

# Lecture 28

## Review

STAT 8010 Statistical Methods I  
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### Inferences for One Population Mean

**Goal:** To infer  $\mu = \mathbb{E}(X)$  from a random sample  $\{X_1, X_2, \dots, X_n\}$

- Point estimation:

$$\bar{X}_n = \frac{\sum_{i=1}^n X_i}{n}$$

- Interval Estimation:  $100 \times (1 - \alpha)\%$  Confidence Interval (CI)

- $\sigma = \sqrt{\text{Var}(X)}$  is known:

$$\left( \bar{X}_n - z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{X}_n + z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \right)$$

- $\sigma$  is unknown:

$$\left( \bar{X}_n - t_{\alpha/2, df=n-1} \frac{\sigma}{\sqrt{n}}, \bar{X}_n + t_{\alpha/2, df=n-1} \frac{\sigma}{\sqrt{n}} \right)$$



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### Margin of error & Sample Size Calculation

- Margin of error:

$$z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \quad \text{if } \sigma \text{ known}$$

$$t_{\alpha/2, df=n-1} \frac{s}{\sqrt{n}} \quad \text{if } \sigma \text{ unknown}$$

$\Rightarrow$  CI for  $\mu = \bar{X}_n \pm \text{margin of error}$

- Sample size determination:

$$n = \left( \frac{z_{\alpha/2} \times \sigma}{\text{margin of error}} \right)^2,$$

if  $\sigma$  is given



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## Hypothesis Testing for $\mu$

- State the null and alternative hypotheses:

$$H_0: \mu = \mu_0 \text{ vs. } H_a: \mu > \text{ or } \neq \text{ or } < \mu_0$$

- Compute the test statistic:

$$z_{obs} = \frac{\bar{X}_n - \mu_0}{\sigma/\sqrt{n}}, \quad \sigma \text{ known}; \quad t_{obs} = \frac{\bar{X}_n - \mu_0}{s/\sqrt{n}}, \quad \sigma \text{ unknown}$$

- Make the decision of the test:

Rejection Region/ P-Value Methods

- Draw the conclusion of the test:

We (do/do not) have enough statistical evidence to conclude that ( $H_a$  in words) at  $\alpha\%$  significant level.



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## Type I, II Error & Power

| True State     | Decision     |                      |
|----------------|--------------|----------------------|
|                | Reject $H_0$ | Fail to reject $H_0$ |
| $H_0$ is true  | Type I error | Correct              |
| $H_0$ is false | Correct      | Type II error        |

- Type I error:  $\mathbb{P}(\text{Reject } H_0 | H_0 \text{ is true}) = \alpha$

- Type II error:  $\mathbb{P}(\text{Fail to reject } H_0 | H_0 \text{ is false}) = \beta$

- The power (PWR):  $\mathbb{P}(\text{Reject } H_0 | H_0 \text{ is false}) = 1 - \beta$ .

$$\Rightarrow \text{PWR}(\mu_a) = 1 - \beta(\mu_a) = 1 - \mathbb{P}(z^* \leq z_\alpha - \frac{|\mu_0 - \mu_a|}{\sigma/\sqrt{n}})$$

(see the figure in page 5, Lecture 20)



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## Duality of Hypothesis Test with Confidence Interval

There is an interesting relationship between CIs and hypothesis tests. If  $H_0$  is rejected with significance level  $\alpha$  then the corresponding confidence interval does not contain the value  $\mu_0$  targeted in the hypotheses with the confidence level  $(1 - \alpha)$ , and vice versa

| Hypothesis testing at $\alpha$ level                | $(1 - \alpha)$ -level CI                          |
|---|---|
| $H_0: \mu = \mu_0 \text{ vs. } H_a: \mu \neq \mu_0$ | $\bar{X} \pm t(\alpha/2, n-1)s/\sqrt{n}$          |
| $H_0: \mu = \mu_0 \text{ vs. } H_a: \mu > \mu_0$    | $(\bar{X} - t(\alpha/2, n-1)s/\sqrt{n}, \infty)$  |
| $H_0: \mu = \mu_0 \text{ vs. } H_a: \mu < \mu_0$    | $(-\infty, \bar{X} + t(\alpha/2, n-1)s/\sqrt{n})$ |



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## Statistical Inference for $\mu_1 - \mu_2$

- Point estimation:  $\bar{X}_1 - \bar{X}_2$

- Interval estimation:

$$\bar{X}_1 - \bar{X}_2 \pm \text{margin of error},$$

where margin of error =

$$t_{\alpha/2, df^*} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}, \quad df^* = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\frac{(s_1^2/n_1)^2}{n_1-1} + \frac{(s_2^2/n_2)^2}{n_2-1}}$$

When  $s_1$  and  $s_2$  "similar enough", we replace

$$\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}} \text{ by } s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}, \text{ where}$$
$$s_p = \sqrt{\frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1+n_2-2}}$$



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## Hypothesis Testing for $\mu_1 - \mu_2$

- State the null and alternative hypotheses:

- Upper-tailed test:  $H_0 : \mu_1 - \mu_2 = 0$  vs.  
 $H_a : \mu_1 - \mu_2 > 0$

- Lower-tailed test:  $H_0 : \mu_1 - \mu_2 = 0$  vs.  
 $H_a : \mu_1 - \mu_2 < 0$

- Two-tailed test:  $H_0 : \mu_1 - \mu_2 = 0$  vs.  $H_a : \mu_1 - \mu_2 \neq 0$

- Compute the test statistic:

$$t_{obs} = \frac{\bar{X}_1 - \bar{X}_2}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}, \quad \sigma_1 = \sigma_2$$
$$t_{obs} = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}, \quad \sigma_1 \neq \sigma_2$$

- Make the decision of the test:

Rejection Region/ P-Value Methods

- Draw the conclusion of the test



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## Paired T-Tests

- When to use:** before/after study, pairing subjects, study on twins, etc
- $H_0 : \mu_{diff} = 0$  vs.  $H_a : \mu_{diff} > 0$  or  $\mu_{diff} < 0$  or  $\mu_{diff} \neq 0$ , where  $\mu_{diff}$  is the population mean of the paired difference
- Test statistic:  $t_{obs} = \frac{\bar{X}_{diff} - 0}{s_{diff}/\sqrt{n}}$



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ANOVA and Overall F Test

Overall F-Test

- $H_0 : \mu_1 = \mu_2 = \dots = \mu_J$   
 $H_a$  : At least one mean is different

- ANOVA Table:

| Source    | df      | SS   | MS                        | F statistic            |
|-----------|---------|------|---------------------------|------------------------|
| Treatment | $J - 1$ | SSTr | $MSTr = \frac{SSTr}{J-1}$ | $F = \frac{MSTr}{MSE}$ |
| Error     | $N - J$ | SSE  | $MSE = \frac{SSE}{N-J}$   |                        |
| Total     | $N - 1$ | SSTo |                           |                        |

- Test Statistic:  $F_{obs} = \frac{MSTr}{MSE}$ . Under  $H_0$ ,  
 $F^* \sim F_{df_1=J-1, df_2=N-J}$

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Family-Wise Error Rate (FWER) and Multiple Comparisons

- Family-Wise Error Rate (FWER)  $\bar{\alpha}$ : the probability of making 1 or more type I errors in a set of hypothesis tests
- Bonferroni Correction: Adjust the significant level for each of the  $m$  tests to be  $\frac{\alpha}{m}$  to control the **FWER**
- Fisher's LSD and Tukey's HSD

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Linear Contrasts

- Definition: Let  $c_1, c_2, \dots, c_J$  are constants where  $\sum_{j=1}^J c_j = 0$ , then  $L = \sum_{j=1}^J c_j \mu_j$  is called a **linear contrast** of the population means.
- Point Estimation:

$$\hat{L} = \sum_{j=1}^J c_j \bar{X}_j$$

- Interval Estimation:

$$(\hat{L} - t_{(\alpha/2, df=N-J)} \hat{se}_{\hat{L}}, \hat{L} + t_{(\alpha/2, df=N-J)} \hat{se}_{\hat{L}}),$$

where  $\hat{se}_{\hat{L}} = \sqrt{MSE \left( \frac{c_1^2}{n_1} + \dots + \frac{c_J^2}{n_J} \right)}$

- Hypothesis Testing for linear contrasts

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