

Paired data. No independence assumption.

Data: (X_{1i}, X_{2i}) are dependent $((X_{1i}, X_{2i}) \text{ indep from } (X_{1j}, X_{2j}))$

Forex, measure "before" and "after" on same subject.

Look at difference $Y_i = X_{1i} - X_{2i}$. \leftarrow one Data pt for each subject.
 \uparrow before \uparrow after

$\Rightarrow Y_1, \dots, Y_n$ new RS. assume $\frac{\bar{Y} - \mu_Y}{s_Y/\sqrt{n}} \sim t_{n-1}$ if $Y_i \sim \text{Normal}$.

Perform a one-sample t-test of $H_0: \mu_Y = 0$ vs. $H_1: \mu_Y \neq 0$.

If n large use $N(0,1)$ instead of t_{n-1} .

Ex: Data: # seizures before & after drug. $n=20$, $\alpha=0.05$. $\bar{Y} = 12.5$, $s_Y = 14.12$.

$$\frac{12.5 - 0}{14.12/\sqrt{20}} = 3.96. \quad p\text{-value} = P(|t| \geq 3.96) \dots < 0.005 \quad (*)$$

$$CR: |t| > t_{\frac{\alpha}{2}, n-1} = t_{0.025, 19} = 2.096.$$

\downarrow
so Reject H_0 .

So Reject H_0 . There is a difference in # of seizures.

§ 13.4 tests concerning variance

One-sample test.

Ex: population SD is 0.17, but higher variation one day. test whether SD increased.

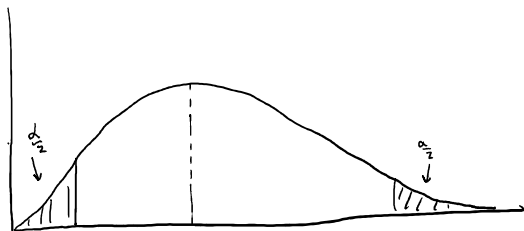
Test: $H_0: \sigma^2 = \sigma_0^2 (= 0.17^2)$ vs. $H_1: \sigma^2 > \sigma_0^2$

$$\text{Note } V^2 = \frac{(n-1)S^2}{\sigma_0^2} \sim \chi^2_{n-1} \quad (1+1 \quad \sigma_0^2 \quad \dots)$$

Note $\chi^2 = \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$ (Thm 8.11) when H_0 is true.

To find a level α test, use χ_{n-1}^2 dist to determine CR.

* Reject H_0 in favor of $H_1: \sigma^2 \neq \sigma_0^2$ if $\chi^2 \notin [\chi_{1-\frac{\alpha}{2}, n-1}^2, \chi_{\frac{\alpha}{2}, n-1}^2]$
 $< \chi_{1-\alpha, n-1}^2$
 $> \chi_{\alpha, n-1}^2$



Ex: $n=30$, $S=0.21$, original pop $\sigma_0=0.17$. evidence that σ increased? $\alpha=0.05$.

Sol: $H_0: \sigma^2 = 0.17^2$ vs $H_1: \sigma^2 > 0.17^2$

$$\chi^2 = \frac{(30-1)0.21^2}{0.17^2} = 44.25$$

Method 1: $\chi_{0.05, 29}^2 = 42.557 < \chi^2 \Rightarrow$ reject H_0 .

Method 2: $p\text{-val} = P(\chi^2 > 44.25) \in (0.025, 0.05) \Rightarrow$ reject H_0 .

Two-sample test:

Compare SDs from independent samples

Exercise 12.26 \Rightarrow LRT based on $\frac{S_1^2}{S_2^2}$ when we assume the populations are normal.

Theorem 8.15: $F = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} \sim F_{n_1-1, n_2-1}$. n_1, n_2 sample size of the two samples.

Consider test $H_0: \sigma_1^2 = \sigma_2^2$. under H_0 , $F = \frac{S_1^2}{S_2^2} \sim F_{n_1-1, n_2-1} = (f_{\frac{\alpha}{2}, n_2-1, n_1-1})^{-1}$

Test: reject H_0 in favor of $H_1: \sigma_1^2 \neq \sigma_2^2$ if $\frac{S_1^2}{S_2^2} \notin [f_{1-\frac{\alpha}{2}, n_1-1, n_2-1}, f_{\frac{\alpha}{2}, n_1-1, n_2-1}]$
 $\frac{S_1^2}{S_2^2} < f_{1-\alpha, n_1-1, n_2-1}$
 $\frac{S_1^2}{S_2^2} > f_{\alpha, n_1-1, n_2-1}$

Note: Can't use these formulas if not normal populations.