

Hypothesis testing framework

Recap: Constructing a test

$$H_0 : \theta \in \Theta_0 \quad H_A : \theta \in \Theta_1$$

Observe data X_1, \dots, X_n .

- + Calculate a test statistic $T_n = T(X_1, \dots, X_n)$
- + Choose a rejection region $R = \{(x_1, \dots, x_n) : \text{reject } H_0\}$
- + Reject H_0 if $(X_1, \dots, X_n) \in R$

Goal: minimize probability of a type II error
while controlling probability of a type I error

$$P(\text{type I error}) = P(\text{reject } H_0 \mid H_0 \text{ is true})$$

$$P(\text{type II error}) = 1 - P(\text{reject } H_0 \mid H_A \text{ is true})$$

More specifically: we want $P(\text{reject } H_0 \mid \text{true parameter} = \theta)$

Power function

Suppose we reject H_0 when $(X_1, \dots, X_n) \in R$. The **power function** $\beta(\theta)$ is

$$\beta(\theta) = P_\theta((X_1, \dots, X_n) \in R)$$

function of θ probability of rejecting H_0 if the parameter is θ

Goal: want $\beta(\theta)$ small when $\theta \in H_0$
and large when $\theta \in H_1$,

Formally:

- ① Fix $\alpha \in [0, 1]$
- ② Try to maximize $\beta(\theta)$ for $\theta \in H_1$, subject to
 $\beta(\theta) \leq \alpha$ for $\theta \in H_0$

if $\sup_{\theta \in H_0} \beta(\theta) = \alpha$, our test is size α
if $\sup_{\theta \in H_0} \beta(\theta) \leq \alpha$, our test is level α

Φ = denote $N(0,1)$ cdf

Example

X_1, \dots, X_n iid from a population with mean μ and variance σ^2 .

$$H_0 : \mu = \mu_0 \quad H_A : \mu > \mu_0$$

reject H_0 when $Z_n = \frac{\bar{X}_n - \mu_0}{\sigma/\sqrt{n}} > c$

Power function:

$$\begin{aligned} \beta(\mu) &= P_{\mu} \left(\frac{\bar{X}_n - \mu_0}{\sigma/\sqrt{n}} > c \right) \\ &= P_{\mu} \left(\frac{\bar{X}_n - \mu + \mu - \mu_0}{\sigma/\sqrt{n}} > c \right) \\ &= P_{\mu} \left(\underbrace{\frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}}}_{\approx N(0,1)} > c - \frac{(\mu - \mu_0)}{\sigma/\sqrt{n}} \right) \\ &\approx 1 - \Phi \left(c - \frac{(\mu - \mu_0)}{\sigma/\sqrt{n}} \right) \end{aligned}$$

For $\mu = \mu_0$:

$$\beta(\mu_0) = 1 - \Phi(c)$$

To get $\beta(\mu_0) = \alpha$

$$1 - \Phi(c) = \alpha$$

$$\Rightarrow c = \Phi^{-1}(1-\alpha) \\ = Z\alpha$$

Class activity

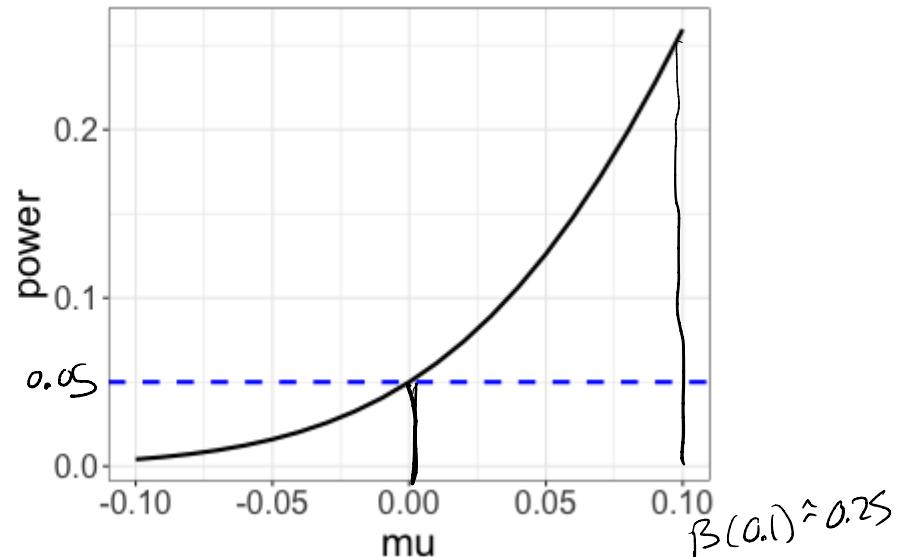
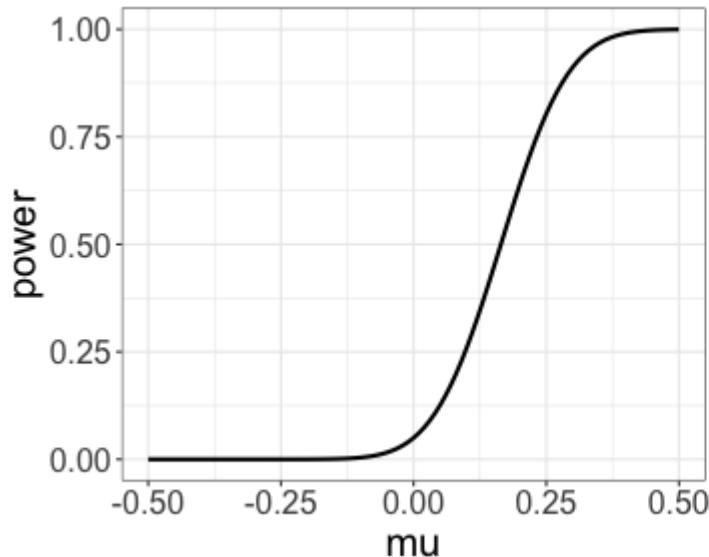
$$\beta(\mu) \approx 1 - \Phi \left(z_\alpha - \frac{(\mu - \mu_0)}{\sigma/\sqrt{n}} \right)$$

- + Suppose that $\mu_0 = 0$, $n = 100$, and $\sigma = 1$. Make a plot of $\beta(\mu)$ vs. μ for $\alpha = 0.05$.
- + Now consider testing $H_0 : \mu \leq \mu_0$ vs. $H_A : \mu > \mu_0$. Will this change our rejection region if we want a size α test?

$$\beta(\mu) = 1 - \Phi\left(z_\alpha - \frac{(\mu - \mu_0)}{\sigma/\sqrt{n}}\right)$$

Class activity

if $\mu < \mu_0$, $z_\alpha - \frac{(\mu - \mu_0)}{\sigma/\sqrt{n}} > z_\alpha$



$\beta(\mu)$ increases as

μ increases

(easier to reject H_0)

$$\sup_{\mu \leq \mu_0} \beta(\mu) = \beta(\mu_0)$$

$$H_0: \mu \leq \mu_0$$

vs.

$$H_A: \mu > \mu_0$$

$$H_0: \mu = \mu_0$$

vs.

$$H_A: \mu > \mu_0$$

same rejection
region for a size
 α test

$$\beta(u) = 1 - \Phi\left(z_\alpha - \frac{(u - u_0)}{\sigma/\sqrt{n}}\right)$$

$$\text{if } u < u_0, z_\alpha - \frac{(u - u_0)}{\sigma/\sqrt{n}} > z_\alpha$$

$$\Rightarrow \Phi\left(z_\alpha - \frac{(u - u_0)}{\sigma/\sqrt{n}}\right) > \Phi(z_\alpha)$$

$$\Rightarrow \beta(u) = 1 - \Phi\left(z_\alpha - \frac{(u - u_0)}{\sigma/\sqrt{n}}\right) < \beta(u_0) = 1 - \Phi(z_\alpha)$$

Rejecting H_0

$$H_0 : \theta \in \Theta_0 \quad H_A : \theta \in \Theta_1$$

A hypothesis test rejects H_0 if (X_1, \dots, X_n) is in the rejection region R . Are there any issues if we only use a rejection region to test hypotheses?

- picking one rejection region only allows us to look at one α
- reject / fail to reject only provides binary information?

Question : what other values of α do we reject for?

p-values

$$H_0 : \theta \in \Theta_0 \quad H_A : \theta \in \Theta_1$$

Given α , we construct a rejection region R and reject H_0 when $(X_1, \dots, X_n) \in R$. Let (x_1, \dots, x_n) be an observed set of data.

Definition: The **p-value** for the observed data (x_1, \dots, x_n) is the smallest α for which we reject H_0 .

Intuition: reject H_0 when p-value $< \alpha$

$$\Rightarrow \inf \{\alpha : \text{reject } H_0\} = \inf \{\alpha : \text{p-value} < \alpha\}$$
$$= \text{p-value}$$

Suppose we have a test which rejects H_0 when $T(X_1, \dots, X_n) > c_\alpha$, where c_α is chosen such that $\sup_{\theta \in \Theta_0} \beta(\theta) = \sup_{\theta \in \Theta_0} P_\theta(T(X_1, \dots, X_n) > c_\alpha) = \alpha$

Let (x_1, \dots, x_n) be a set of observed data.

Theorem: The p-value for (x_1, \dots, x_n) is $\underbrace{P_\theta(T(X_1, \dots, X_n) > T(x_1, \dots, x_n))}_{\text{observed test statistic}}$ ("probability of our test statistic or more extreme")

$$P = \sup_{\theta \in \Theta_0} P_\theta(T(X_1, \dots, X_n) > T(x_1, \dots, x_n))$$

Proof: $P = \inf \{\alpha : \text{reject } H_0\} = \inf \{\alpha : T(x_1, \dots, x_n) > c_\alpha\}$

$$\text{As } \alpha \downarrow, c_\alpha \uparrow \Rightarrow c_p = \sup \{c_\alpha : T(x_1, \dots, x_n) > c_\alpha\} = T(x_1, \dots, x_n)$$

$$\sup_{\theta \in \Theta_0} P_\theta(T(X_1, \dots, X_n) > c_p) = P$$

$$\Rightarrow \sup_{\theta \in \Theta_0} P_\theta(T(X_1, \dots, X_n) > T(x_1, \dots, x_n)) = P \quad //$$

Example

X_1, \dots, X_n iid from a population with mean μ and variance σ^2 .

$$H_0 : \mu = \mu_0 \quad H_A : \mu > \mu_0$$

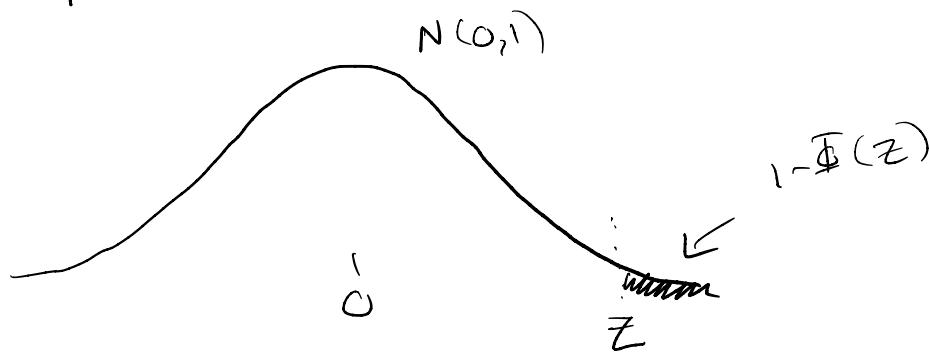
reject H_0 when $Z_n = \frac{\bar{X}_n - \mu_0}{\sigma/\sqrt{n}} > z_\alpha$

Let z be observed test statistic

$$\text{p-value} = P_{\mu_0}(Z_n > z)$$

If $\mu = \mu_0$, $Z_n \approx N(0,1)$

$$\Rightarrow \text{p-value} = 1 - \Phi(z)$$



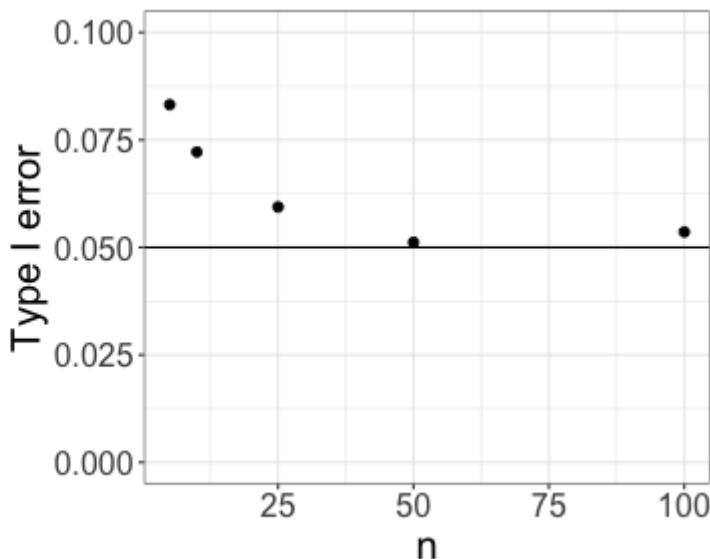
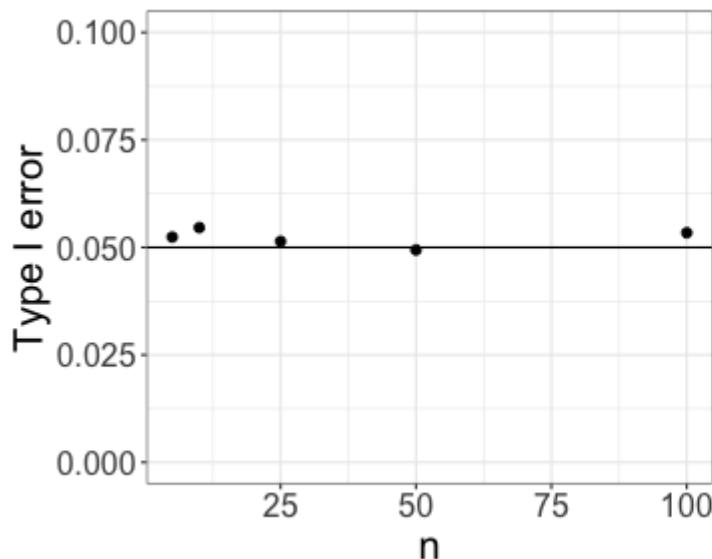
Next steps

So far, we have discussed the Wald test in detail. What other hypothesis tests have you seen in statistics courses?

Class activity

https://sta711-s23.github.io/class_activities/ca_lecture_20.html

Class activity



If we reject $H_0 : \mu = \mu_0$ when $\frac{\sqrt{n}(\bar{X}_n - \mu_0)}{s} > z_\alpha$, why does type I error increase as n decreases?