

AUA CS 108, Statistics, Fall 2019

Lecture 39

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Contents

- ▶ Tests for the Normal Model Variance
- ▶ Large Sample Hypothesis Testing

Last Lecture ReCap

- ▶ What are we testing when using a (one-sample) t -Test?

Last Lecture ReCap

- ▶ What are we testing when using a (one-sample) t -Test?
- ▶ Describe the t -Test.

Note

Note: In **R** `t.test` command, the default values for parameters are:

- ▶ `mu = 0`
- ▶ `alternative = "two.sided"`
- ▶ `conf.level = 0.95`

Note

Note: In many textbooks, you will find the Critical Values and quantiles, calculated using areas of the Right-Tail. So you can meet in textbooks t -Test with the Rejection Region

$$|t| > t_{n-1, \alpha/2}.$$

In fact, here $t_{n-1, \alpha/2}$ is the point such that the area under the PDF of $t(n-1)$ **right to that point** is $\alpha/2$. This coincides with our standard quantile $t_{n-1, 1-\alpha/2}$, where we are calculating the area to the **left**.

R can calculate also these type of quantiles:

```
qt(1-0.05, df = 15)
```

```
## [1] 1.75305
```

```
qt(0.05, df = 15, lower.tail = FALSE)
```

```
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Large Sample Hypothesis Testing

aka

Asymptotic Testing

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Asymptotic Test for the Mean of General Distribution

Model: X_1, X_2, \dots, X_n are IID from some Distribution with Mean μ and Variance σ^2 . σ^2 is not known. The Parameter is μ ;

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Asymptotic Significance Level: $\alpha \in (0, 1)$; This means that we want to have

$$\mathbb{P}(\text{Type I Error}) = \mathbb{P}(\text{Reject } \mathcal{H}_0 \mid \mathcal{H}_0 \text{ is True}) \rightarrow \alpha, \quad \text{as } n \rightarrow +\infty$$

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Test Statistics: $W = \frac{\hat{\theta}^{MLE} - \theta_0}{1/\sqrt{n \cdot \mathcal{I}(\theta_0)}}$ or $W = \frac{\hat{\theta}^{MLE} - \theta_0}{1/\sqrt{n \cdot \mathcal{I}(\hat{\theta}^{MLE})}}$

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Note: People use this Test only if $n \cdot p_0 > 5$ and $n \cdot (1 - p_0) > 5$.