

# IE605: Engineering Statistics

## Lecture 14

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## Previous Lecture:

- ▶ Hypothesis testing
- ▶ Likelihood Ratio Test (LRT)

## This Lecture:

- ▶ Method of Evaluating Tests
- ▶ Power Functions
- ▶ Uniformly Most Powerful (UMP) test
- ▶ Neyman-Pearson Lemma

# Error in Hypothesis Testing

$H_0 : \theta \in \Theta_0$  null hypothesis

$H_1 : \theta \in \Theta_0^c$  alternate hypothesis

- ▶ Rejection set  $\mathcal{R} = \{\mathbf{x} : \lambda(\mathbf{x}) \leq c\}$ . If  $\mathbf{x} \in \mathcal{R}$  hypothesis  $H_0$  is rejected, otherwise  $H_1$  is accepted
- ▶ In accepting or rejecting hypothesis the experimenter could be making mistake
- ▶ How to control the error?

# Types of Error

There could be two types of error

- ▶ Type I error: If  $\theta \in \Theta_0$  ( $H_0$  is true), but the hypothesis test incorrectly **rejects** the null hypothesis  $H_0$
- ▶ Type II error: If  $\theta \in \Theta_0^c$  ( $H_1$  is true), but the hypothesis test incorrectly **accepts** the null hypothesis  $H_0$

Truth/Decision	Accept	Reject
$H_0$	Correct	<b>Type I error</b>
$H_1$	<b>Type II error</b>	Correct

# Power Function

- ▶ If  $\theta \in \Theta_0$ ,  $P_\theta(\mathbf{X} \in \mathcal{R})$  gives probability of Type I error
- ▶ If  $\theta \in \Theta_0^c$ ,  $1 - P_\theta(\mathbf{X} \in \mathcal{R})$  gives probability of Type II error.

**Power Function** of a hypothesis test with rejection region  $\mathcal{R}$  is a function of  $\theta$  given by  $\beta(\theta) = P_\theta(\mathbf{X} \in \mathcal{R})$

For an 'ideal' hypothesis test

- ▶ what should be the value of  $P_\theta(\mathbf{X} \in \mathcal{R})$  if  $\theta \in \Theta_0$ ?
- ▶ What should be the value of  $P_\theta(\mathbf{X} \in \mathcal{R})$  if  $\theta \in \Theta_0^c$ ?
- ▶ The ideal case cannot be attained in practical problems
- ▶ A good hypothesis test should be such that  $P_\theta(\mathbf{X} \in \mathcal{R}) \rightarrow 1$  when  $\theta \in \Theta_0$  and  $P_\theta(\mathbf{X} \in \mathcal{R}) \rightarrow 0$  when  $\theta \in \Theta_0^c$

## Type I vs Type II error

Example:  $X \in \text{Bin}(5, \theta)$

$H_0 : \theta \leq 1/2$  null hypothesis

$H_1 : \theta > 1/2$  alternate hypothesis

- ▶ Test 1: Reject  $H_0$  iff all success are observed

$$\beta_1(\theta) = \Pr(\{1, 1, 1, 1, 1\}) = \theta^5$$

Type 1 error:  $\beta_1(\theta) \leq (1/2)^5 = 0.0312$  for all  $\theta \leq 1/2$

Type II error:  $1 - \beta_1(\theta) > 0.87$

- ▶ Test 2: Reject  $H_0$  if 3, 4, 5 success are observed

$$\beta_2(\theta) = \binom{5}{3}\theta^3(1-\theta)^2 + \binom{5}{4}\theta^4(1-\theta) + \theta^5$$

## Example:

Let  $\mathbf{X} = \{X_1, X_2, \dots, X_n\}$  is iid drawn from population  $\mathcal{N}(\theta, \sigma^2)$  with known  $\sigma^2$ . We want to test the hypothesis

$H_0 : \theta \leq \theta_0$  null hypothesis

$H_1 : \theta > \theta_0$  alternate hypothesis

$$\mathcal{R} = \{\mathbf{x} : \lambda(\mathbf{x}) \leq c\} = \left\{ \mathbf{x} : \frac{\bar{\mathbf{x}} - \theta_0}{\sigma/\sqrt{n}} \geq c \right\}$$

$$\begin{aligned} \beta(\theta) &= P\left(\frac{\bar{\mathbf{x}} - \theta_0}{\sigma/\sqrt{n}} \geq c\right) \\ &= P\left(\frac{\bar{\mathbf{x}} - \theta}{\sigma/\sqrt{n}} \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}}\right) \\ &= P\left(Z \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}}\right) \end{aligned}$$

where  $Z \sim \mathcal{N}(0, 1)$ .

## Example Contd.

Suppose we want Type I error to be at most 0.1 and Type II error to be at most 0.2 when  $\theta > \theta_0 + \sigma$ . What should be the value of  $c$  and  $n$ ?

$$\beta(\theta) = P\left(Z \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}}\right)$$

- ▶ Increasing in  $\theta$
- ▶ set  $\beta(\theta_0) = 0.1$  and  $\beta(\theta_0 + \sigma) = 0.2$
- ▶ For a fixed sample, it is impossible to make both types of errors arbitrarily small.
- ▶ To find a good test, it is common to restrict Type I at a fixed level.



## Size and Levels

**Definition:** For  $0 \leq \alpha \leq 1$ , a test with power function  $\beta(\theta)$  is a size  $\alpha$  test if  $\sup_{\theta \in \Theta_0} \beta(\theta) = \alpha$ .

**Definition:** For  $0 \leq \alpha \leq 1$ , a test with power function  $\beta(\theta)$  is a level  $\alpha$  test if  $\sup_{\theta \in \Theta_0} \beta(\theta) \leq \alpha$ .

- ▶ Set of level  $\alpha$  tests contains set of size  $\alpha$  tests
- ▶ By fixing  $\alpha$  we are only controlling Type I Error, not Type II error.

## Unbiased Test

A test with power function  $\beta(\theta)$  is unbiased if we have  $\beta(\theta') \geq \beta(\theta'')$  for any  $\theta' \in \Theta_0^c$  and  $\theta'' \in \Theta_0$

**Example:** Let  $\mathbf{X} = \{X_1, X_2, \dots, X_n\}$  is iid drawn from population  $\mathcal{N}(\theta, \sigma^2)$  with known  $\sigma^2$ . We want to test the hypothesis

$H_0 : \theta \leq \theta_0$  null hypothesis

$H_1 : \theta > \theta_0$  alternate hypothesis

$$\beta(\theta) = P\left(Z \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}}\right)$$

Since  $\beta(\theta)$  is increasing. It is clear that  $\beta(\theta)$  is unbiased.

## Most Powerful test

- ▶ There can be class of level  $\alpha$  tests
- ▶ Level  $\alpha$  tests have type I Error probability at most  $\alpha$  in  $\theta \in \Theta_0$
- ▶ A good test in such a class would also have a small Type II Error probability

**Definition:** Let  $\mathcal{C}$  be a class of test for testing  $H_0 : \theta \in \Theta_0$  versus  $H_1 : \theta \in \Theta_0^c$ . A test in class  $\mathcal{C}$  with power function  $\beta(\theta)$  is a uniformly most powerful (UMP) class  $\mathcal{C}$  test if for every power function  $\beta'(\theta)$  in  $\mathcal{C}$  we have  $\beta(\theta) \geq \beta'(\theta)$  for every  $\theta \in \Theta_0^c$ .

# Neyman-Pearson Lemma

**Lemma:** Consider  $H_0 : \theta = \theta_0$  versus  $H_1 : \theta = \theta_1$ . Using a rejection region  $\mathcal{R}$  that satisfies

$$\begin{aligned} \mathbf{x} \in \mathcal{R} & \quad \text{if} \quad f(\mathbf{x}|\theta_1) > \kappa f(\mathbf{x}|\theta_0) \\ \mathbf{x} \in \mathcal{R}^c & \quad \text{if} \quad f(\mathbf{x}|\theta_1) < \kappa f(\mathbf{x}|\theta_0) \end{aligned}$$

for some  $\kappa \geq 0$ , and Let  $\alpha = P_{\theta_0}(\mathbf{X} \in \mathcal{R})$ . Then

- ▶ Any test satisfying above rejection region is a UMP level  $\alpha$  test (necessary condition)
- ▶ Every UMP level  $\alpha$  test satisfies the above rejection regions (sufficient condition)