

Lecture 10 :

$$x_1, \dots, x_n \sim N(\mu, \sigma^2)$$

$$\begin{cases} H_0: \mu = \mu_0 \\ H_a: \mu > \mu_0 \end{cases}$$

- step 1: (to design T)

$$T(x_1, \dots, x_n) = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$

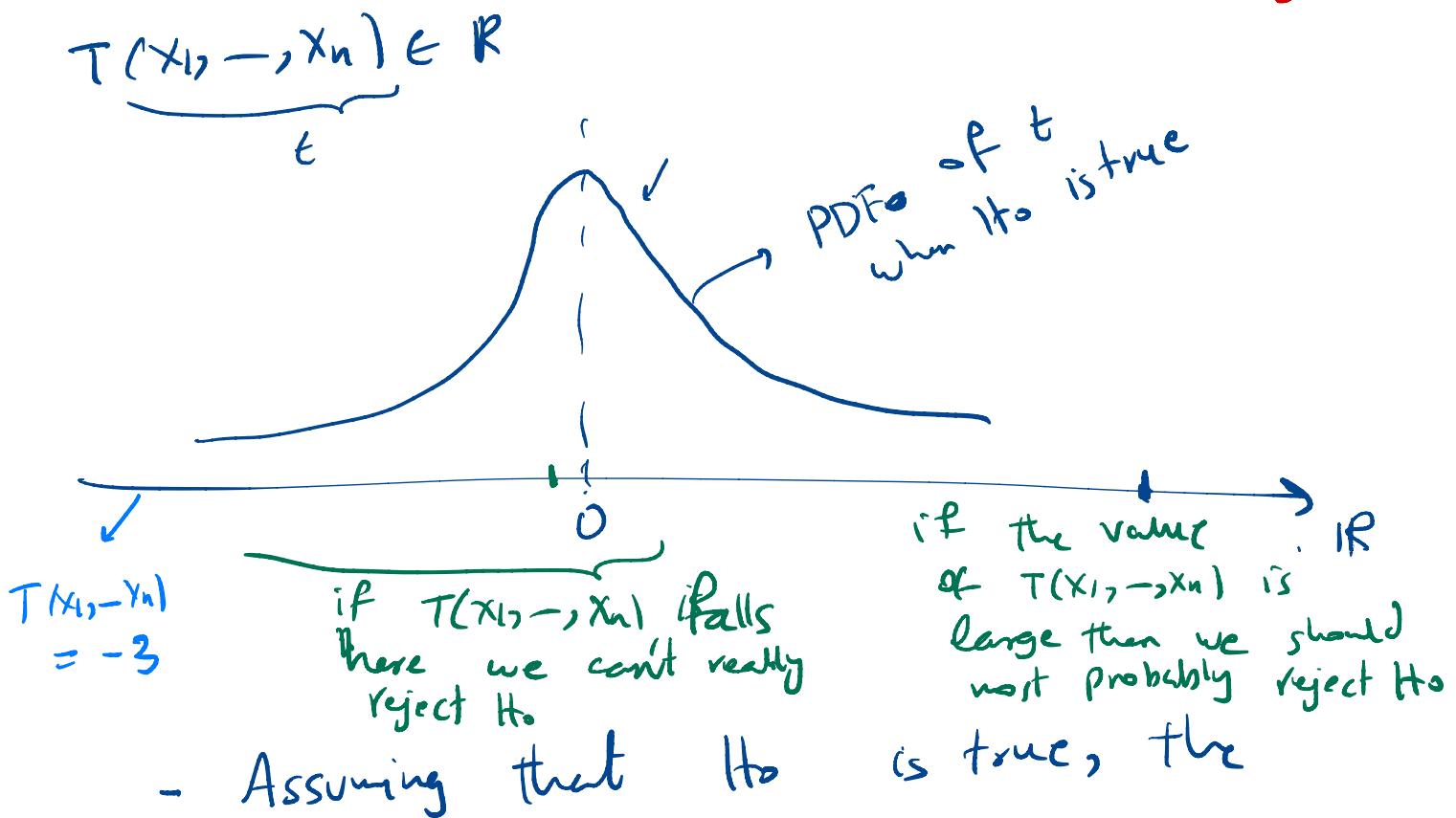
- Assuming that H_0 is true, the distribution of the outcome of T

is:

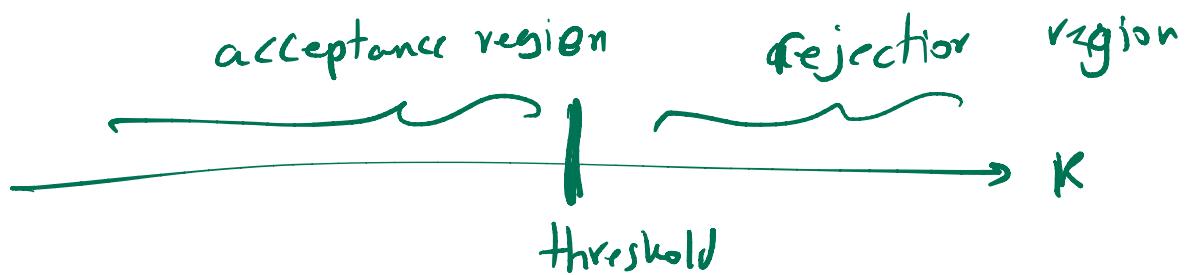
$$T(x_1, \dots, x_n) = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1)$$

- step 2: (identify acceptance/reject regions)

Constraint: $\alpha = \Pr \{ H_0 \text{ is rejected} \mid H_0 \text{ is true} \}$
 \leq significance level
 (α_0)
 $\underline{\alpha_0}$ given



$$T(X_1, \dots, X_n) \sim N(0, 1)$$

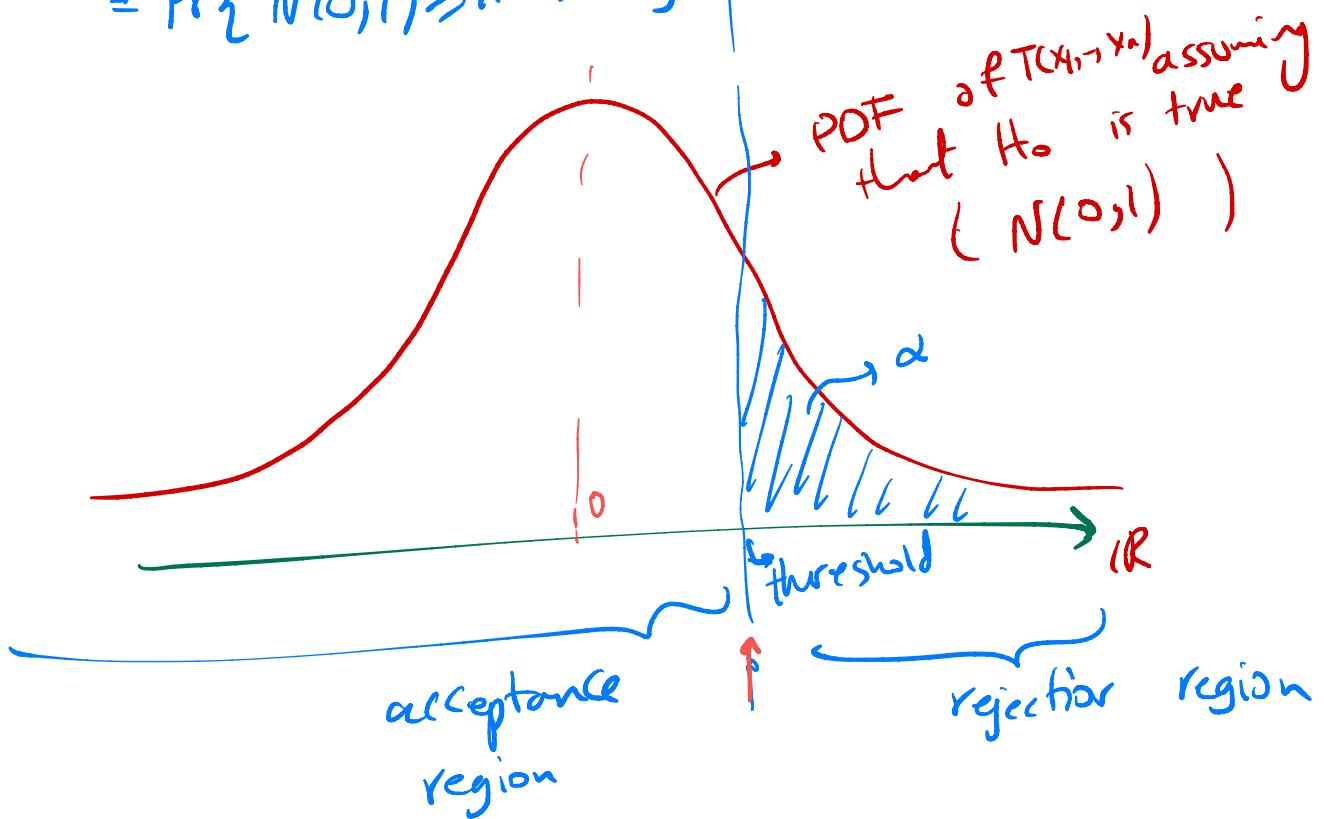


How should we find the threshold?

→ we still need to make sure that

$$\alpha \leq \alpha_0 \quad \text{e.g. } 0.01$$

$$\begin{aligned}\alpha &= \Pr \left\{ H_0 \text{ is rejected} \mid H_0 \text{ is true} \right\} \\ &= \Pr \left\{ T(x_1, \dots, x_n) \in \text{rejection region} \mid T(x_1, \dots, x_n) \sim N(0, 1) \right\} \\ &= \Pr \left\{ N(0, 1) > \text{threshold} \right\}\end{aligned}$$



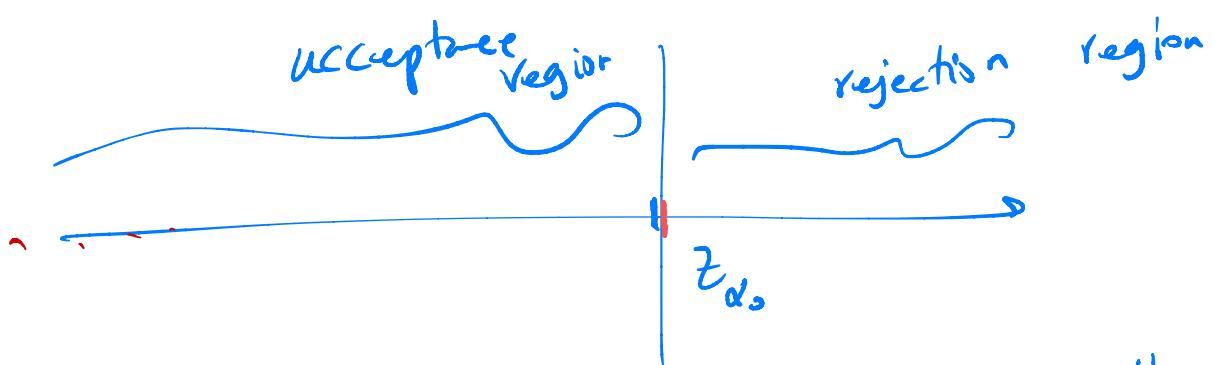
We want α to be less than α_0 .

$$\alpha = \alpha_0 \iff \text{threshold} = z_{\alpha_0}$$

Hypothesis testing problem

$$\begin{cases} H_0: \mu = \mu_0 \\ H_a: \mu > \mu_0 \end{cases}$$

$$\rightarrow T(\bar{X}_1, \dots, \bar{X}_n) = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}}$$



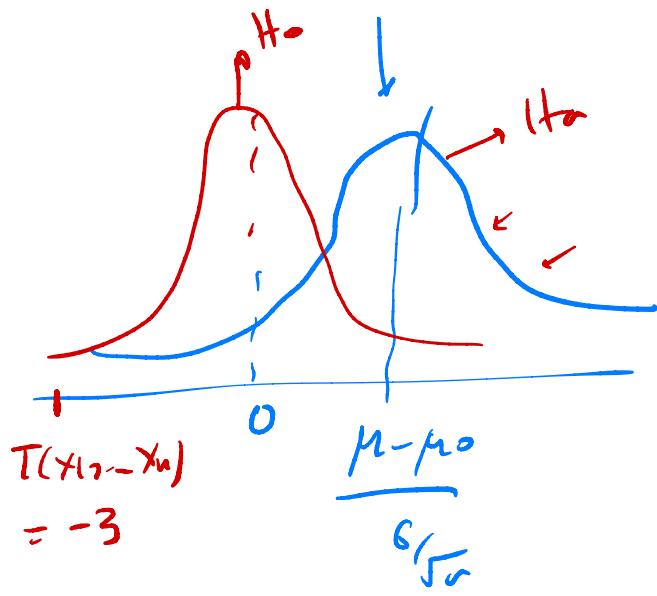
$$T(\bar{X}_1, \dots, \bar{X}_n) \geq z_{\alpha_0}$$

reject H_0

do not
reject H_0

if H_a was true $\rightarrow \mu > \mu_0$

$$T(\bar{X}_1, \dots, \bar{X}_n) = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}} = \underbrace{\frac{\mu - \mu_0}{\sigma / \sqrt{n}}}_{> 0} + N(0, 1)$$



Let's compute β for the example above.

$$\beta = \Pr \{ \text{type II error} \}$$

$$= \Pr \left\{ \begin{array}{l} H_0 \text{ is accepted} \\ H_0 \text{ is false} \end{array} \right\}$$

$$\left\{ \begin{array}{l} H_0: \mu = \mu_0 \\ H_0: \mu > \mu_0 \end{array} \right.$$

$$\left\{ \begin{array}{l} T(X_1, \dots, X_n) = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} \\ T(X_1, \dots, X_n) \stackrel{\substack{\text{reject} \\ \downarrow \\ \text{accept}}}{\geq} Z_{\alpha} \end{array} \right.$$

Assuming that H_0 is true, $X_1, \dots, X_n \sim N(\mu, \sigma^2)$

Assuming $X_1, \dots, X_n \sim N(\mu, \sigma^2) \rightarrow \bar{X} = \frac{\sum_{i=1}^n X_i}{n} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$

$\mu > \mu_0$

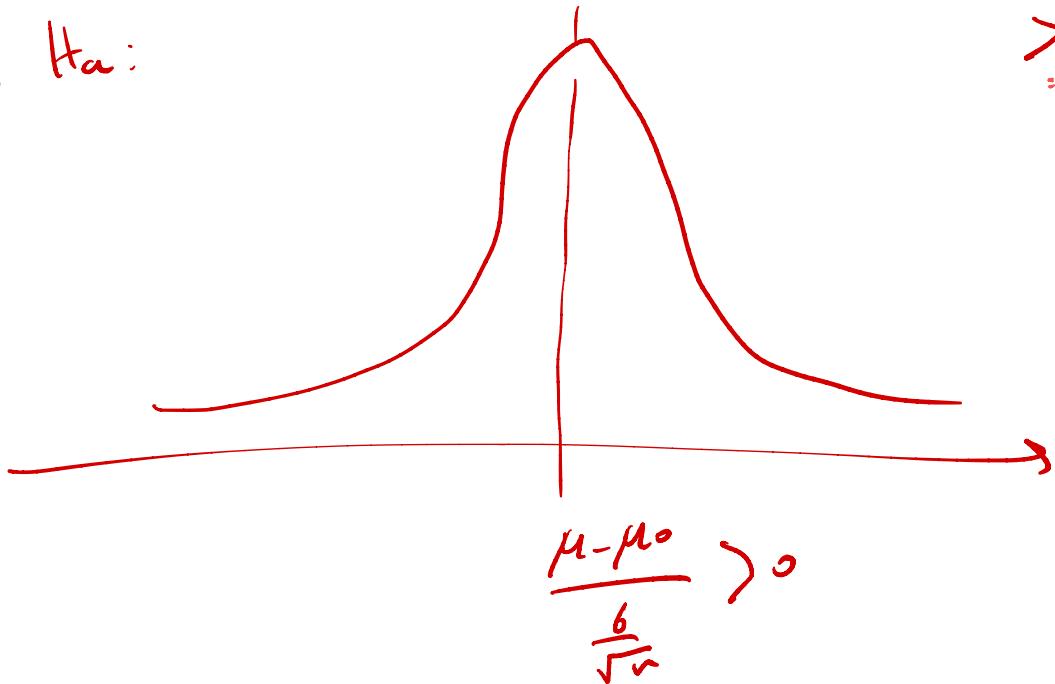
$$\Rightarrow \bar{X} = \mu + \frac{\mu - \mu_0}{\sqrt{n}} + N(0, 1)$$

recall: $T(X_1, \dots, X_n) = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$

$$\Rightarrow T(X_1, \dots, X_n) \sim \frac{\mu - \mu_0}{\frac{\sigma}{\sqrt{n}}} + N(0, 1)$$

$\underbrace{\quad}_{> 0}$

Assuming $H_a:$



$$\beta = \Pr \left\{ H_0 \text{ is accepted} \middle| \begin{array}{l} H_a \text{ is true} \\ \text{if } H_0 \text{ is false} \end{array} \right\}$$

$$= \Pr \left\{ T(X_1, \dots, X_n) \leq z_{\alpha_0} \middle| \begin{array}{l} T(X_1, \dots, X_n) \sim \\ \frac{\mu - \mu_0}{\frac{\sigma}{\sqrt{n}}} + N(0, 1) \end{array} \right\}$$

$$\begin{aligned}
 - &= \Pr \left\{ \frac{\mu - \mu_0}{\sigma/\sqrt{n}} + N(0,1) \leq z_{d_0} \right\} \\
 &= \Pr \left\{ N(0,1) \leq z_{d_0} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}} \right\} \\
 &= \Phi(z_{d_0} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}})
 \end{aligned}$$

$\text{area} = \beta$
 $z_{d_0} = \frac{\mu - \mu_0}{\sigma/\sqrt{n}}$ $z_{d_0} = b_{\text{threshold}}$

Notation:

$\text{area} = \varphi(u)$
 $\int_{-\infty}^u \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = \varphi(u)$

gaussian error function
 ↳ CDF of normal dist.

Example 2:

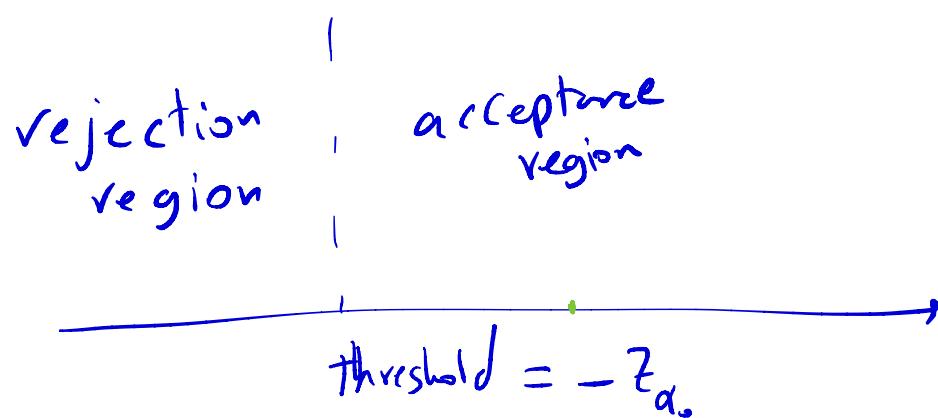
$$X_1, \dots, X_n \sim N(\mu, \sigma^2)$$

$$\begin{cases} H_0: \mu = \mu_0 \\ H_a: \mu < \mu_0 \end{cases}$$

Similar to the previous example

we can take

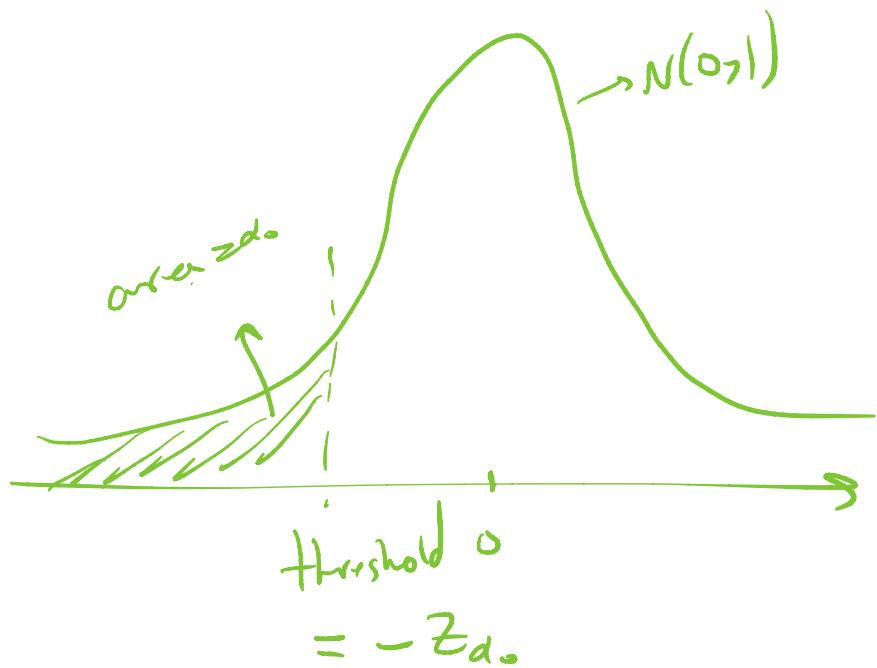
$$T(X_1, \dots, X_n) = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$



$$\Pr \{ \text{type I error} \mid H_0 \text{ is true} \}$$

$$= \Pr \{ T(X_1, \dots, X_n) \leq \text{threshold} \mid H_0 \text{ is true} \}$$

$$= \Pr \{ N(0, 1) \leq \text{threshold} \} = \alpha$$



$$T(x_1, \dots, x_n) = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}} \begin{cases} < -z_{\alpha/2} & \text{accept } H_0 \\ > z_{\alpha/2} & \text{reject } H_0 \end{cases}$$

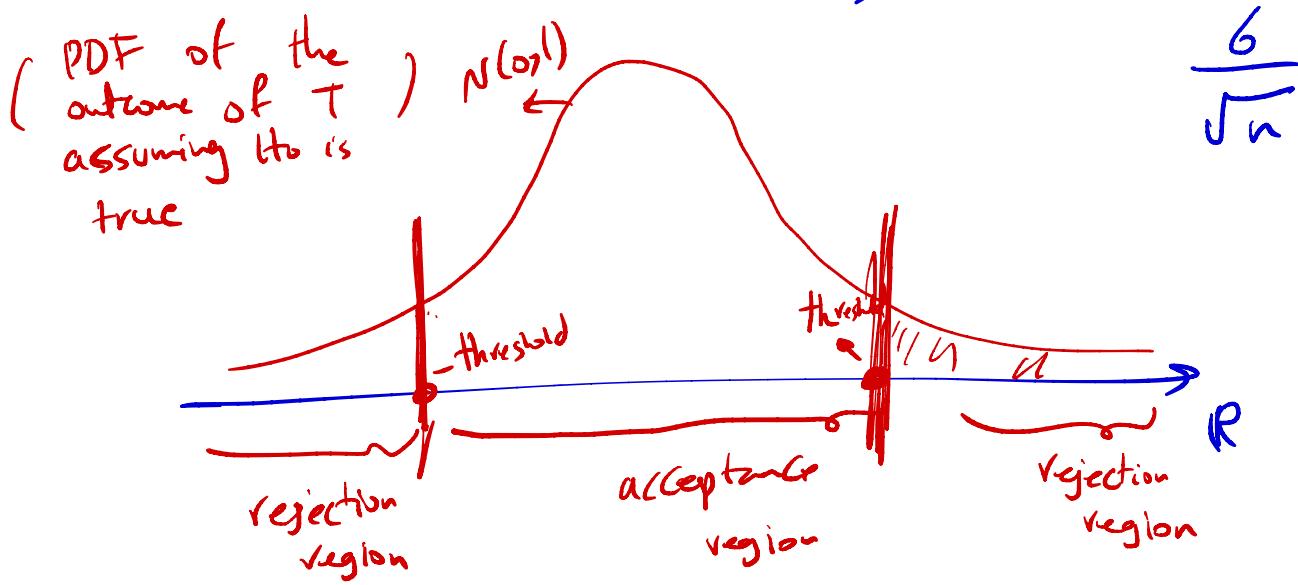
Example 3:

$$X_1, \dots, X_n \sim N(\mu, \sigma^2)$$

$$\left. \begin{array}{l} H_0: \mu = \mu_0 \\ H_a: \mu \neq \mu_0 \end{array} \right\}$$

Similar to the previous examples, since we're testing the mean, a good statistic can be:

$$T(X_1, \dots, X_n) = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$

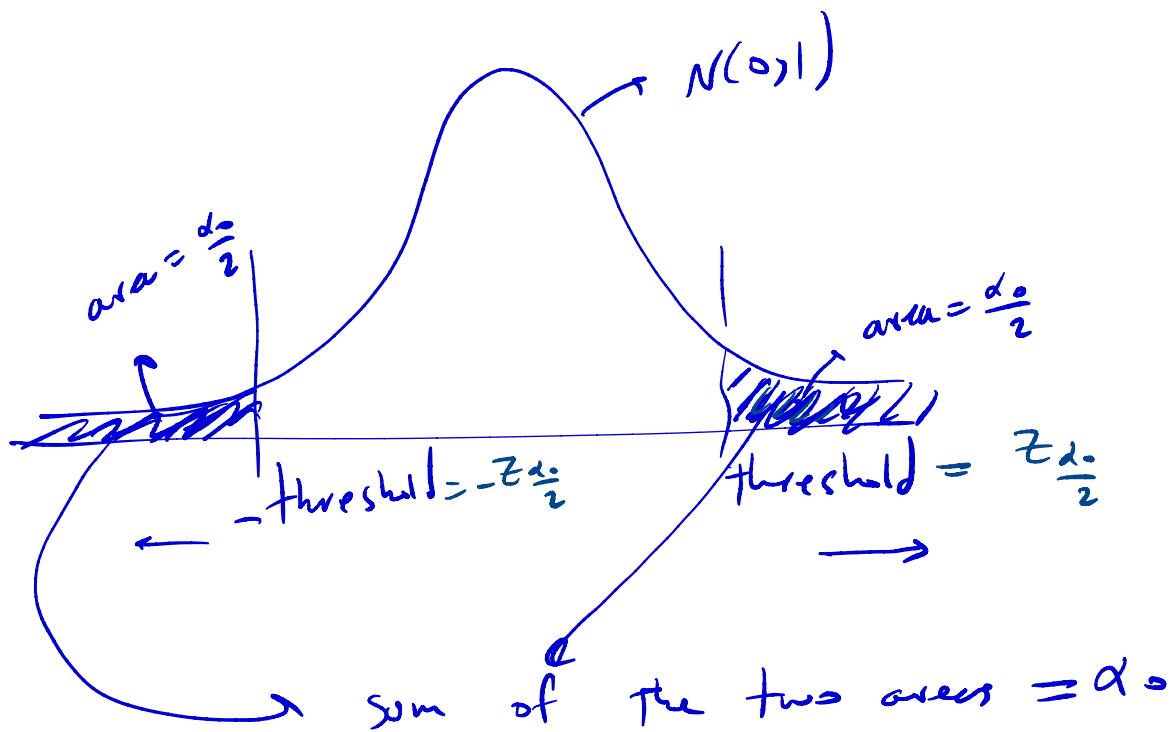


In order to find the threshold,
we need use the constraint and
make sure it's satisfied:

$$\Pr \{ \text{type I error} \} = \alpha_0$$

$$\Pr \left\{ \begin{array}{l} \text{or } T(x_1, \dots, x_n) \geq \text{threshold} \\ \text{or } T(x_1, \dots, x_n) \leq -\text{threshold} \end{array} \middle| \begin{array}{l} H_0 \text{ is true} \end{array} \right\} = \alpha_0$$

$$= \Pr \left\{ \begin{array}{l} N(0, 1) \geq \text{threshold} \\ \text{or} \\ N(0, 1) \leq -\text{threshold} \end{array} \right\}$$



$$\Rightarrow \left| T(x_1, \dots, x_n) \right| \begin{cases} > \frac{\alpha}{2} \\ < \frac{\alpha}{2} \\ \text{accept } H_0 \end{cases} \begin{cases} \text{reject } H_0 \\ Z \frac{\alpha}{2} \end{cases}$$

A more general setting (^{large n} _{known}):

$$x_1, \dots, x_n \sim \text{dist}(\mu, \sigma^2)$$

$$\left\{ \begin{array}{l} H_0 : \mu = \mu_0 \\ H_a : \mu > \mu_0 \end{array} \right. \quad \left(\begin{array}{l} \text{remember} \\ \text{when } n \text{ is} \\ \text{large, by CLT,} \\ \text{we have} \\ \bar{x} \sim N(\mu, \sigma^2/n) \end{array} \right)$$

$$T(x_1, \dots, x_n) = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$

since \bar{x} is (approximately, for large n) distributed according to $N(\mu, \sigma^2/n)$, all the derivations follow exactly as the previous examples.

$$T(X_1, \dots, X_n) = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}} \begin{cases} \text{reject} \\ \text{accp} \end{cases} z_{\alpha}$$

Similar cases follow for the other examples of hypothesis testing that we considered above.

Finally, so far we have assumed that σ^2 is known.

$$T(X_1, \dots, X_n) = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}}$$

What should we do if σ is not known? Use $\hat{\sigma}$ instead.

$$\hat{\sigma} = \sqrt{\frac{1}{n-1} \left(\sum_{i=1}^n (x_i - \bar{x})^2 \right)}$$