$$\underbrace{\sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)^2}_{Z^T Z} = \underbrace{\sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)^2}_{Z^T Z} + \underbrace{n \left(\frac{\bar{x} - \mu}{\sigma} \right)^2}_{Z^T \left(\frac{1}{n} J_n \right) Z}$$

↓

$$\begin{aligned} Z^T Z - Z^T \left(\frac{1}{n} J_n \right) Z \\ = Z^T \left(I - \frac{1}{n} J_n \right) Z \end{aligned}$$

$$\Rightarrow A_1 = I - \frac{1}{n} J_n \qquad A_2 = \frac{1}{n} J_n$$

Now want to find the ranks:

Fact: if A is idempotent ($A = A^2$), then
 $\text{rank}(A) = \text{tr}(A)$

$$\text{tr}(A_2) = \frac{1}{n} \text{tr}(J_n) = \frac{1}{n} \cdot n = 1$$

$$\begin{aligned} \text{tr}(A_1) &= \text{tr}\left(I - \frac{1}{n} J_n\right) = \text{tr}(I) - \text{tr}\left(\frac{1}{n} J_n\right) \\ &= n - 1 \end{aligned}$$

Application to t-tests

$$\text{So: } \underbrace{\sum_i \left(\frac{x_i - \mu}{\sigma} \right)^2}_{Z^T Z} = \underbrace{\sum_i \left(\frac{x_i - \bar{x}}{\sigma} \right)^2}_{Z^T A_1 Z} + \underbrace{n \left(\frac{\bar{x} - \mu}{\sigma} \right)^2}_{Z^T A_2 Z}$$
$$\sim \chi_n^2$$

$$\text{rank}(A_1) = n-1$$

$$\text{rank}(A_2) = 1$$

By Cochran's theorem:

$$\cdot Z^T A_1 Z \perp Z^T A_2 Z \Rightarrow \frac{(n-1)s^2}{\sigma^2} \perp \sqrt{n} \left(\frac{\bar{x} - \mu}{\sigma} \right)$$

$$\cdot Z^T A_1 Z \sim \chi_{n-1}^2, \quad n \left(\frac{\bar{x} - \mu}{\sigma} \right)^2 \sim \chi_1^2$$

$$\Rightarrow \frac{(n-1)s^2}{\sigma^2} \sim \chi_{n-1}^2$$

$$\Rightarrow \frac{\sqrt{n}(\bar{x} - \mu)}{s} = \frac{\sqrt{n}(\bar{x} - \mu)/\sigma}{\sqrt{((n-1)s^2/\sigma^2)/(n-1)}} \sim t_{n-1} //$$