

Review of Inference: Key Terms

1. Estimation v. Inference

- Estimation: What are your best parameter estimates? [BLUE]
 - Inference: Are you sure/close? [confidence intervals; hypothesis tests]
- Confidence intervals:** Intervals generated in this fashion will contain the true parameter value some percentage of the time
 - Hypothesis testing:** Can you reject the null hypothesis at some acceptable statistical significance level?

Our focus: Estimating the Population Mean

- $Y \sim N(\mu, \sigma^2)$... random sampling: $\{Y_1, Y_2, \dots, Y_n\}$ iid $\sim Y$
- Use $\bar{Y} = \frac{1}{n} \sum Y_i$ to estimate μ .
- $\bar{Y} = \frac{1}{n} \sum Y_i \sim N(\mu, \frac{\sigma^2}{n})$... $\frac{\sigma}{\sqrt{n}}$ is the standard deviation (sd) of \bar{Y} ...
 $\frac{\bar{Y} - \mu}{\sigma / \sqrt{n}} \sim N(0,1)$
- But the variance is unknown... so use the unbiased estimator $S_Y^2 = \frac{1}{n-1} \sum (Y_i - \bar{Y})^2$ to estimate σ^2
- $\frac{S_Y}{\sqrt{n}}$ is the *Standard Error* (se) of the sample mean, and an estimator of $\frac{\sigma}{\sqrt{n}}$, $\text{sd}(\bar{Y})$
- The **t statistic** is the **cornerstone of inference**: $\frac{\bar{Y} - \mu}{S_Y / \sqrt{n}} \sim t_{n-1}$
 - Has a t distribution with n-1 dofs

Confidence Intervals

- Confidence interval estimator: $\left[\bar{Y} - c \frac{S_Y}{\sqrt{n}}, \bar{Y} + c \frac{S_Y}{\sqrt{n}} \right] = \left[\bar{Y} \pm c \frac{S_Y}{\sqrt{n}} \right]$
 - c comes from the t statistic and t_{n-1} , the t distribution with n-1 dofs
 - Effectively: estimate \pm c std. errors (c=2 is not a bad place to start)

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Classical Hypothesis Testing

11. Null and Alternative Hypotheses

- H_0 : the *Null* Hypothesis (the hypothesis we are testing)
- H_1 : the *Alternative* Hypothesis, the alternative to H_0
- Almost always focus on H_0 : The true parameter value is 0 ($\mu = 0$)

12. Two types of Error

- Type I – False Rejection; traditionally receives the most attention; *protecting the null*; reject the Null only if there is overwhelming evidence to the contrary
- Type II – False Acceptance

13. Significance Level: α

- the acceptable probability of a Type I error = $P(\text{Reject } H_0 \mid H_0 \text{ is true})$
- small is beautiful: 10%, 5%, 1% , ...

14. Test: Reject the Null Hypothesis $H_0 : \mu = 0$ if $\left| \frac{\bar{Y}}{S_Y / \sqrt{n}} \right| > c_\alpha$

- c_α is defined by $P(|t_{n-1}| > c_\alpha) = \alpha$, the significance level (if t statistic is larger in magnitude than c_α)
- So: Reject the Null Hypothesis if the observed sample mean is at least c_α standard errors (*se's*) away from 0

15. *p* value: $P(|t_{n-1}| > \left| \frac{\bar{y}}{s_y / \sqrt{n}} \right|) = p$

- probability of being $\left| \frac{\bar{y}}{s_y / \sqrt{n}} \right|$ standard errors (*se's*) or further away from 0

16. Hypothesis testing is easier with p values: $p < \alpha \Leftrightarrow \left| \frac{\bar{y}}{s_y / \sqrt{n}} \right| > c_\alpha$

- Look for small p-values below the significance level
- You don't need to worry about different dofs (impacting critical values) just look at the p-value and see if it's small.