

Model:

One Sample: X_1, \dots, X_n iid $N(\mu, \sigma^2)$

R.V.s with $E(X_i) = \mu$

$Var(X_i) = \sigma^2$

ALWAYS UNKNOWN

\Rightarrow PARAMETER OF INTEREST

MAY BE

KNOWN $= \sigma_0^2$

Z-test

$$\frac{\bar{X} - \mu_0}{\sigma_0 / \sqrt{n}}$$

to test

$$H_0: \mu = \mu_0$$

KNOWN
ST. ERROR

SINCE

$$\sqrt{Var(\bar{X})} = \frac{\sigma_0}{\sqrt{n}}$$

$\sim N(0,1)$

if H_0

true

SO LONG AS

\bar{X} is (approx.)

normally distributed

X_i 's normal

"large n"

Central Limit Thm

UNKNOWN

MUST ESTIMATE

ST. ERROR

USING

$$\frac{S}{\sqrt{n}} \text{ where } S = \sqrt{\frac{1}{n-1} \sum (X_i - \bar{X})^2}$$

t-test

$$\frac{\bar{X} - \mu_0}{S / \sqrt{n}} \sim t_{n-1}$$

if H_0 true

SO LONG AS

X_i 's NORMAL



Two-Sample Model

- Paired T-test
- 2 "indep sample" t-test

~~X₁, ..., X_n~~

Paired test: 1-sample t-test on differences.

2 indep samples.

$$X_1, \dots, X_n \sim N(\mu_x, \sigma_x^2)$$

$$Y_1, \dots, Y_n \sim N(\mu_y, \sigma_y^2)$$

Parameter of interest is $\mu_x - \mu_y$.

The Estimator is $\bar{X} - \bar{Y} \sim N\left(\mu_x - \mu_y, \frac{\sigma_x^2}{m} + \frac{\sigma_y^2}{n}\right)$

"TRUE" St. Error is $\sqrt{\frac{\sigma_x^2}{m} + \frac{\sigma_y^2}{n}}$.

2 Sub-cases:

$$1) \sigma_x^2 = \sigma_y^2 (= \sigma^2 \text{ say}).$$

~~We est~~

$$\text{Then St. Error} = \sigma \sqrt{\frac{1}{m} + \frac{1}{n}}$$

We estimate ~~st. error~~ by using

$$S_P = \sqrt{\frac{(m-1)S_x^2 + (n-1)S_y^2}{m+n-2}}$$

Then the t-statistic is

$$\frac{(\bar{X} - \bar{Y}) - \underbrace{(\mu_{x0} - \mu_{y0})}_{\text{often } = 0}}{S_P \sqrt{\frac{1}{m} + \frac{1}{n}}} \sim t_{m+n-2} \quad \text{if } \mu_x - \mu_y = \mu_{x0} - \mu_{y0}$$

2) $\sigma_x^2 \neq \sigma_y^2$: Then we estimate the st. error using.

$$\sqrt{\frac{S_x^2}{m} + \frac{S_y^2}{n}}$$

The t-stat is

$$\frac{\bar{X} - \bar{Y} - (\mu_{X0} - \mu_{Y0})}{\sqrt{\frac{s_X^2}{m} + \frac{s_Y^2}{n}}} \sim ??$$

under
 $H_0: \mu_X - \mu_Y = \mu_{X0} - \mu_{Y0}$

approx t-dist

with "data-dependent" degrees of freedom

"Welch test"