

$$f(x) = 3x^2, \quad 0 \leq x \leq 1$$

$$n = 2$$

$$E(x) = \int x f(x) dx = \int 3x^3 dx = \frac{3}{4} x^4 \Big|_0^1 = \frac{3}{4}$$

$$E(x^2) = \int x^2 f(x) dx = \int 3x^5 dx = \frac{3}{5} x^6 \Big|_0^1 = \frac{3}{5}$$

$$\therefore V(x) = E(x^2) - E(x)^2 = \frac{3}{5} - \frac{9}{16} = \frac{48-45}{80} = \frac{3}{80}$$

$$V(\bar{x}) = \frac{V(x)}{n} = \frac{3}{160}$$

$$\begin{aligned} V(\bar{x}) &= V\left(\frac{x_1 + \dots + x_n}{n}\right) = \frac{1}{n^2} (V(x_1) + \dots + V(x_n)) \\ &= \frac{1}{n^2} \cdot n \cdot \sigma^2 \\ &= \frac{\sigma^2}{n} = \frac{V(x)}{n} \end{aligned}$$

Chapter 9:

Hypothesis Test

C.I. Definition

1000 C.I., 95% C.I.

950 of C.I. contains the true mean.

A statistical hypothesis:

A statement about the parameters of the population.

Hypothesis Test:

$$\begin{aligned} 1. \text{ Hypothesis: } & H_0: \mu = 100,000 && \text{Average Income} \\ & H_A: \mu < 100,000 \end{aligned}$$

Null hypothesis: H_0 is the population parameter that equal to some value.

Alternative hypothesis: H_A (H_1) is that the

that equal

Alternative hypothesis: H_a (H_A) is that the population parameter " $>$ ", " $<$ " " $\neq$ " to some value.

e.g. $H_0: \underline{\mu = 170}$, $H_A: \mu \neq 170$.

$H_0: \underline{\mu = 170}$, $H_A: \mu > 170$.

2. Test Statistic:

$\bar{X}_1 = 50,000$, $\bar{X}_2 = 90,000$ ✓

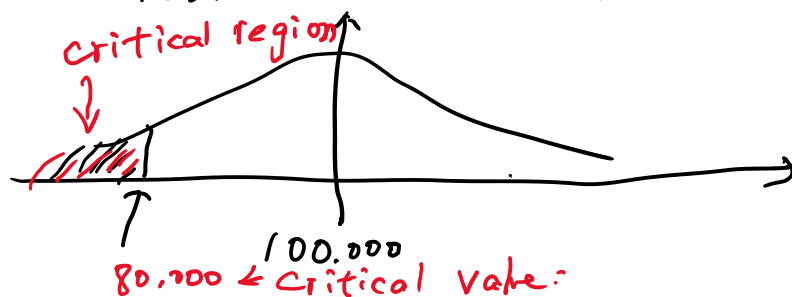
is a statistic: whose value is used to make a decision about the hypothesis.

3. Critical region, Critical value.

$H_0: \mu = 100,000$

$H_A: \mu < 100,000$

Test statistic: $\bar{X} = 50,000$



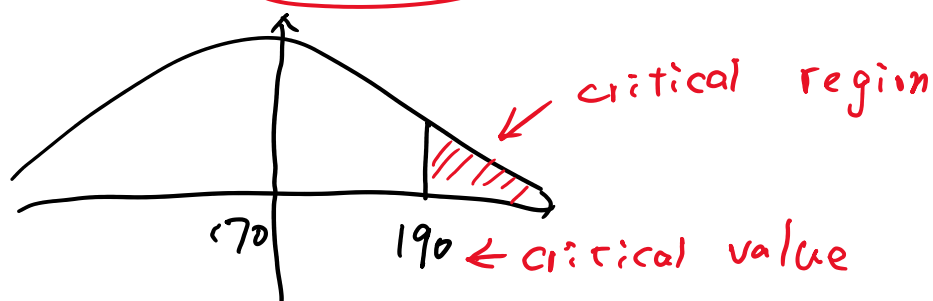
when

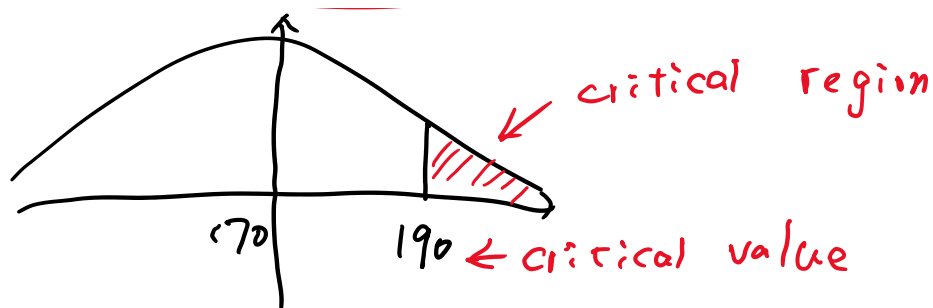
$\bar{X} = 50,000 < 80,000$

The critical region consists of values of the test statistic that resulting in rejecting the null hypothesis.

$H_0: \mu = 170$

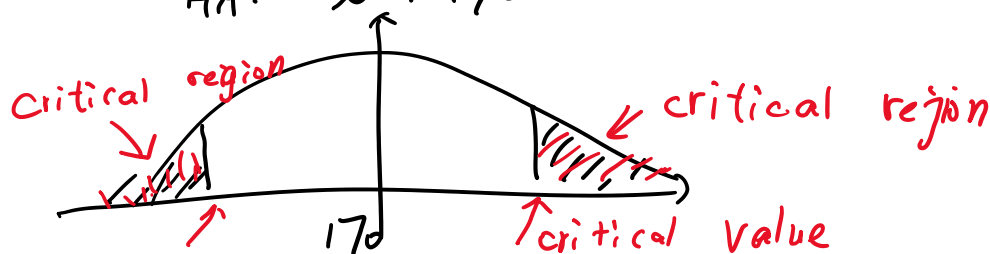
$H_A: \mu > 170$





$$H_0: \mu = 170$$

$$H_A: \mu \neq 170$$



4. Make a conclusion.

When Test statistic lies in the critical region, then reject the null hypothesis.

Otherwise, do not.

Type of errors:

	H_0 is True	H_A is True.
Reject H_0	Type I error	Correct
Accept H_0	Correct	Type II error

α : A type I error occurs if we reject H_0 but H_0 is actually true. (Patient with disease, fail to diagnosis)

β : A type II error occurs if we do not reject H_0 when H_A is actually true (Healthy people diagnosed wrongly)

α : type I error, significance level or size of the test

β = type II error.

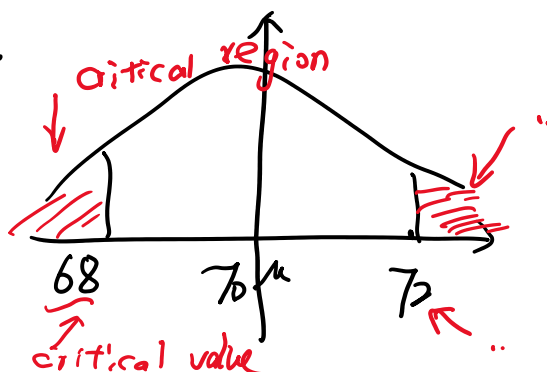
$1 - \beta$: power of the test

Example: Sample $n = 64$ from a normal population.

$H_0: \mu = 70$

$H_A: \mu \neq 70$

Suppose we reject H_0 if $\bar{X} \geq 72$ or $\bar{X} \leq 68$. Find the α , assuming $\sigma = 16$.



α = Type I error

$= P(\text{rejecting } H_0 \mid \text{H}_0 \text{ is actually true})$
 $\mu = 70$

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

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

$= P\left(Z \leq \frac{68 - 70}{\frac{16}{\sqrt{64}}}\right) + P\left(Z \geq \frac{72 - 70}{\frac{16}{\sqrt{64}}}\right)$
 σ $\mu = 70$

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

$= 2P(Z \leq -1)$

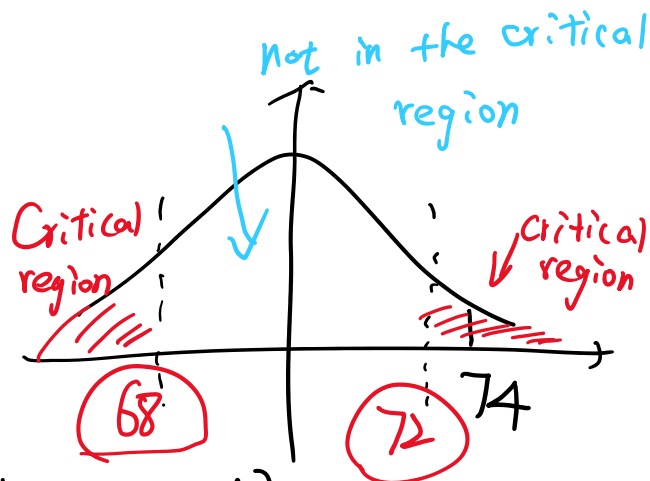
$= 2 \cdot 0.1587$

β , additional condition

true mean : $\mu = 74$

β = type II error

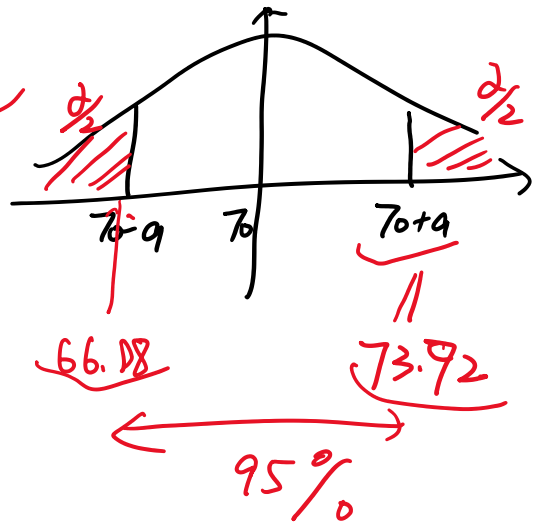
$= P(\text{fail to reject } H_0 \mid H_A \text{ is correct})$



$$\begin{aligned}
 &= P(\text{fail to reject } H_0 \mid H_A \text{ is correct}) \\
 &= P(68 \leq \bar{X} \leq 72 \mid \mu = 74) \\
 &= P\left(\frac{68-74}{16/\sqrt{64}} \leq Z \leq \frac{72-74}{16/\sqrt{64}}\right) \\
 &= P(-3 \leq Z \leq -1) \\
 &= 0.9986 - 0.8413 \\
 &= 0.1573
 \end{aligned}$$

Example: Find critical value, $\alpha = 5\%$

$$\begin{aligned}
 &\therefore P(\bar{X} > 70+a) + P(\bar{X} \leq 70-a) \\
 &= P\left(Z \geq \frac{70+a-70}{16/\sqrt{64}}\right) + P\left(Z \leq \frac{70-a-70}{16/\sqrt{64}}\right) \\
 &= P\left(Z \geq \frac{a}{2}\right) + P\left(Z \leq \frac{-a}{2}\right) \\
 &= 5\% = \alpha \\
 &= 2 \cdot P\left(Z \leq \frac{-a}{2}\right) \\
 &\therefore P\left(Z \leq \frac{-a}{2}\right) = 0.025
 \end{aligned}$$



$$\therefore \frac{a}{2} = 1.96 \Rightarrow a = 3.92$$

In general: (Relationship between C.I and Hypothesis Test)

When testing $H_0: \mu = \mu_0$ vs $H_A: \mu \neq \mu_0$

we reject H_0 at α (significance level)

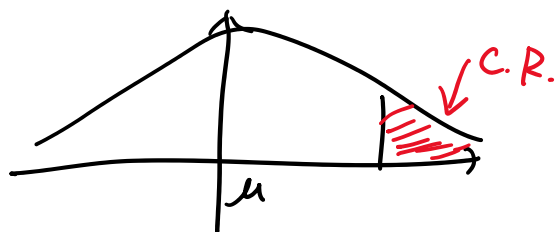
$\Leftrightarrow$ the $100(1-\alpha)\%$ C.I for average does not contain the true mean μ .

Comment: α is fixed, $\alpha \uparrow \Rightarrow \beta \downarrow$

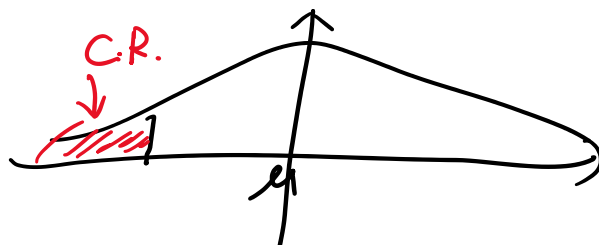
Comment: 1. n is fixed, $\alpha \uparrow \Rightarrow \beta \downarrow$
 $\beta \uparrow \Rightarrow \alpha \downarrow$
 2. Increasing n , reduce α, β .

One-sided Hypothesis Test:

1. $\begin{cases} H_0: \mu = 70 \\ H_A: \mu > 70 \end{cases} \Rightarrow$



2. $\begin{cases} H_0: \mu = 70 \\ H_A: \mu < 70 \end{cases}$



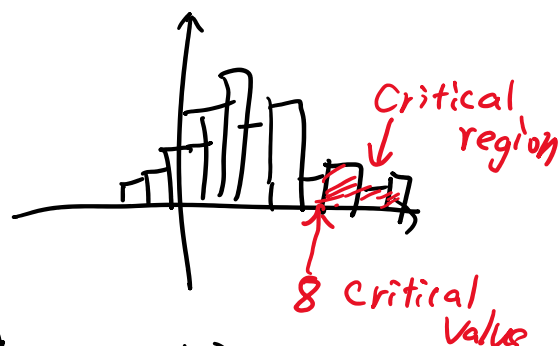
Example: Suppose that
 Test: A coin is biased and having larger chance for head.
 Flip it 10 times and if find 8 heads or more
 then we claim is biased.

$$\begin{cases} H_0: p = \frac{1}{2} \\ H_A: p > \frac{1}{2} \end{cases}$$

Binomial ($n=10, p=\frac{1}{2}$)
 $\underbrace{\hspace{10em}}_{H_0 \text{ is true}}$

(1) Find α .

$$\begin{aligned} \alpha &= P(\text{reject } H_0 \mid \underline{H_0 \text{ true}}) \\ &= P(X \geq 8 \mid p = \frac{1}{2}) \end{aligned}$$



$$\begin{aligned} &= \left(\binom{10}{8} \left(\frac{1}{2}\right)^8 \left(\frac{1}{2}\right)^2 + \binom{10}{9} \left(\frac{1}{2}\right)^9 \left(\frac{1}{2}\right) + \binom{10}{10} \left(\frac{1}{2}\right)^{10} \right) \\ &= 0.0547 \end{aligned}$$

(2) Find β if $p = \frac{2}{3}$

$$\beta = P(\text{not reject } H_0 \mid \underline{H_A \text{ is true}})$$

$$\beta = P(\text{not reject } H_0 \mid \underline{H_A \text{ is true}})$$

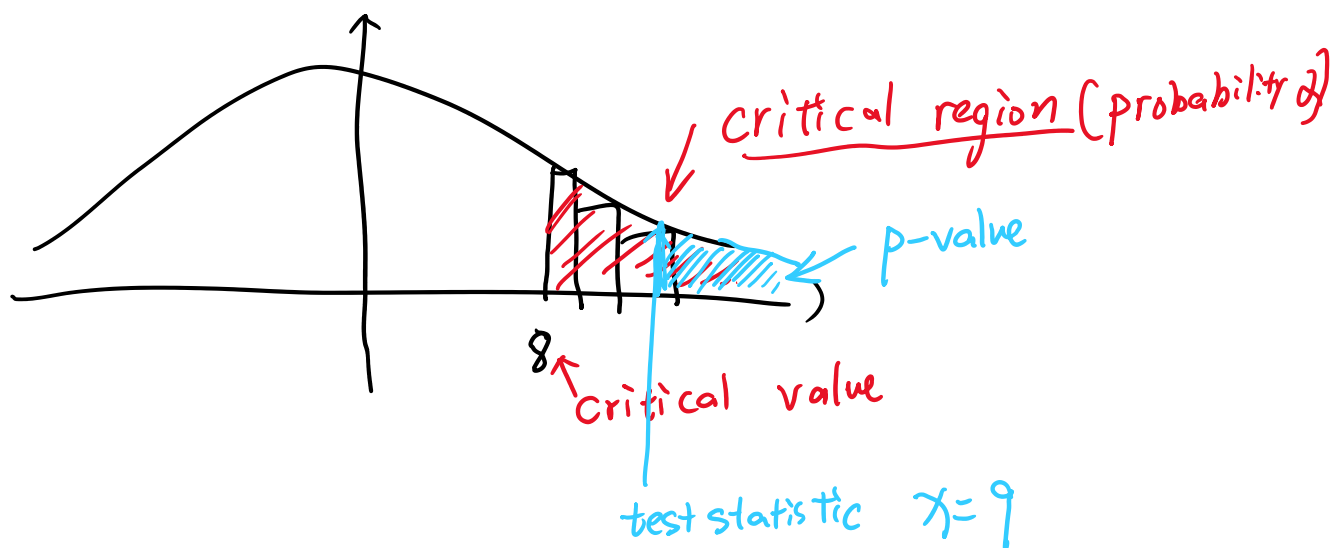
$p = \frac{2}{3}$

$$= P(X \leq 7 \mid p = \frac{2}{3})$$

$$= 1 - \left[\binom{10}{8} \left(\frac{2}{3}\right)^8 \left(\frac{1}{3}\right)^2 + \binom{10}{9} \left(\frac{2}{3}\right)^9 \left(\frac{1}{3}\right) + \binom{10}{10} \left(\frac{2}{3}\right)^{10} \left(\frac{1}{3}\right)^0 \right]$$

$$= 0.7009$$

p-value is the probability that obtain a value of test statistic at least as extreme as the observed value when H_0 is true.



$$\therefore \begin{aligned} p\text{-value} < \alpha &\Rightarrow \text{reject } H_0 \\ p\text{-value} > \alpha &\Rightarrow \text{do not reject } H_0 \end{aligned}$$

(c) Suppose we have an observation $X=6$.

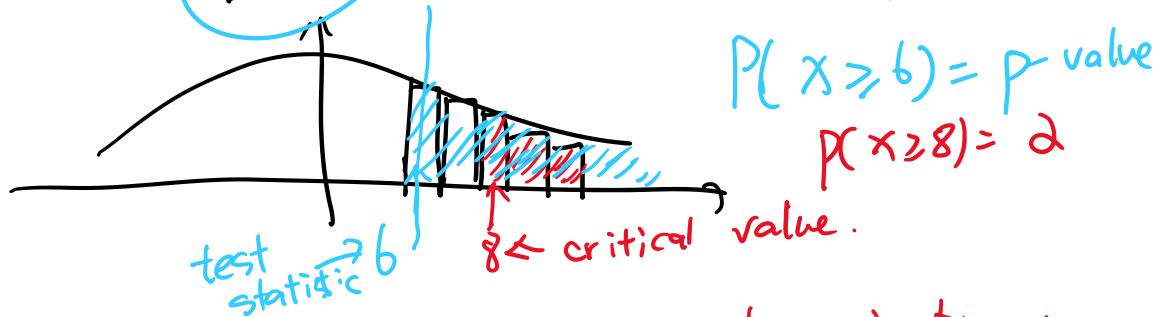
Find the p-value.

$$\therefore \text{p-value} = P(\underbrace{X=6 \text{ or more}}_{\text{tail probability}} \mid \underbrace{p=\frac{1}{2}}_{H_0 \text{ is true}})$$

$$= \binom{10}{6} \left(\frac{1}{2}\right)^6 \left(\frac{1}{2}\right)^4 + \binom{10}{7} \left(\frac{1}{2}\right)^7 \left(\frac{1}{2}\right)^3 + \binom{10}{8} \left(\frac{1}{2}\right)^8 \left(\frac{1}{2}\right)^2 + \binom{10}{9} \left(\frac{1}{2}\right)^9 \left(\frac{1}{2}\right)^1 + \binom{10}{10} \left(\frac{1}{2}\right)^{10} \left(\frac{1}{2}\right)^0$$

$$= \binom{10}{6} \left(\frac{1}{2}\right)^6 \cdot \left(\frac{1}{2}\right)^4 + \dots + \binom{10}{10} \left(\frac{1}{2}\right)^{10} \left(\frac{1}{2}\right)^0$$

$\alpha > \alpha \Rightarrow$ do not reject H_0



1. Tests on the mean of normal distribution. σ is known.

Sample: $X_1, \dots, X_n$ with mean μ and variance σ^2 .

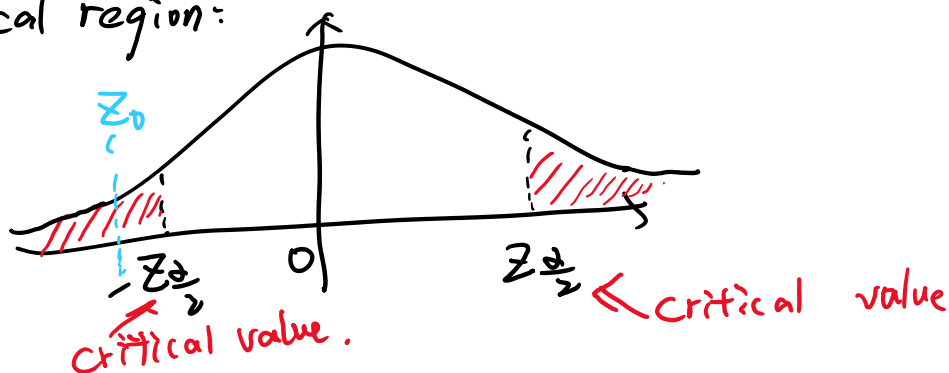
Then, we want to test

$$\begin{cases} H_0: \mu = \mu_0 \\ H_A: \mu \neq \mu_0 \end{cases}$$

Test statistic:

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

Critical region:



Then the p-value:

$$= 2 P(Z < z_0) \text{ then compare with } \alpha.$$

of reject H_0 | H_0 is true)

$$\alpha = P(\text{reject } H_0 \mid H_0 \text{ is true})$$

$\mu = \mu_0$

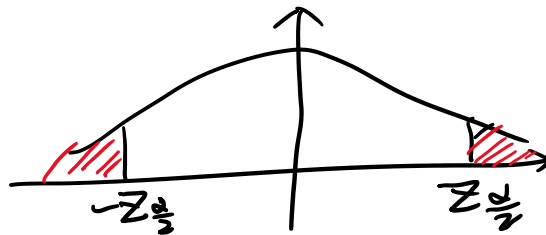
Conclusion:

Reject H_0 , then test is significant,
we could conclude that H_A is correct.

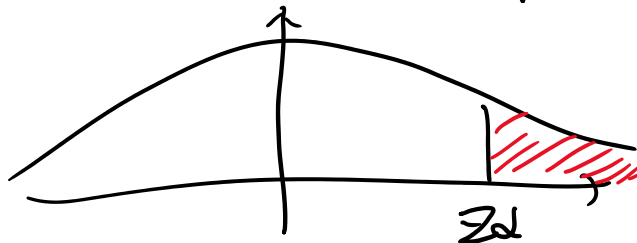
Not reject H_0 , the test is not conclusive,
we can not prove H_0 is wrong based on the sample.

When to reject?

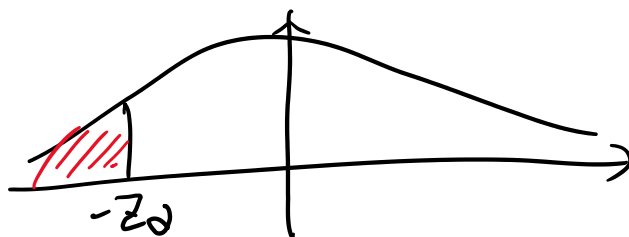
1. $H_1: \mu \neq \mu_0$, reject H_0 at α if $Z > Z_{\frac{\alpha}{2}}$ or $Z < -Z_{\frac{\alpha}{2}}$



2. $H_1: \mu > \mu_0$, reject H_0 at α if $Z > Z_\alpha$



3. $H_1: \mu < \mu_0$, reject H_0 at α if $Z < -Z_\alpha$



or when p-value is less α , reject H_0
 $p\text{-value} < \alpha$

Example:

$\mu = 98.6$ to test

$n = 100$, $\bar{x} = 98.2$ then to test whether the true mean

$n=100$, $\bar{x} = 98.2$ then to test whether the true mean is 98.6 or not. σ is given as 0.62. ($\alpha=0.01$)

1. Hypothesis: $\begin{cases} H_0 = \mu = 98.6 \\ H_A = \mu \neq 98.6 \end{cases}$

2. Test statistic:

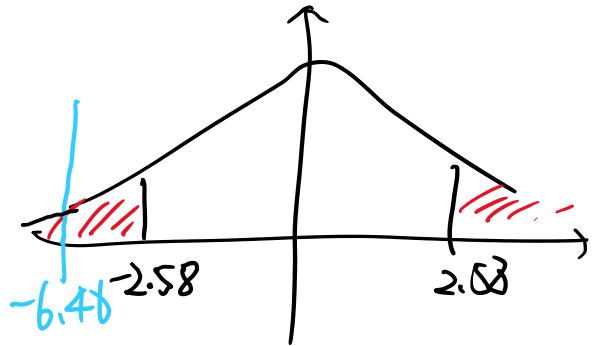
$$Z = \frac{98.2 - 98.6}{0.62 / \sqrt{106}} = -6.64.$$

3. Critical value: $+$

two tail, $\alpha = 0.01$

$$Z_{\frac{\alpha}{2}} = Z_{0.005} = 2.58$$

$$-Z_{\frac{\alpha}{2}} = -Z_{0.005} = -2.58$$



4. $\textcircled{2}$ P-value = $P(\bar{Z} < -6.64) \cdot 2 < 0.01$
Conclusion:

4. Conclusión:

As $Z < -Z_{\frac{\alpha}{2}}$, then we reject the H_0 and claim that the true mean may not be 98.6.

Type II error : β :

$\mu = \mu_0 + \delta$

true mean mean in H_0

$\delta \rightarrow \text{delta}$
 $\sigma \rightarrow \text{sigma}$

difference

$$\beta = P\left(Z \leq Z_{\frac{\alpha}{2}} - \frac{8\sqrt{n}}{\sigma}\right) - P\left(Z \leq -Z_{\frac{\alpha}{2}} - \frac{8\sqrt{n}}{\sigma}\right)$$

From CLT.

Choice of sample size:

$$(Z_A + Z_B)^2 \leq 2 \quad , \quad C = 1, 1, 1, 1$$

$$n = \frac{(Z_{\frac{\alpha}{2}} + Z_{\beta})^2}{\sigma^2} \sigma^2 \text{ where } \delta = \mu - \mu_0$$

For given α, β .

2. Test on the mean of Normal when Variance Unknown

μ σ unknown

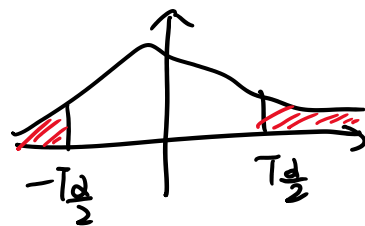
1. Hypothesis: $H_0: \mu = \mu_0$ $H_A: \mu \neq \mu_0$

2. Test Statistic:

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

3. Critical region

$$-T_{\frac{\alpha}{2}}, T_{\frac{\alpha}{2}}$$



4. Conclusion.

1. reject if $p\text{-value} < \alpha$

2. T (test statistic) in the critical region.

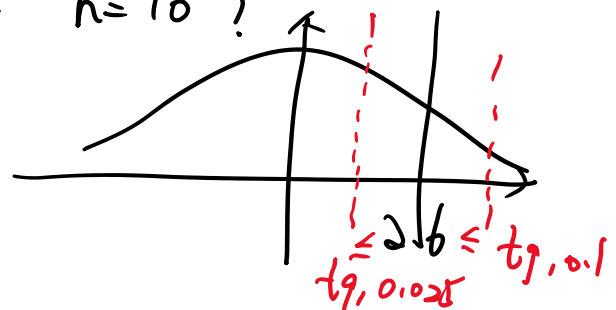
T table can not help you find p -value exactly.

Example: $T = 2.6$ (test statistic) one tail ($\mu > \mu_0$)
 what's p -value if $n = 10$?

$$t_{9, 0.1} = 2.821$$

$$t_{9, 0.025} = 2.237$$

$$0.025 < p\text{-value} < 0.1$$



... in approximations:

$$0.025 - p \dots \dots \dots$$

3. Tests on population proportions:

1 Hypothesis: $H_0: p = p_0$

$H_A: p \neq p_0$

2. Test statistic

$$Z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}} \quad \text{where } \hat{p} = \frac{x}{n}$$

3. Critical region. (Similar to μ with σ known)

4. Conclusion.

Type II Error: ① When $p < p_0$

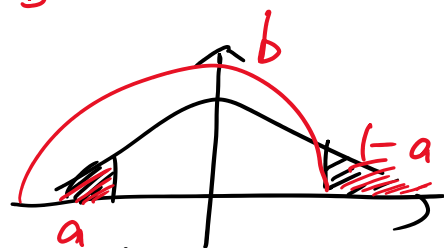
$$\beta = 1 - P\left(Z \leq \frac{p_0 - p - Z_{\alpha} \sqrt{p_0(1-p_0)/n}}{\sqrt{p_0(1-p_0)/n}}\right)$$

② When $H_1: p > p_0$

$$\beta = P\left(Z \leq \frac{p_0 - p + Z_{\alpha} \sqrt{p_0(1-p_0)/n}}{\sqrt{p_0(1-p_0)/n}}\right)$$

③ $H_1: p \neq p_0$

$$\beta = b - a$$



Choice of sample size. (α, β given)

$$n = \left[\frac{Z_{\alpha} \sqrt{p_0(1-p_0)} + Z_{\beta} \sqrt{p_0(1-p_0)}}{p - p_0} \right]^2 \quad (\text{one tail})$$

Chapter 10: Difference between means from Normal population

Chapter 10: Difference between two population

4. Tests on μ_1 and μ_2 when assuming equal variance ($\sigma_1^2 = \sigma_2^2$) assuming σ_1, σ_2 is known

1. Hypothesis: $H_0: \mu_1 = \mu_2$
 $H_A: \begin{cases} \mu_1 \neq \mu_2 \\ \mu_1 > \mu_2 \\ \mu_1 < \mu_2 \end{cases}$

2. Test statistic:

$$Z = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{Sp^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim N(0, 1)$$

where $Sp^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}$

$\left(\begin{array}{cc} \frac{S_1^2}{\uparrow} & \frac{S_2^2}{\uparrow} \\ \sigma_1 & \sigma_2 \end{array} \right)$ $\begin{matrix} \swarrow SD(X_1) \\ \searrow SD(X_2) \end{matrix}$

$\swarrow$ pooled sample variance.

3. Critical Region

4. Conclusion.

5. Tests on μ_1 and μ_2 assuming σ unknown but equal variance ($\sigma_1^2 = \sigma_2^2$)

Test statistic:

$$T = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{Sp^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim t(\nu)$$

$\nwarrow$ pivot

$$\nu = n_1 + n_2 - 2$$

C.I for $(\mu_1 - \mu_2)$

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

6. Tests on μ_1 and μ_2 assuming variance unknown but unequal variance.

Test statistic:

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

$$\sim t(\nu)$$

Degree of freedom:

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

C.I:

$$\bar{x}_1 - \bar{x}_2 \pm t_{\nu, \alpha/2} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

Example:

Two supplies manufacture a plastic gear used in a laser printer.

Supplier 1 had a mean impact $\bar{x}_1 = 290$

$$s_1 = 12, \quad n_1 = 10.$$

$$\text{Supplier 2: } \bar{x}_2 = 321, \quad s_2 = 22, \quad n_2 = 16.$$

Assume that population variances are unknown and not equal. $\alpha = 0.05$.

Assume that $\sigma_1 \neq \sigma_2$
not equal. $\alpha = 0.05$.

Test the hypothesis that two suppliers have difference
in mean impact. two tail

1. Hypothesis:

$$H_0: \mu_1 = \mu_2$$

$$H_A: \mu_1 \neq \mu_2$$

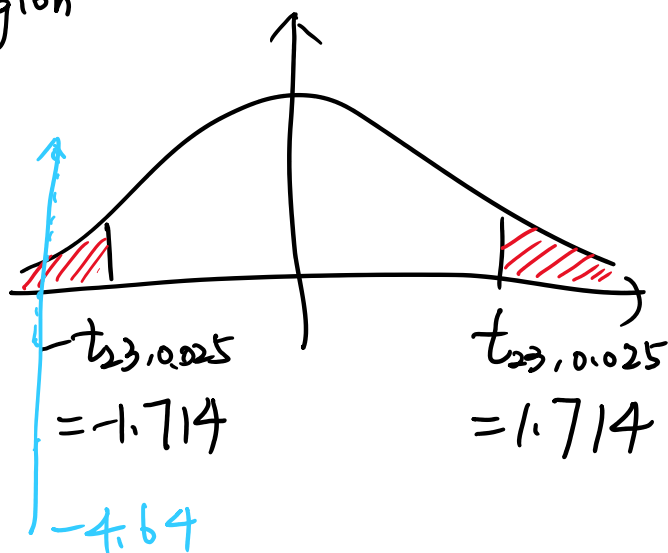
2. Test statistic:

$$T = \frac{290 - 321}{\sqrt{\frac{12^2}{10} + \frac{22^2}{16}}} = -4.64 \sim t(2)$$

$\sim t(23)$

$$df = \frac{(\frac{12^2}{10} + \frac{22^2}{16})^2}{\frac{(12^2/10)^2}{10-1} + \frac{(22^2/16)^2}{16-1}} = \boxed{23.72} \approx \underline{23}$$

3. Critical Region



$$\therefore T = -4.64 < -t_{23, 0.025} = -1.714$$

$\Rightarrow$ reject H_0 .

4. Conclusion: the test is significant and we could conclude that they have different mean of impact.

C.I. 95%

$$\downarrow \underline{t_{23, 0.025} \text{ (two tail } \alpha = 5\%)}$$

12 22

C.I . 95 %

$$(290 - 321) \pm 1.714 \sqrt{\frac{12^2}{10} + \frac{22^2}{16}}$$

0.25.0023 ... 10%

$$= (-42.45, -19.55)$$