

# Lecture IV

## Confidence intervals

## Hypothesis tests

STA9750

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# Logistics

- HW1 up on the website, **due 09/26**
- New class dynamic:
  - I'll upload a handout and datasets at least 24h **before** lecture with SAS commands
  - I'll expect you to have looked at the handout and tried things out
  - In lecture, I'll go through the handout and answer questions you had as you went through it
  - After that, I'll lecture about some new topic, which will be the object of the new handout

# Today

- Confidence intervals
- Hypothesis testing
- Review of last week's handout

# Confidence intervals

- **Want:** estimate some unknown  $\theta$
- **Goal:** Provide a range of values that seem “good,” not just one point estimate...
- Suppose that the data  $X$  are independent and identically distributed (iid)

$$x_i \stackrel{\text{iid}}{\sim} P_\theta, \theta \text{ unknown}$$

- A **random** interval  $[L(X), U(X)]$  is a  $(1 - \alpha) \cdot 100\%$  confidence interval (CI) for  $\theta$  if, for all possible  $\theta$ :

$$P(L(X) \leq \theta \leq U(X)) \geq 1 - \alpha$$

# CIs... Interpretation?

- If you use 95% CIs all your life, roughly 95% (or more) will catch the “true value” [if the assumptions under which you derived the CI sort of hold...]
- **What about this particular one?** The 95% guarantee refers to repeated sampling, not a particular dataset

<https://istats.shinyapps.io/ExploreCoverage/>

# Examples

- Normal mean:  $x_i \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$ , then  $(1 - \alpha)\%$  CI for  $\mu$

$n$  sample size

$\bar{x}$  sample mean

$S$  sample standard deviation

$t_{n-1, \alpha/2}$  quantile of Student- $t$ ,  $n-1$  deg. of freedom

$$C_\alpha = \bar{x} \pm t_{n-1, \alpha/2} \frac{s}{\sqrt{n}}$$

It's "exact":  $P(\mu \in C_\alpha) = 1 - \alpha$

If you need more review, this might be helpful:

<http://www2.stat.duke.edu/~vp58/sta111/lecture13.pdf>

# Hypothesis testing

- Want to choose between competing hypotheses  $H_0$  (null) and  $H_1$  (alternative)

		Our decision	
		Don't Reject $H_0$	Reject $H_0$
Truth	$H_0$ is true	OK	Type I error (false positive)
	$H_1$ is true	Type II error (false negative)	OK

# Common practice

1. Assume that  $H_0$  is true
2. Find the  **$p$ -value**: probability of finding data as “extreme” or “more extreme” (in a direction favorable to  $H_1$ ) than the observed data under  $H_0$
3. Reject  $H_0$  if the  **$p$ -value** is below a pre-specified threshold  $\alpha$ . Otherwise, don't reject  $H_0$

If you follow this procedure, you will wrongly reject the null  $\alpha \cdot 100\%$  of the time



# CIs and hypothesis tests

- Given a  $(1 - \alpha) \cdot 100\%$  CI, you can do an  $\alpha$ -level hypothesis test for  $H_0 : \theta = \theta_0$  against  $H_1 : \theta \neq \theta_0$  by checking whether the interval contains  $\theta_0$  or not
- Similarly, given a test, we can find CIs by finding the values of  $\theta_0$  for which  $H_0 : \theta = \theta_0$  isn't rejected

# Two-sample tests, normal mean

$$\begin{array}{ll} y_{i1} \stackrel{\text{iid}}{\sim} N(\mu_1, \sigma^2) & H_0 : \mu_1 = \mu_2 \\ y_{l2} \stackrel{\text{iid}}{\sim} N(\mu_2, \sigma^2) & H_1 : \mu_1 \neq \mu_2 \end{array}$$

- If  $i \neq l$ ,  $y_{i1}$  and  $y_{l2}$  are independent
- **Independent samples:** also independent if  $i = l$
- **Paired tests:** dependent if  $i = l$
- Usually, paired observations are 2 measurements on the same experimental unit
  - *Example:* measurements before and after some treatment

# Contingency tables

- We have two categorical variables
- We want to know if the distribution of one of the variables depends upon the levels of another
- *Example:*
  - Variables: Policy support and party affiliation
  - Does policy support depend on party affiliation?
- Chi-squared test:  $H_0$ : variables are independent  
 $H_1$ : variables are dependent

# ANOVA

- Testing whether all the means from different groups are equal
- We have normal data coming from  $k$  groups

$$y_{ij} \stackrel{\text{ind}}{\sim} N(\mu_j, \sigma^2) \quad \begin{array}{l} j \in \{1, 2, \dots, k\} \text{ group} \\ i \in \{1, 2, \dots, n_j\} \text{ observation} \\ \text{within group } j \end{array}$$

$$H_0 : \mu_1 = \mu_2 = \dots = \mu_k$$

$H_1$  : at least 2 pop. means are different

# Now... Go through handout

- Find  $p$ -values and CIs with SAS