

Inference Part 2

STAT-S520

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- ▶ These slides complement material from ISI Chapter 9

Hypothesis Testing

- ▶ A claim about μ is made (null hypothesis) about the mean of the population
- ▶ A random sample from this population is obtained and used to obtain the sample mean \bar{x}_n
- ▶ Assuming the claim about μ is correct, we determine how likely is to observe a value as extreme as or more extreme than \bar{x}_n
 - ▶ If, getting \bar{x}_n is very unlikely, we reject the claim (reject the null hypothesis)

Hypotheses

- ▶ The null hypothesis, H_0 , is what we initially assume to be true. It's a statement about μ .
 - ▶ For procedural reasons, we **always** include the equal sign in the statement under H_0
- ▶ The alternative hypothesis, H_1 , is what we would conclude (about μ) if we were to reject H_0 .

ISI Example 9.3

(Under H_0)

$$X_1, X_2, \dots, X_{150} \sim P$$

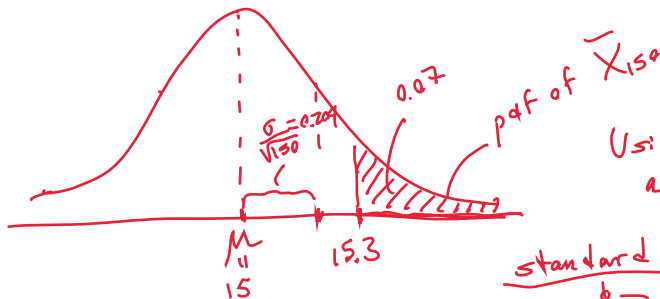
$$E X_i = \mu = 15$$

$$\text{Var } X_i = \sigma^2$$

$$\bar{X}_{150} \sim \text{Normal}(\mu, \frac{\sigma^2}{150}) \quad E \bar{X}_{150} = \mu$$

$$\text{Var } \bar{X}_{150} = \frac{\sigma^2}{150}$$

$$\begin{aligned} \text{SD } \bar{X}_{150} &= \sqrt{\frac{\sigma^2}{150}} \\ &= \frac{\sigma}{\sqrt{150}} \end{aligned}$$



Using $s = 2.5$
as an estimate
of σ

$$\begin{aligned} \text{standard error of } \bar{X}_{150} \\ \text{SD of } \bar{X}_{150} &= \frac{s}{\sqrt{150}} = \frac{2.5}{\sqrt{150}} = 0.204 \end{aligned}$$

Hypotheses:

Null hypothesis

$$H_0: \mu \leq 15$$

μ_0

$$H_1: \mu > 15$$

Alternative hypothesis

City traffic office
perspective.

$$H_0: \mu \geq 15$$

$$H_1: \mu < 15$$

parent's perspective

Aside:

$$H_0: \mu = 15$$

$$H_1: \mu \neq 15$$

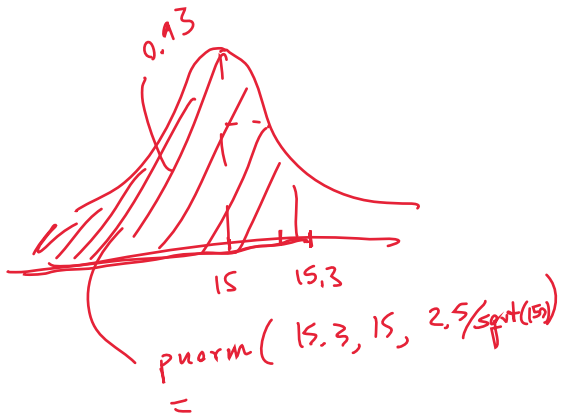
Parents' perspective

$$H_0: \mu \geq 15$$

$$H_1: \mu < 15$$

$$\bar{x} = 15.3$$

$$s = 2.5$$



Test Statistic

The statistic is the method we use to assess the claim made under the null hypothesis. Many statistics exist, the one we encounter often looks like this:

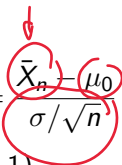
$$\text{statistic} = \frac{\overline{x} - \mu_0}{\text{standard error}}$$

Handwritten annotations in red:

- A bracket above μ_0 and the text "parameter under H_0 " are positioned over the denominator's numerator.
- A bracket below "standard error" is positioned over the denominator's denominator.
- Below the "standard error" bracket, the text "SD \overline{x}_n " is written.

Hypothesis about μ when σ is known

When the hypothesis is made about μ , and σ is known, the test statistic is given by

$$Z = \frac{\bar{X}_n - \mu_0}{\sigma / \sqrt{n}}$$


and due to the CLT, $Z \sim N(0, 1)$.

Hypothesis about μ when σ is unknown

When σ is unknown (as in most real-life problems), the test statistic is given by

$$T = \frac{\bar{X}_n - \mu_0}{\underbrace{S_n}_{\text{red circle}} / \sqrt{n}}$$

where $S_n = \sqrt{\frac{\sum (X_i - \bar{X}_n)^2}{n-1}}$ is the sample standard deviation, another estimator. Given the added uncertainty of S_n , T is no longer normal, but T follows a T-distribution with $n - 1$ degrees of freedom and we write $T \sim T_{n-1}$

Observed test statistic

Once you collect a (random) sample of n observations, the sample observed is $\vec{x} = (x_1, \dots, x_n)$. If σ is known, the observed test statistic is

$$z = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}}$$

and if σ is unknown, the observed test statistic is

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

where $s = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n-1}}$ is the sample standard deviation.

Back to the example
office's perspective

$$t = \frac{15.3 - 15}{2.5 / \sqrt{150}}$$

Significance Probability (p -value)

- ▶ The p -value is the probability of observing a test statistic that is as extreme or more extreme than the one observed in our data, when we assume that H_0 is true.
- ▶ The p -value depends on the hypotheses statements made

Example

$$1 - pt(\underline{t}, 150-1)$$

Conclusion and Interpretation of results

- ▶ We want to reject H_0 if the observed estimate is highly unlikely to have been obtained by chance
 - ▶ The smaller the p -value, the more evidence to reject H_0 .
- ▶ While how small p -value needs to be is somewhat arbitrary, it should be guided by how important is not to make a mistake by rejecting H_0 when it is actually true (Type I Error) or by failing to reject H_0 when it is actually false (Type II Error).

Reasonable ranges for the p -value

To guide you in the decision process, some reasonable values (although still arbitrary) can be:

- ▶ If $p\text{-value} > 0.1$, do not reject H_0 .
- ▶ If $p\text{-value} < 0.001$, reject H_0 .
- ▶ If $0.001 \leq p\text{-value} \leq 0.1$, decide based on your own perception of this uncertainty (different people may make different decisions).
- ▶ Alternatively, come up with a significance level, α , such that if $p\text{-value} \leq \alpha$, we reject H_0 , and if $p\text{-value} > \alpha$, we fail to reject H_0 .
 - ▶ The choice of α should be set before collecting and/or observing the random sample.

ISI Example 9.3 (continued)