

Note:

1. 95% C.I. for people's height is (165, 185), how to interpret this interval.

We are 95% confident/certain that the true mean of people's height is between 165, 185.

2. Independence: X and Y

$$\textcircled{1} \quad \text{Cov}(X, Y) = E(XY) - E(X) \cdot E(Y)$$

$$\textcircled{2} \quad p(X \text{ and } Y) = p(X) \cdot p(Y)$$

$$\textcircled{3} \quad p(X|Y) = p(X)$$

$$\textcircled{4} \quad p(Y|X) = p(Y)$$

$$\left. \begin{array}{l} \text{Mutually Exclusive} \\ p(X \cap Y) = 0 \end{array} \right\}$$

## 9.1 Hypothesis Testing

**Statistical hypothesis:**

is a statement about the parameter of one or more populations.

— Null Hypothesis ( $H_0$ ) is that the population parameter is equal to some value.

$$H_0: \mu = 65$$

— Alternative Hypothesis ( $H_a/H_1$ ) is that the population parameter is  $>$ ,  $<$ ,  $\neq$  to some value.

$$H_1: \mu > 65$$

or

$$H_1: \mu < 65$$

or

$$H_1: \mu \neq 65$$

**Critical region:**

consists of values of the test statistic that would result in rejecting  $H_0$ .

**Test statistic:**

is a statistic whose value is used to make a decision about hypothesis from a sample.

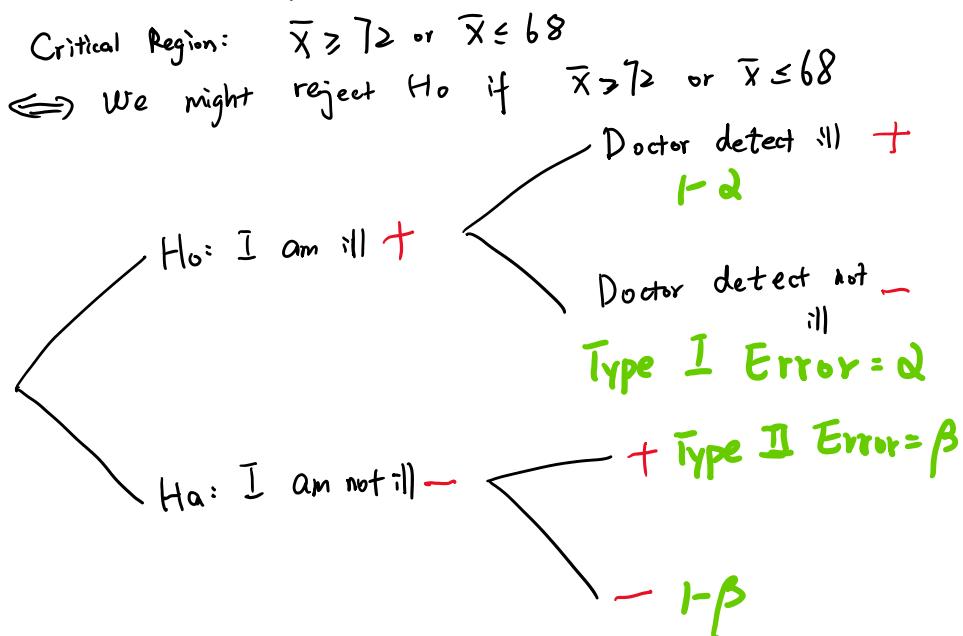
$$(\bar{x})$$

$$\text{Example: } H_0: \mu = 70$$

$$H_1: \mu \neq 70$$

$$\text{Critical Region: } \bar{x} \geq 72 \text{ or } \bar{x} \leq 68$$

... i.e. if  $\bar{x} \geq 72$  or  $\bar{x} \leq 68$



### Type I Error:

A Type I Error occurs if we reject  $H_0$  but  $H_0$  is actually true.  $\alpha$  = significance level.

### Type II Error:

A Type II Error occurs if we fail to  $H_0$  but  $H_1/H_0$  is true.  $\beta$

$$1 - \beta = \text{power of the test.}$$

Example: In testing,  $H_0: \mu = 70$ ,  $H_1: \mu \neq 70$   
based on a sample of size 64 from Normal.

Suppose that we reject  $H_0$  if  $\bar{X} \geq 72$  or  $\bar{X} \leq 68$

Find  $\alpha$  when  $\sigma = 16$ .

$$\alpha = P(\text{Type I Error}) \quad (\text{Assume } H_0 \text{ is True} \Leftrightarrow \mu = 70)$$

$$= P(\bar{X} \geq 72) + P(\bar{X} \leq 68)$$

$$\left( Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \right) = P(Z \geq \frac{72 - 70}{16/\sqrt{64}}) + P(Z \leq \frac{68 - 70}{16/\sqrt{64}})$$

$$= P(Z \geq 1) + P(Z \leq -1)$$

$$= 0.3173$$

Note: Increasing Sample size will always decrease  $\alpha$ .

$$\beta = P(\text{Type II Error})$$

$$= P(\text{failing to reject } H_0 \text{ when } H_1 \text{ is true})$$

$$= P(\text{do not reject } H_0 \mid H_1 \text{ True})$$

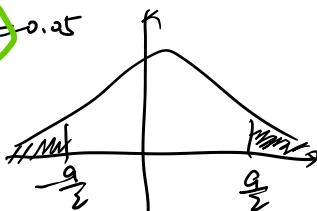
$$\begin{aligned}
 &= P(\text{do not reject } H_0 \mid H_1 \text{ True}) \\
 \text{Find } \beta \text{ if } M = 74 &\quad \text{true mean} \\
 &= P(68 \leq \bar{X} \leq 72 \mid M = 74) \\
 &= P\left(\frac{68-74}{16/\sqrt{64}} \leq Z \leq \frac{72-74}{16/\sqrt{64}}\right) \\
 &= P(-1 \leq Z \leq -3) \\
 &= 0.1573
 \end{aligned}$$

Example: Find the boundary of the critical region

if we want  $\alpha = 0.05$ .

$$\begin{aligned}
 P(\bar{X} \geq 70+a) + P(\bar{X} \leq 70-a) &= 0.05 \\
 P(Z \geq \frac{70+a-70}{16/\sqrt{64}}) + P(Z \leq \frac{70-a-70}{16/\sqrt{64}}) &= 0.05
 \end{aligned}$$

$$P(Z \geq \frac{a}{2}) + P(Z \leq -\frac{a}{2}) = 0.05$$



$$\Leftrightarrow 2P(Z \geq \frac{a}{2}) = 0.05$$

$$\frac{a}{2} = 1.96$$

$$a = 3.92$$

Critical Region:  $\bar{X} \geq 73.92$  or  $\bar{X} \leq 66.08$ .

Note: 1. For a fixed sample size, decreasing  $\alpha$ , increases  $\beta$ .

2. For a fixed  $\alpha$ , increase sample size,  $n$  will decrease  $\beta$ .

## P-value

The p-value is the probability of obtaining a value of the test statistic at least as extreme as the observed value when  $H_0$  is true

(Same direction with  $H_1$ )

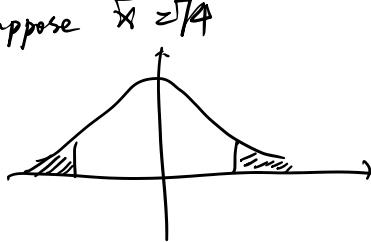
Example:  $H_0: \mu = 70$   
 $H_1: \mu \neq 70$  two tailed

Example:  $H_0: \mu = 70$   
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$n=64, \sigma=16$ . Suppose  $\bar{x}=74$

Find p-value.  $\downarrow$  alternative ' $\neq$ '

$$p\text{-value} = 2 \cdot P(\bar{X} > 74)$$



$$= 2 \cdot P\left(Z > \frac{74-70}{16/\sqrt{64}}\right)$$

$$\approx 2 \cdot P(Z > 2)$$

$$\approx 0.0455$$

(the probability of mistakes to reject  $H_0$ )  
when  $\bar{x}=74$ .

If for a given  $\alpha$  for a test.

p-value  $< \alpha$ , then reject  $H_0$ .

### Hypothesis Test

① Hypothesis:  $H_0: \mu = 0$  always equal

$H_1:$   
 $>$   
 $<$   
 $\neq$

② Test statistic:

③ Critical Region / Critical Value

p-value

④ Conclusion:

Test stat is in the critical region

$\Leftrightarrow$  p-value  $< \alpha$

$\Leftrightarrow$  reject  $H_0$ .

### q.2 Test on the mean of Normal ( $\sigma$ known)

①  $H_0: \mu = \mu_0$

$H_1: \mu \neq \mu_0$

$H_1$ : ①  $\mu \neq \mu_0$

②  $\mu > \mu_0$

③  $\mu < \mu_0$

② Test statistic

$$Z_0 = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$$

③

Critical Region:	① $Z_0 > Z_{\alpha/2}^*$ or $Z_0 < -Z_{\alpha/2}^*$	② $Z_0 > Z_\alpha^*$	③ $Z_0 < -Z_\alpha^*$
p-value :	$2 p(Z >  Z_0 )$	$p(Z > Z_0)$	$p(Z < Z_0)$

④ Conclusion:  $p\text{-value} < \alpha$ , reject  $H_0$ .

For all tests  
or  $| \text{Test stat} | > | \text{Critical value} |$ , reject  $H_0$

Example:

The average house price in Canada in 2010 was  $\$339100$  with standard deviation  $\$194300$ . according to CREA.

Today, a consumer randomly select 54 house sample, find an average of  $\$418200$ .

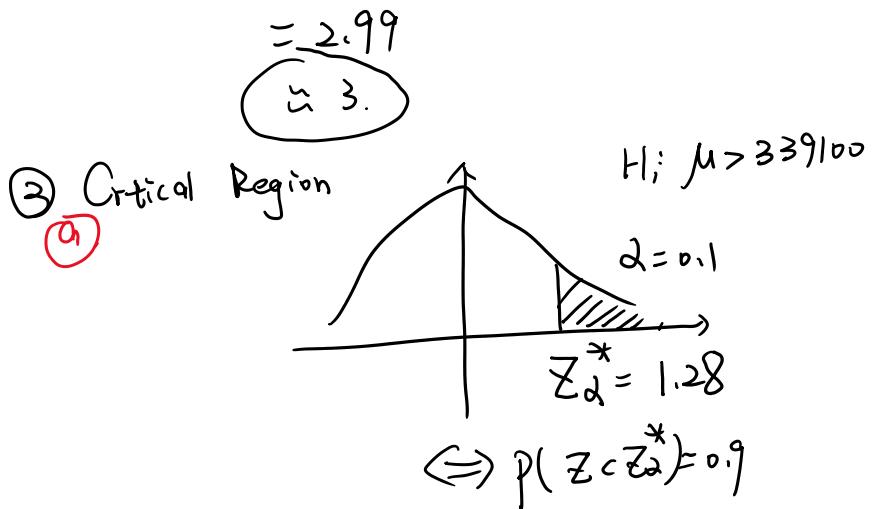
State a hypothesis to test whether the price has increased significantly,  $\alpha = 0.1$

① Hypothesis:  $H_0: \mu = 339100$   
 $H_a: \mu > 339100$

② Test Stat:

$$Z_0 = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} = \frac{418200 - 339100}{194300/\sqrt{54}}$$

$$\approx 2.99$$



④ p-value =  $p(Z > Z_0)$   
 $= p(Z > 3)$   
 $= 0.001$

④ Conclusion:

①  $| \text{Test Stat} | = 3 \rightarrow | \text{Critical Value} | = | z_{\alpha}^* |$   
 $= 1.28$

②  $p\text{-value} = 0.001 < \alpha = 0.01$

$\Leftrightarrow$  reject  $H_0$ .

9.3 Tests on the Mean of a Normal Distribution ( $\sigma$  unknown)

② Test Stat:

$$t_{n-1} = \frac{\bar{x} - \mu}{s/\sqrt{n}}$$

All other steps same 9.1

Example: A company claims that its batteries last at least 35 hours.

Here is sample of their products:

35 34 32 31 34 34 32 33 35 55 32

$n=12$

Do a hypothesis test for battery does not last longer than 35 hours.  $\alpha = 0.05$

$$H_1: \begin{cases} > \\ < \end{cases} \text{ one tail}$$

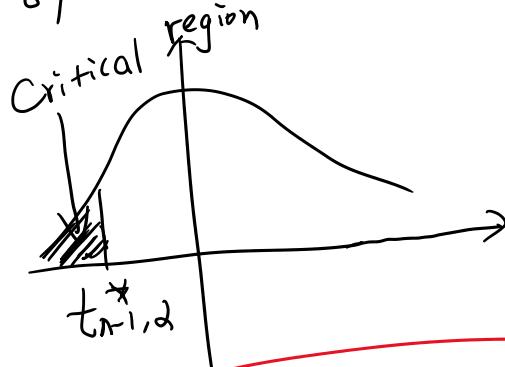
$\neq$  two tail

$$\textcircled{1} H_0: \mu = 35$$

$$H_1: \mu < 35 \quad \text{one tail}$$

$$\bar{x} = 34.8333 \quad S = 6.506$$

$$\textcircled{2} \text{ Test Stat: } t_{n-1} = \frac{\bar{x} - \mu}{S/\sqrt{n}} = -0.089$$



$$\textcircled{3} H_1: \mu < 35$$

$$\begin{aligned} -t_{n-1, 2}^* &= \text{critical value} \\ &= -t_{11, 0.05}^* \Leftrightarrow \text{one tail) } t, \alpha = 0.05, df = 11 \\ &\approx -1.796 \end{aligned}$$

$$\textcircled{4} | \text{Test Stat} | < | \text{Critical Value} |$$

∴ do not reject  $H_0$ .

## 9.5 Tests on a Population Proportion.

$$\textcircled{1} H_0: p = p_0$$

$$H_1: p > p_0$$

$$p < p_0$$

$$p \neq p_0$$

where  $p_0$  is a given number.

$$\textcircled{2} \text{ Test Stat: }$$

$$Z_0 = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}} \sim N(0,1) \quad (\hat{p} \text{ sample statistic})$$

③ > same with 9.1

④ Example: In 2006,  $P_0 = 0.34$  of students had not been absent from school even once.

In 2011 survey responses from  $n = 8302$  students showed the rate is  $\hat{P} = 0.33$ . Officials would be concerned if student attendance were declining. Do a hypothesis test  $\alpha = 0.05$ .

①  $H_0: P = 0.34$

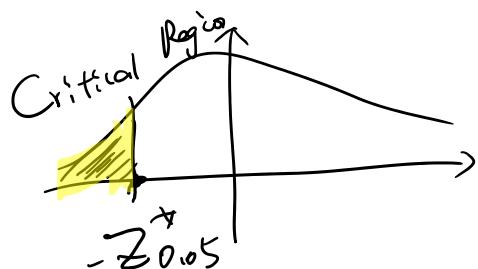
$H_1: P < 0.34$

② Test Stat

$$Z_0^* = \frac{\hat{P} - P_0}{\sqrt{\frac{P_0(1-P_0)}{n}}} = \frac{0.33 - 0.34}{\sqrt{\frac{0.34(1-0.34)}{8302}}} = -1.923$$

③

Critical value  $|Z_{0.05}^*| = 1.645$



④  $|Test Stat| > |Critical Value|$

$\Leftrightarrow$  reject  $H_0$ .

$$\Leftrightarrow p\text{-value} = P(Z < -1.923) = 0.0272 < 0.05$$

$\therefore$  calculation for the exams.

There will be no  $\beta$ 's calculation for the exams.

## 10.1 - 10.2 Inference for difference of two means

Case I:

Two independent samples from population ( $\sigma_1, \sigma_2$  unknown)

- ① Hypothesis:  
 $H_0: \mu_1 = \mu_2$   
 $H_1: \mu_1 > \mu_2$   
 $\mu_1 < \mu_2$   
 $\mu_1 \neq \mu_2$

② Test Statistic

$$Z = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0, 1)$$

- ③ ④ Same with 9.)

Case II

Two independent sample ( $\sigma_1, \sigma_2$  unknown, assuming equal variances for two populations)

- ①

② Test Stat:

$$t_{df} = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{s_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

where

$$df = n_1 + n_2 - 2$$

$$s_p^2 = \frac{(n_1 - 1)\sigma_1^2 + (n_2 - 1)\sigma_2^2}{n_1 + n_2 - 2}$$

$$S_p^2 = \frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1+n_2-2}$$

Pooled Variance

③, ④ same with 9.1

Example: A developer wants to know if the houses in two different neighbourhoods were built at roughly the same time. She took random samples

Neighbour 1

$$n_1 = 30$$

$$\bar{x}_1 = 57.2$$

$$s_1 = 7.5$$

Neighbour 2

$$n_2 = 35$$

$$\bar{x}_2 = 47.6$$

$$s_2 = 7.85$$

Does N1 has significantly larger age than N2?  $\alpha = 0.05$

Assume the population variances are equal. ( $\sigma_1^2 = \sigma_2^2$ )

①  $H_0: \mu_1 = \mu_2$

$H_1: \mu_1 > \mu_2$  one tail

② Test Stat:

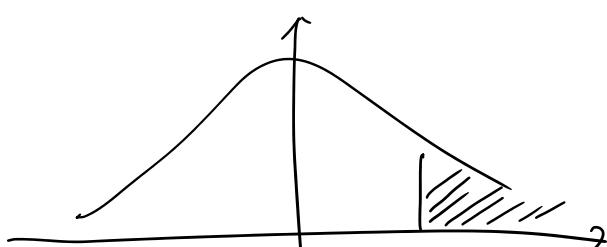
$$t_{df} = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

$$S_p^2 = \frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1+n_2-2} = 59.218$$

$$df = n_1 + n_2 - 2 = 63$$

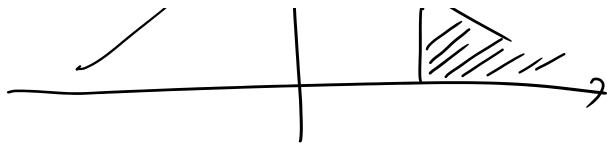
$$t_{df} = 5.01$$

③ critical value



③ Critical value

$$t_{63, 0.05}^* = 1.671$$



④  $| \text{Test Stat} | > | \text{Critical Value} |$

∴ reject  $H_0$ .

CI for  $(\mu_1 - \mu_2)$ : ( $\sigma_1^2 = \sigma_2^2$ )

$$\bar{x}_1 - \bar{x}_2 \pm t_{n_1 + n_2 - 2, \frac{\alpha}{2}, \text{two tail}}^* \cdot \sqrt{s_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$$

Case III: two independent sample ( $\sigma_1^2 \neq \sigma_2^2$ )

② Test Stat:

$$t_{df} = \frac{\bar{x}_1 - \bar{x}_2 - (\mu_1 - \mu_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$\text{where } df = \frac{\left( \frac{s_1^2}{n_1} + \frac{s_2^2}{n_2} \right)^2}{\frac{(s_1^2/n_1)^2}{n_1-1} + \frac{(s_2^2/n_2)^2}{n_2-1}}$$

CI for  $\mu_1 - \mu_2$  ( $\sigma_1^2 \neq \sigma_2^2$ )

$$\bar{x}_1 - \bar{x}_2 \pm t_{df, \frac{\alpha}{2}, \text{two tail}} \cdot \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}.$$

Example: Two suppliers manufacture a plastic gear used in a laser printer.

Supplier 1:  $\bar{x}_1 = 296$        $s_1 = 12$        $n_1 = 10$

“ 2:  $\bar{x}_2 = 321$        $s_2 = 22$        $n_2 = 16$ .

→ ... we hypothesize that the suppliers provide gears with

Test the hypothesis that the suppliers provide gears with different mean. Assume variances are not equal,  $\alpha = 0.05$ .

① Hypothesis:  $H_0: \mu_1 = \mu_2$   
 $H_1: \mu_1 \neq \mu_2$  two tail.

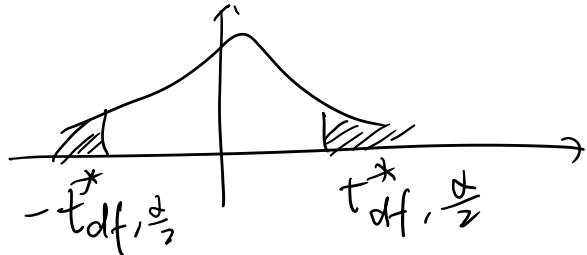
② Test Statistic:

$$t_{df} = \frac{290 - 321 - (0)}{\sqrt{\frac{12^2}{10} + \frac{22^2}{16}}} = -4.64$$

$$df = 23.72 \approx \begin{matrix} 23 \\ \text{round down} \end{matrix}$$

③ Critical Value

$$t_{23, \frac{0.05}{2}}^*, \text{two tail} \\ = 1.714$$



④  $| \text{Test Stat} | > | \text{Critical Value} |$   
 $\Leftrightarrow \text{reject } H_0$ .