

# Large-Sample Tests of Hypotheses (Part 1)

# Outline

- A statistical test of hypothesis
- Large sample test about a population mean

# A statistical test of hypothesis

Five components of a statistical test

- (1) The **null hypothesis,  $H_0$**
- (2) The **alternative hypothesis,  $H_a$**
- (3) The **test statistic** and its **p-value**
- (4) The **rejection region**
- (5) The **conclusion**

# A statistical test of hypothesis

- (1) The **null hypothesis,  $H_0$**

The hypothesis contradicting  $H_a$ , e.g.  $H_0: \mu = \$456$

- (2) The **alternative hypothesis,  $H_a$**

The hypothesis that we wish to support, for example

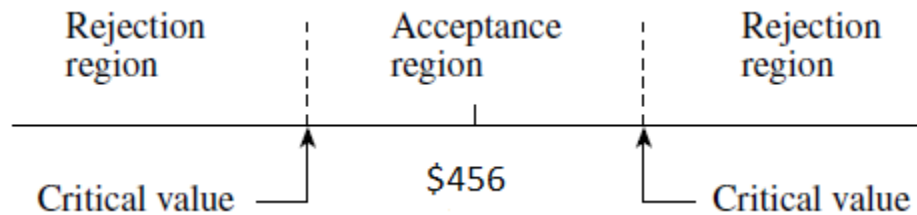
Example 1:  $H_a: \mu \neq \$456$  (2-tailed test of hypothesis)

Example 2:  $H_a: \mu < \$456$  (1-tailed test of hypothesis)

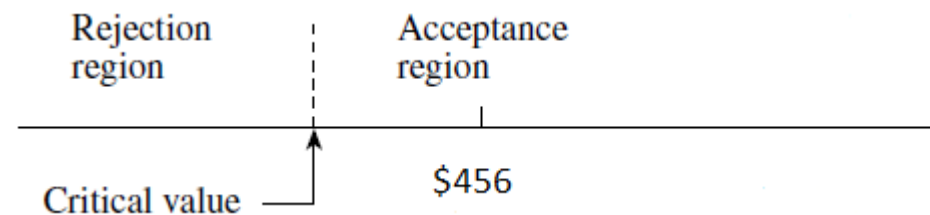
Example 3:  $H_a: \mu > \$456$  (1-tailed test of hypothesis)

# A statistical test of hypothesis

- (3) **Test statistic** is a single value calculated from the sample data and **p-value** is a probability of observing an example as large (or as small) as the test statistic.
- (4) The set of possible values of test statistic can be divided into 2 regions
  - **Rejection region** – includes values that support the alternative hypothesis  $H_a$  and rejects the null hypothesis  $H_0$
  - **Acceptance region** – includes values that support the null hypothesis  $H_0$



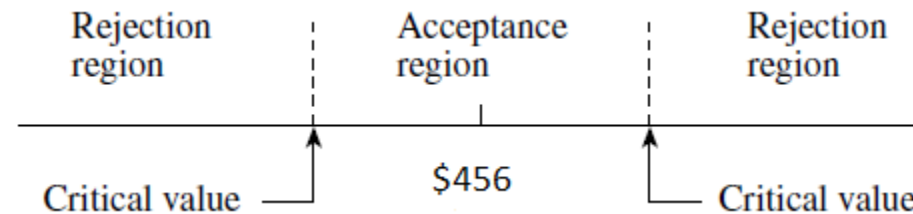
$$H_a: \mu \neq \$456$$



$$H_a: \mu < \$456$$

# A statistical test of hypothesis

- (5) **Conclusions** – we *always* begin with assuming that the null hypothesis is true, then use sample data as evidence to decide one of the 2 conclusions
  - Reject  $H_0$  and conclude  $H_a$  is true
  - Accept  $H_0$  as true or the test is inconclusive



- The critical values are decided based on the **significance level  $\alpha$** , which represents the probability of rejecting  $H_0$  when it is true.
- **Type I error** – the error of rejecting the null hypothesis when it is true.

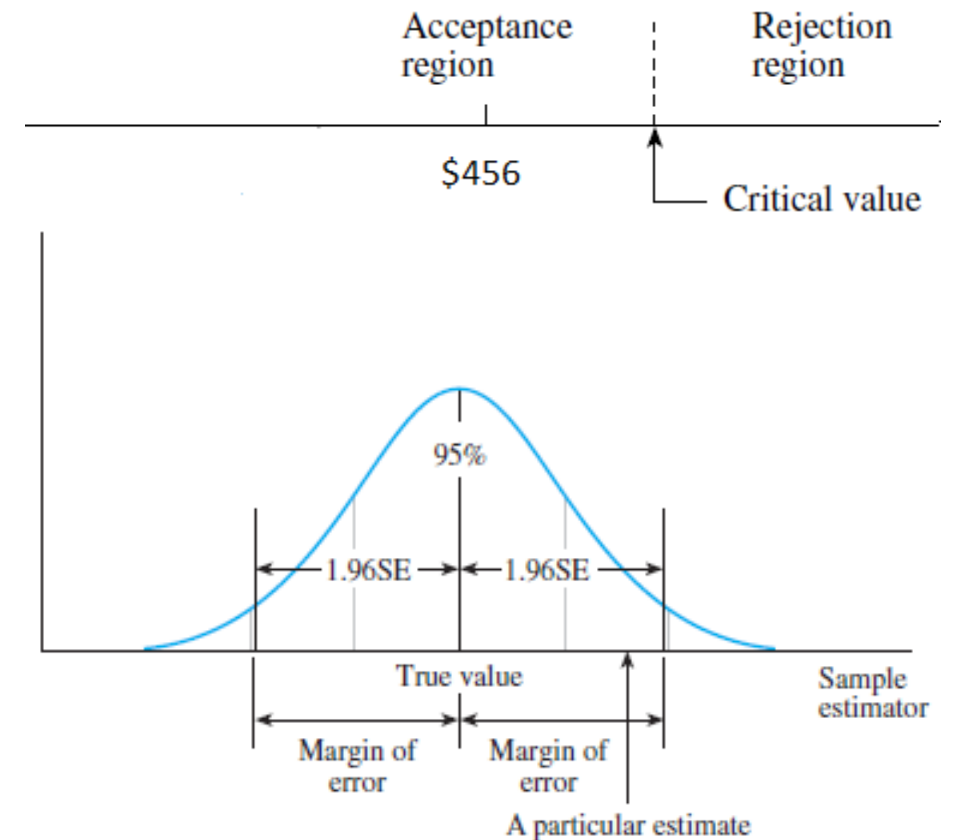
# A large-sample test about a population mean

**Example** – The average monthly income of people in HCMC is \$456. A random sample of  $n=51$  IT professionals in HCMC showed that average income  $\bar{x} = \$500$ , with standard deviation  $s = \$155$ . Do IT professionals have higher monthly income than the city average? Test the hypothesis with significance level  $\alpha = .05$  (or 5%).

- (1) The **null hypothesis**,  $H_0: \mu = \$456$
- (2) The **alternative hypothesis**,  $H_a: \mu > \$456$

# A large-sample test about a population mean

- Because  $n$  is fairly large, the sample mean  $\bar{x} = \$500$  is the best estimate of the true average income  $\mu$  of IT professionals in HCMC (the Central Limit Theorem).
- How large  $\bar{x}$  needs to be compared to  $\mu_0 = \$456$  for us to reject the null hypothesis?
- Because the sampling distribution of  $\bar{x}$  follows a normal distribution, the mean of which is  $\mu$ , if  $\mu_0$  is *many standard errors* (SEs) away from  $\mu$  we can fairly sure that the probability to see  $\mu_0$  is very low, i.e.  $\mu_0$  does not equal  $\mu$ .





# A large-sample test about a population mean

- But how many SEs are enough? We need to rely on the significance level  $\alpha$ .
- Standard error of  $\bar{x}$ ,  $SE = \frac{s}{\sqrt{n}} = \frac{155}{\sqrt{51}} = \$21.9$
- (3) **Test statistic:** The number of SEs  $\mu_0 = \$456$  is away from  $\bar{x}$  is calculated by

$$z = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} = \frac{500 - 456}{21.9} = 2.03$$

In other words,  $\bar{x} = \mu_0 + 2.03 * SE$ .

- (4) Rejection region: For significance level  $\alpha = .05$ , the corresponding z-score is 1.64. Any observed z-value larger than this will be in the rejection region.
- (5) Conclusions: Because the test statistic  $z = 2.03$  is larger than the critical value of 1.64, we reject the null hypothesis, and conclude that the *average monthly income of IT professionals is higher than the city average*.
- The probability of this conclusion being wrong is  $\alpha = 5\%$ .

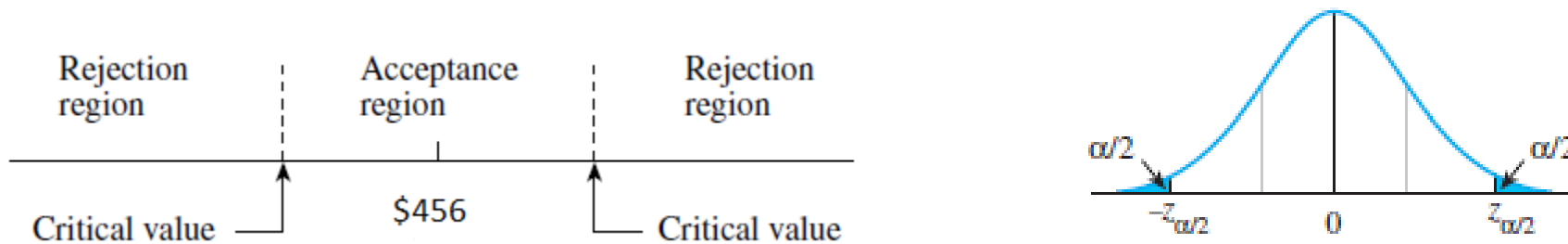
# A large-sample test about a population mean

**Example** – The average monthly income of people in HCMC is \$456. A random sample of  $n=51$  IT professionals in HCMC showed that average income  $\bar{x} = \$500$ , with standard deviation  $s = \$155$ . Do IT professionals have monthly income **different to** the city average? Test the hypothesis with significance level  $\alpha = .05$  (or 5%).

- (1) The **null hypothesis**,  $H_0: \mu = \$456$
- (2) The **alternative hypothesis**,  $H_a: \mu \neq \$456$

# A large-sample test about a population mean

- (3) **Test statistic** – We use the same reasoning as before and come up with the test statistic  $z = 2.03$
- (4) **Rejection region** – In 2 tailed test using significance level  $\alpha = .05$ , the critical values separating the rejection region and the acceptance region corresponds to  $\alpha/2 = .025$  to the right and left of the tail of the standardized normal distribution. These values are  $z = \pm 1.96$ . The rejection region includes  $z < -1.96$  or  $z > 1.96$ .



- (5) **Conclusion** – Because  $z = 2.03$  is larger than 1.96, we ignore the null hypothesis and conclude that the average monthly income of IT professionals *is different to the city average*. The probability of making the wrong decision is  $\alpha = 5\%$ .

# A large-sample test about a population mean

In summary:

1. Null hypothesis:  $H_0 : \mu = \mu_0$

2. Alternative hypothesis:

**One-Tailed Test**

$$H_a : \mu > \mu_0$$

(or,  $H_a : \mu < \mu_0$ )

**Two-Tailed Test**

$$H_a : \mu \neq \mu_0$$

3. Test statistic:  $z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$  estimated as  $z = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$

4. Rejection region: Reject  $H_0$  when

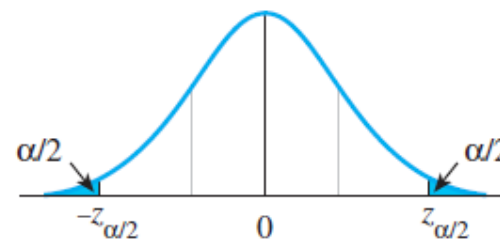
**One-Tailed Test**

$$z > z_{\alpha}$$

(or  $z < -z_{\alpha}$  when the  
alternative hypothesis is  
 $H_a : \mu < \mu_0$ )

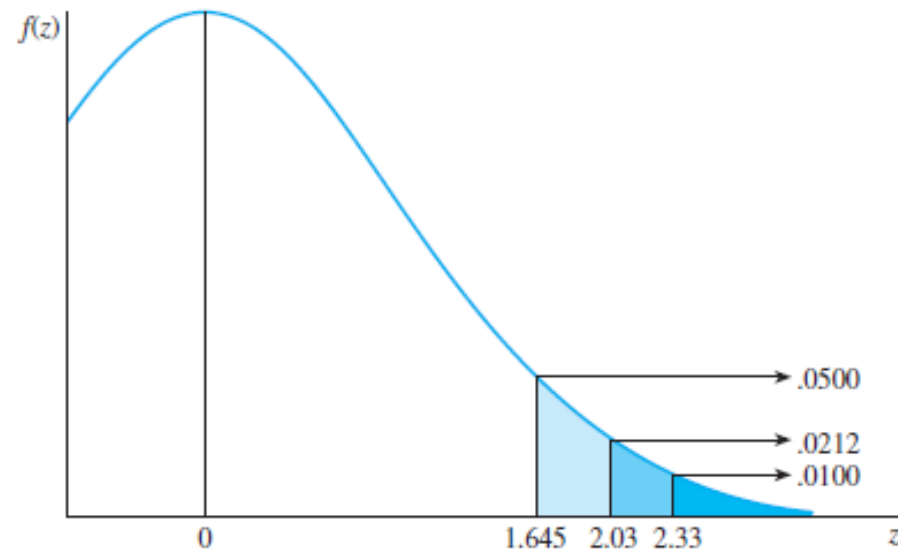
**Two-Tailed Test**

$$z > z_{\alpha/2} \quad \text{or} \quad z < -z_{\alpha/2}$$



# A Large-sample test about a population mean

- In previous examples, the decision to reject a null hypothesis was based on value of  $z$  determined from a significance level  $\alpha$ .
- In Example 1,  $\alpha = .05$ , the critical value of  $z$  is 1.64. We rejected the null hypothesis because the observed value of  $z_0 = 2.03$  is larger the critical value.
- However if  $\alpha = .01$ , the critical value of  $z$  is 2.33, we do not reject the null hypothesis because  $z_0 = 2.03$  is smaller the critical value. (The conclusion in this case is that the *average monthly income of IT professionals is not higher than the city average*)



# A Large-sample test about a population mean

- The smallest critical value that we can use to reject  $H_0$  is 2.03. The probability of this reject decision being wrong is  $P(z > 2.03) = .0212$ , which is the p-value for the test.
- Smaller p-value means larger  $z_0$ , which means larger distance between  $\mu_0 = \$456$  and sample mean  $\bar{x} = \$500$ , which means higher chance of rejecting the null hypothesis.
- p-value can also be compared *directly* with the significance level  $\alpha$ .
  - *If  $p\text{-value} \leq \alpha$ , we reject the null hypothesis and report that the results are **statistically significant** at level  $\alpha$ .*
- In example 1 (one-tailed test),  
p-value =  $P(z > 2.03) = .0212$
- In example 2 (two-tailed test),  
p-value =  $P(z > 2.03) + P(z < -2.03) = .0212 + .0212 = .0424$

# A large-sample test about a population mean

- In example 1, if we set the significant level  $\alpha = .01$ , because p-value = .0212 is larger than  $\alpha$ , we do not reject the null hypothesis and conclude that the average monthly income of IT professionals is not higher than the city average.
- Note that we do NOT say that we *accept* the null hypothesis, i.e. we do NOT conclude that the average monthly income of IT professionals *equals* the city average.
- This is because if we choose to accept the null hypothesis, we need to know the probability of error associate with such a decision.
- ***Type II error*** for statistical test is the error of accepting the null hypothesis when it is false and an alternative hypothesis is true, represented by a probability  $\beta$ .

Decision	Null Hypothesis	
	True	False
Reject $H_0$	Type I error	Correct decision
Accept $H_0$	Correct decision	Type II error