

## 8.2

### Tests About a Population Mean

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### Tests About a Population Mean

- Confidence intervals for a population mean  $\mu$  focused on three different cases.
- We now develop test procedures for these cases.

Case I : Normal Population with Known  $\sigma$

Case II : Large-Sample Tests ພົບກຳນົດໃນລາຍງານ

Case III : Normal Population Distribution ພົບກຳນົດທີ່ມີຄວາມ

→ t-test

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## Case I: A Normal Population with Known $\sigma$

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### Case I: A Normal Population with Known $\sigma$

- Although assumption that value of  $\sigma$  is known is rarely met in practice, this case provides a good starting point because of the ease with which general procedures and their properties can be developed.  $H_0: \mu = \mu_0$
- Null hypothesis in all three cases will state that  $\mu$  has a particular numerical value, the null value, which we will denote by  $\mu_0$ .
- Let  $X_1, \dots, X_n$  represent random sample of size  $n$  from normal population.

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## Case I: A Normal Population with Known $\sigma$

- Then sample mean  $\bar{X}$  has a normal distribution with expected value  $\mu_{\bar{X}} = \mu$  and standard deviation  $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$

- When  $H_0$  is true,  $\mu_{\bar{X}} = \mu_0$

- Consider now the statistic  $Z$  obtained by standardizing  $\bar{X}$  under assumption that  $H_0$  is true:

$$Z = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}}$$

- Substitution of computed sample mean  $\bar{x}$  gives  $z$ , distance between  $\bar{x}$  and  $\mu_0$  expressed in “standard deviation units.”

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## Case I: A Normal Population with Known $\sigma$

- For example, if null hypothesis is

$$H_0 : \mu = 100 \quad \sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}} = \frac{10}{\sqrt{25}} = 2.0 \quad \text{and} \quad \bar{x} = 103$$

then the test statistic value is

$$z = \frac{\bar{x} - \mu}{\sigma_{\bar{X}}} \rightarrow z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}} \rightarrow z = \frac{103 - 100}{\frac{10}{\sqrt{25}}} = 1.5$$

- That is, the observed value of  $\bar{x}$  is 1.5 standard Deviations (of  $\bar{X}$ ) larger than what we expect it to be when  $H_0$  is true.

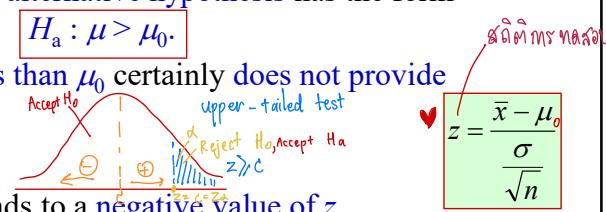
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## Case I: A Normal Population with Known $\sigma$

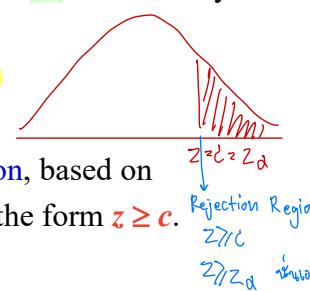
- Statistic  $Z$  is natural measure of **distance** between  $\bar{X}$ , estimator of  $\mu$ , and its **expected value** when  $H_0$  is true.
- If this **distance is too great** in a **direction consistent with  $H_a$** , **null hypothesis** should be **rejected**.
- Suppose first that the **alternative hypothesis** has the form  

$$H_a: \mu > \mu_0.$$
- Then an  $\bar{x}$  value **less than  $\mu_0$**  certainly **does not provide support for  $H_a$** .
- Such an  $\bar{x}$  corresponds to a **negative value of  $z$**  (since  $\bar{x} - \mu_0$  is **negative** and the divisor  $\sigma/\sqrt{n}$  is **positive**). 68



## Case I: A Normal Population with Known $\sigma$

- Similarly, an  $\bar{x}$  value that **exceeds  $\mu_0$**  by only a **small amount** (corresponding to  $z$ , which is **positive but small**) does **not suggest** that  $H_0$  should be **rejected** in favor of  $H_a$ .
- Rejection of  $H_0$  is appropriate only when  $\bar{x}$  considerably exceeds  $\mu_0$ —that is, when the  **$z$  value is positive and large**.
- In summary, **appropriate rejection region**, based on the **test statistic  $Z$**  rather than  $\bar{X}$ , has the form  $z \geq c$ . 69



## Case I: A Normal Population with Known $\sigma$

- As we have discussed earlier, **cutoff value  $c$**  should be chosen to control probability of **type I error** at the **desired level  $\alpha$** .
- The required cutoff  $c$  is  **$z$  critical value** that captures upper-tail area  $\alpha$  under the  **$z$  curve**.

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## Case I: A Normal Population with Known $\sigma$

- As an example,
- let  $c = 1.645$ , value that captures tail area  $0.05$  ( $z_{0.05} = 1.645$ ).

Table A.3 Standard Normal Curve Areas (cont.)

$$\Phi(z) = P(Z \leq z)$$

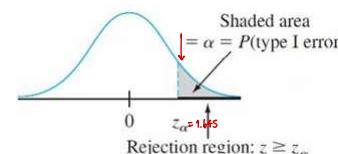
$z$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
1.6	.9452	.9463	.9474	.9484	.9495	.9505	.9515	.9525	.9535	.9545

Then,  $(\alpha) = P(\text{type I error})$

$= P(H_0 \text{ is rejected} \text{ when } H_0 \text{ is true})$

$= P(Z \geq 1.645 \text{ when } Z \sim N(0,1))$

$$= 1 - \Phi(1.645) = 1 - 0.95 = 0.05$$



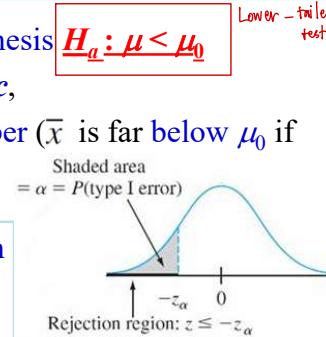
- More generally, rejection region  $z \geq z_\alpha$  has type I error probability  $\alpha$ .

- The test procedure is **upper-tailed** because rejection region consists only of large values of test statistic.

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## Case I: A Normal Population with Known $\sigma$

- Analogous reasoning for alternative hypothesis  $H_a: \mu < \mu_0$  suggests a **rejection region** of the form  $z \leq c$ , where  $c$  is a suitably chosen **negative number** ( $\bar{x}$  is far **below**  $\mu_0$  if and only if  $z$  is quite **negative**).



- Because  $Z$  has **standard normal distribution** when  $H_0$  is true, taking  $c = -z_\alpha$  yields  $P(\text{type I error}) = \alpha$ .

- For example,  $z_{0.10} = 1.28$  implies that the **rejection region**  $z \leq -1.28$  specifies a test with **significance level** 0.10.

$z$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-1.2	.1151	.1131	.1112	.1093	.1075	.1056	.1038	.1020	.1003	.0985

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## Case I: A Normal Population with Known $\sigma$

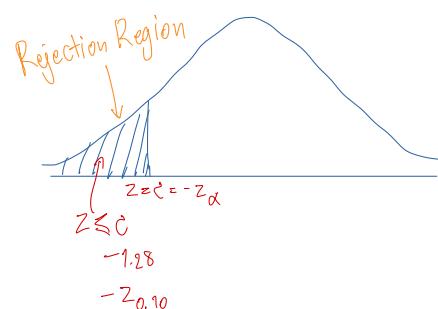
$$\begin{aligned} H_0: \mu &= \mu_0 \\ H_a: \mu &< \mu_0 \end{aligned} \quad \left. \begin{array}{l} \text{Lower-tailed test} \\ \text{Region: } z \leq c \end{array} \right\}$$

$$\begin{aligned} \therefore c &= -z_{\alpha} \\ \text{Ex. } \alpha &= 0.10 \\ \alpha &= P(\text{Type I error}) \\ &= P(\text{Reject } H_0 \text{ when } H_0 \text{ is true}) \\ &= P(z \leq c) \end{aligned}$$

$$0.1 = P(z \leq c)$$

$$\Rightarrow \Phi(-1.28)$$

$$\therefore c = -1.28 \quad \rightarrow c = -z_{\alpha} = -z_{0.10}$$



$z$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-1.2	.1151	.1131	.1112	.1093	.1075	.1056	.1038	.1020	.1003	.0985

Two-tail test

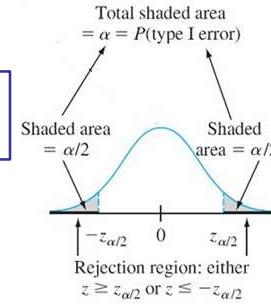
### Case I: A Normal Population with Known $\sigma$

- Finally, when alternative hypothesis is  $H_a: \mu \neq \mu_0$ ,  $H_0$  should be rejected if  $\bar{x}$  is too far to either side of  $\mu_0$ .
- This is equivalent to rejecting  $H_0$  either if  $z \geq c$  or if  $z \leq -c$ .
- Suppose we desire  $\alpha = 0.05$ . Then,

$$0.05 = P(Z \geq c \text{ or } Z \leq -c)$$

when  $Z$  has a standard normal distribution)

$$= \Phi(-c) + (1 - \Phi(c)) = 2[1 - \Phi(c)]$$



- Thus  $c$  is such that  $1 - \Phi(c)$ , area under the  $z$  curve to the right of  $c$ , is 0.025 (and not 0.05!). 76

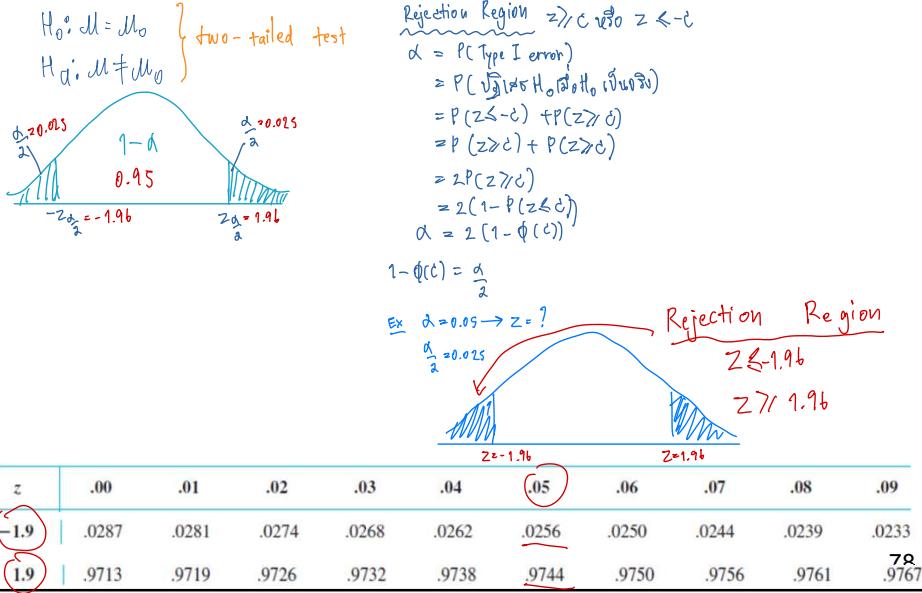
### Case I: A Normal Population with Known $\sigma$

- From Appendix Table A.3,  $c = 1.96$ , and the rejection region is  $z \geq 1.96$  or  $z \leq -1.96$ .

$z$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-1.9	.0287	.0281	.0274	.0268	.0262	.0256	.0250	.0244	.0239	.0233
1.9	.9713	.9719	.9726	.9732	.9738	.9744	.9750	.9756	.9761	.9767

- For any  $\alpha$ , the two-tailed rejection region  $z \geq z_{\alpha/2}$  or  $z \leq -z_{\alpha/2}$  has type I error probability  $\alpha$  (since area  $\alpha/2$  is captured under each of two tails of the  $z$  curve).
- Again, key reason for using standardized test statistic  $Z$  is that because  $Z$  has a known distribution when  $H_0$  is true (standard normal), rejection region with desired type I error probability is easily obtained by using appropriate critical values.

## Case I: A Normal Population with Known $\sigma$



## Case I: A Normal Population with Known $\sigma$

- Test procedure for case I is summarized in accompanying box, and corresponding rejection regions are illustrated in Figure 8.2.

$z$  curve (probability distribution of test statistic  $Z$  when  $H_0$  is true)

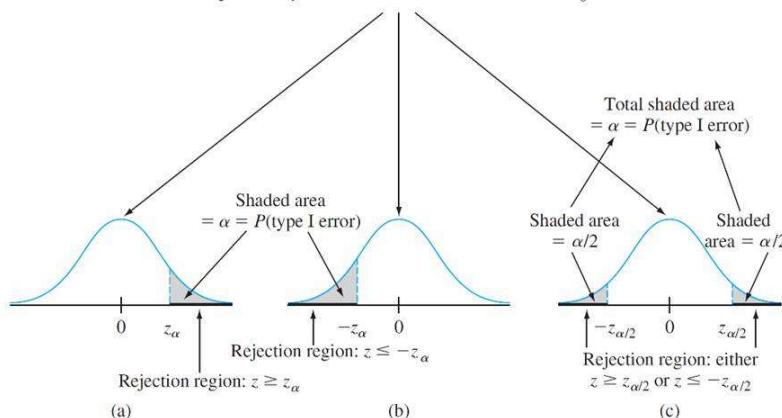


Figure 8.2 Rejection regions for  $z$  tests: (a) upper-tailed test; (b) lower-tailed test; (c) two-tailed test

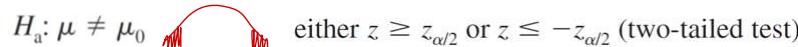
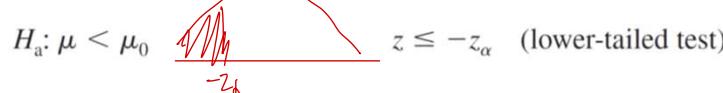
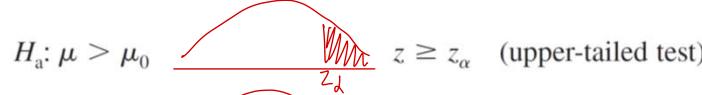
## Case I: A Normal Population with Known $\sigma$

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

Test statistic value :  $z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$



**Alternative Hypothesis      Rejection Region for Level  $\alpha$  Test**



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## Case I: A Normal Population with Known $\sigma$

- Use of the following sequence of steps is recommended when testing hypotheses about a parameter.

1. Identify the parameter of interest and describe it in the context of the problem situation.

2. Determine the null value and state the null hypothesis.

3. State the appropriate alternative hypothesis.

4. Give the formula for the computed value of the test statistic (substituting the null value and the known values of any other parameters, but not those of any sample-based quantities).



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## Case I: A Normal Population with Known $\sigma$

5. State the rejection region for the selected significance level  $\alpha$ .
6. Compute any necessary sample quantities, substitute into the formula for the test statistic value, and compute that value.
7. Decide whether  $H_0$  should be rejected, and state this conclusion in the problem context.

- The formulation of hypotheses (Steps 2 and 3) should be done before examining the data.

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## Example 6



ระบบหัวกระเจยนำดับเพลิงอัดโน้มติ

- A manufacturer of sprinkler systems used for fire protection in office buildings claims that true average system-activation temperature is  $130^\circ$   $\rightarrow \mu_0$
- A sample of  $n = 9$  systems, when tested, yields a sample average activation temperature of  $131.08^\circ\text{F}$ .  $\rightarrow \bar{x}$
- If the distribution of activation times is normal with standard deviation  $1.5^\circ\text{F}$ ,  $\rightarrow \sigma = 1.5$   
does the data contradict the manufacturer's claim at significance level  $(\alpha = 0.01)$ ?

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$$\frac{\alpha}{2} = 0.005$$

## Example 6

cont'd

1. Parameter of interest:  $\mu$  = true average activation temperature.

$$H_0: \mu = \mu_0$$

2. Null hypothesis:  $H_0: \mu = 130$  (null value =  $\mu_0 = 130$ ).

3. Alternative hypothesis:  $H_a: \mu \neq 130$  (a departure from the claimed value in either direction is of concern). two-tailed test

4. Test statistic value:

$$\mu_0 = 130$$

$$z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} = \frac{\bar{x} - 130}{1.5/\sqrt{9}}$$

$$\sigma = 1.5$$

$$n = 9$$

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## Example 6 $H_a: \mu \neq 130$ either $z \geq z_{\alpha/2}$ or $z \leq -z_{\alpha/2}$ (two-tailed test)

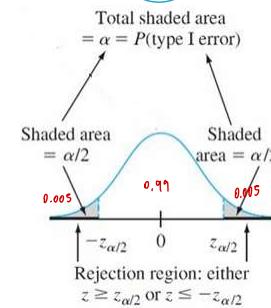
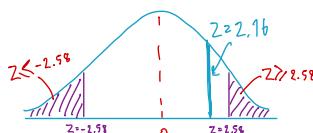
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$$\alpha = 0.01 \quad \frac{\alpha}{2} = 0.005$$

5. Rejection region: The form of  $H_a$  implies use of two-tailed test with rejection region either  $z \geq z_{0.005}$  or  $z \leq -z_{0.005}$ .

$z$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
-2.5	.0062	.0060	.0059	.0057	.0055	.0054	.0052	.0051	.0049	.0038
2.5	.9938	.9940	.9941	.9943	.9945	.9946	.9948	.9949	.9951	.9952

- From Appendix Table A.3,  $z_{0.005} = 2.58$ , so we reject  $H_0$  if either  $z \geq 2.58$  or  $z \leq -2.58$



## Example 6

$H_a: \mu \neq 130$  either  $z \geq z_{\alpha/2}$  or  $z \leq -z_{\alpha/2}$  (two-tailed test)

cont'd

6. Substituting  $n = 9$  and  $\bar{x} = 131.08$ ,

$$z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} = \frac{\bar{x} - 130}{1.5/\sqrt{9}} \rightarrow z = \frac{131.08 - 130}{1.5/\sqrt{9}} = \frac{1.08}{0.5} = 2.16$$

- That is, the observed sample mean is a bit more than 2 standard deviations above what would have been expected were  $H_0$  true.
- 7. The computed value  $z = 2.16$  does not fall in rejection region ( $-2.58 < 2.16 < 2.58$ ), so  $H_0$  cannot be rejected at significance level 0.01.

Data does not give strong support to the claim that the true average differs from the design value of 130.

**End of Section 8.2**