

# Test 2 Review

## Chapters 16, 17, 18, 20, 21, 22, 23

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## 1 Two Types of Parameters

## 2 Means

- One Sample
- Two Samples

## 3 Porportions

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- Two Samples

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# Two Types of Parameters

- We had two types of parameter: means and proportions.
- Means are averages, derived from numerical data.
- Proportions are ratios, derived from categorical data.
- The population mean is  $\mu$  and the sample mean is  $\bar{x}$ .
- The population proportion is  $p$  and the sample proportion is  $\hat{p}$ .

# Outline

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# Means – One Sample, $\sigma$ Known

- The sampling distribution of  $\bar{x}$  is normal, with mean  $\mu$  and standard deviation  $\sigma/\sqrt{n}$ .
- Thus, a confidence interval for  $\mu$  is

$$\bar{x} \pm z^* \left( \frac{\sigma}{\sqrt{n}} \right).$$

- For hypothesis testing, the test statistic is

$$z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}.$$

# Means – One Sample, $\sigma$ Unknown

- If  $\sigma$  is unknown, then we use  $s$  as an estimator of  $\sigma$  and use  $t$  instead of  $z$ .
- Thus, a confidence interval for  $\mu$  is

$$\bar{x} \pm t^* \left( \frac{s}{\sqrt{n}} \right).$$

- For hypothesis testing, the test statistic is

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}.$$

- Degrees of freedom =  $n - 1$ .



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## Means – Two Samples, $\sigma_1, \sigma_2$ Known

- The sampling distribution of  $\bar{x}_1 - \bar{x}_2$  is normal, with mean  $\mu_1 - \mu_2$  and standard deviation

$$\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}.$$

- Thus, a confidence interval for  $\mu_1 - \mu_2$  is

$$(\bar{x}_1 - \bar{x}_2) \pm z^* \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}.$$

- For hypothesis testing, the test statistic is

$$z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}.$$

## Means – Two Samples, $\sigma_1, \sigma_2$ Unknown

- If  $\sigma_1$  and  $\sigma_2$  are unknown, which is usually the case, then we use  $s_1$  and  $s_2$  as estimators of  $\sigma_1$  and  $\sigma_2$  and use  $t$  instead of  $z$ .
- Thus, a confidence interval for  $\mu$  is

$$(\bar{x}_1 - \bar{x}_2) \pm t^* \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}.$$

- For hypothesis testing, the test statistic is

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}.$$

- Degrees of freedom =  $(n_1 - 1) + (n_2 - 1)$ .

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# Proportions – One Sample

- The sampling distribution of  $\hat{p}$  is normal, with mean  $p$  and standard deviation  $\sqrt{\frac{p(1-p)}{n}}$ .
- Thus, a confidence interval for  $p$  is

$$\hat{p} \pm z^* \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

- For hypothesis testing, the test statistic is

$$z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$$

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# Proportions – Two Samples

- The sampling distribution of  $\hat{p}_1 - \hat{p}_2$  is normal, with mean  $p_1 - p_2$  and standard deviation

$$\sqrt{\frac{p_1(1-p_1)}{n_1} + \frac{p_2(1-p_2)}{n_2}}.$$

- Thus, a confidence interval for  $p_1 - p_2$  is

$$(\hat{p}_1 - \hat{p}_2) \pm z^* \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}.$$

- For hypothesis testing, the test statistic is

$$z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}(1-\hat{p}) \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}},$$

where  $\hat{p} = \frac{x_1 + x_2}{n_1 + n_2}$ .