

# W1211 Introduction to Statistics

## Lecture 24

Wei Wang

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# What we talked about last lecture

- ▶ Confidence Intervals for population mean  $\mu$  based on  $t$  distribution.  
What is the key assumption for using  $t$  distribution.
- ▶ Basic Concept of Hypothesis Testing: the form; null hypothesis and alternative hypothesis.

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- ▶ Null hypothesis and alternative hypothesis are not treated equally. In constructing Testing Procedures, we try to protect null hypothesis, i.e., setting a more stringent standard for rejecting  $H_0$

# Motivating Example

- ▶ Suppose we have a coin, we want to test whether it is unbiased or in favor of head,  $H_0 : p = 0.5$  v.s.  $H_a : p > 0.5$ . We flip the coin for several times, and record the number of heads.
- ▶ Intuitively, how should we conduct the test?

# Testing Procedures

- ▶ A test procedure is specified by the following:
  - ▶ Find a test statistic, a function of the sample data on which the decision (reject  $H_0$  or do not reject  $H_0$ ).
  - ▶ Construct a rejection region, the set of all test statistic values for which  $H_0$  will be rejected.
- ▶ The null hypothesis will then be rejected if and only if the observed or computed test Statistic value falls in the rejection region.

# Example Cont'd

- ▶ Following the aforementioned procedures, we can conduct the test by first selecting a test statistic, and then construct a rejection region.
  - ▶ The natural test statistic is the sample proportion  $\bar{X}$ .
  - ▶ And we will reject the null hypothesis  $p = 0.5$  if  $\bar{X}$  is too large. So the rejection region will look like  $\{\bar{X} > a\}$ .
- ▶ To determine  $a$ , we need to utilize the sampling distribution of the test statistic as well as finer analysis of the errors.

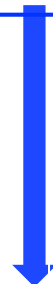


# Two types of errors

- Definition

A type I error  $\alpha$  consists of rejecting the null hypothesis  $H_0$  when it is true.

A type II error  $\beta$  involves not rejecting  $H_0$  when  $H_0$  is false.



	Decide to accept	Decide to reject
Null is true	Right	Type I
Alternative is true	Type II	Right

## Example 8.2 from the Textbook

- ▶ It is known the drying time of a certain type of paint follows a normal distribution with mean 75 min and standard deviation 9 min. A new additive is added to the paint which is believed to lower the mean drying time.
- ▶ If we assume the standard deviation stays the same, then the appropriate Hypotheses are  $H_0 : \mu = 75$  versus  $H_1 : \mu < 75$ . If we use the sample mean of 25 test specimens as our test statistic, and  $\{\bar{X} < c\}$  with cutoff point  $c = 70.8$  as our rejection region.

## Example 8.2 Cont'd

- ▶ We know the sampling distribution of  $\bar{X}$  is  $N(\mu, \frac{9}{25} = 1.8^2)$ .

- ▶ Type I Error

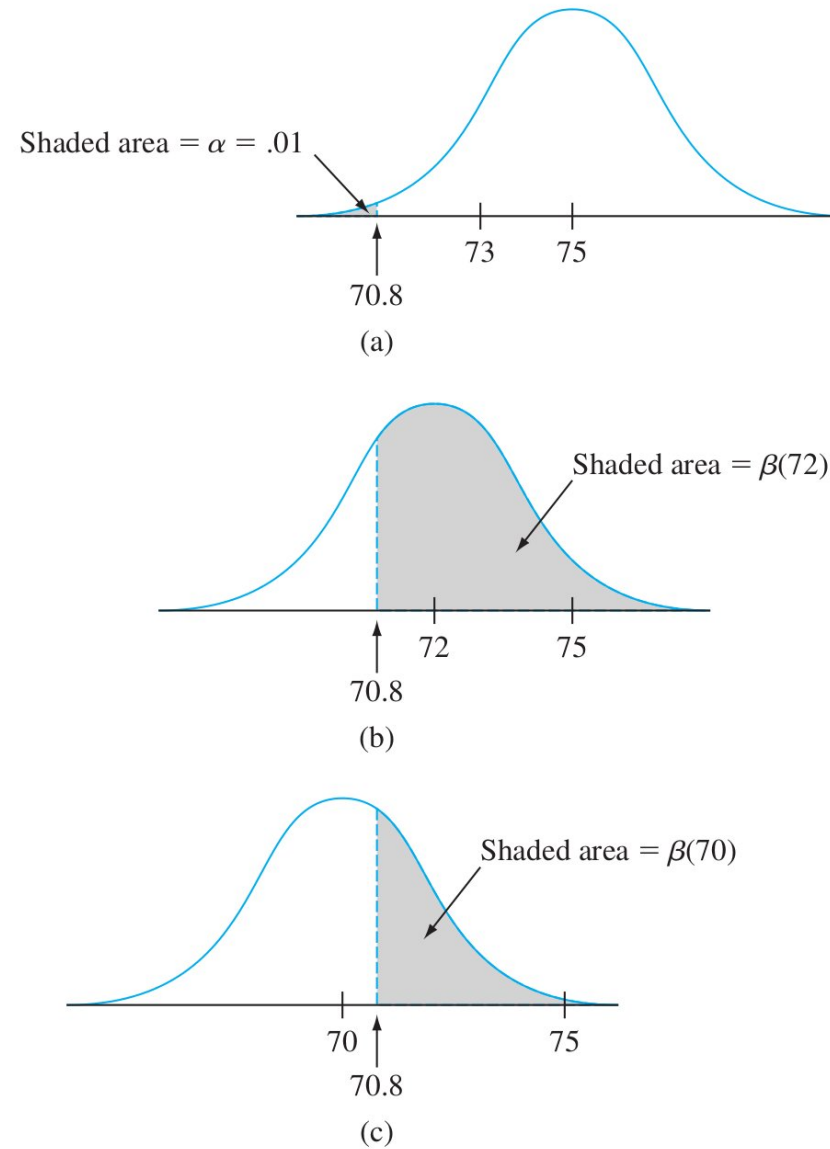
$$\begin{aligned}\alpha &= P(\text{type I error}) = P(H_0 \text{ is rejected when it is true}) \\ &= P(\bar{X} < 70.8 \text{ when } \bar{X} \sim N(75, 1.8^2)) \\ &= P(Z < \frac{70.8 - 75}{1.8}) = 0.01\end{aligned}$$

- ▶ Type II Errors for some values of  $\mu$

$$\begin{aligned}\beta(72) &= P(\text{type II error when } \mu = 72) \\ &= P(\bar{X} > 70.8 \text{ when } \bar{X} \sim N(72, 1.8^2)) \\ &= 1 - P(Z < \frac{70.8 - 72}{1.8}) = 0.7486\end{aligned}$$

$$\beta(70) = 0.33 \quad \beta(67) = 0.0174$$

# Example 8.2 Cont'd



**Figure:** Illustrations of  $\alpha$  and  $\beta$  for the testing procedure: (a)  $\mu = 75$ ; (b)  $\mu = 72$ ; (c)  $\mu = 70$ .

## Example 8.2 Cont'd

- ▶ If we change the cutoff point to 72,  $\alpha$  and  $\beta$  will change correspondingly

- ▶ Type I Error

$$\begin{aligned}\alpha &= P(\text{type I error}) = P(H_0 \text{ is rejected when it is true}) \\ &= P(\bar{X} < 72 \text{ when } \bar{X} \sim N(75, 1.8^2)) \\ &= P(Z < \frac{72 - 75}{1.8}) = 0.05\end{aligned}$$

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$$\beta(70) = 0.1335 \quad \beta(67) = 0.0027$$

# Balancing Two Types of Errors

- ▶ A good test will be aimed to make two types of errors, both  $\alpha$  and  $\beta$ , as small as possible. But simultaneously minimizing the two is impossible once a test statistic is given, so we need to construct a rejection region that effects a good compromise between  $\alpha$  and  $\beta$ .
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- ▶ In practice, people often fix the value of  $\alpha$ , typically at levels such as 0.1, 0.05 and 0.01, which is called **significance level** of the test, and then minimize  $\beta$  subject to the constraint of significance level. The corresponding test procedure is called a **level  $\alpha$  test**.

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- ▶ In applied statistics, another criterion called **power** is also used. It is simply  $1 - \beta$ , which means the probability of rejecting null hypothesis when it is false.