

# Lecture 19

## Hypothesis Testing

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- **Hypothesis Testing:** A statistical procedure that use sample data to decide between two competing hypotheses about a population characteristic (e.g.  $\mu$ )
- **Hypotheses:**
  - **Null Hypothesis ( $H_0$ ):** An initial claim about a population characteristic
  - **Alternative Hypothesis ( $H_a$ ):** The competing claim
- $H_0$  will be rejected in favor of  $H_a$  only if sample evidence strongly suggests that  $H_0$  is false. Otherwise,  $H_0$  will not be rejected. Therefore, the two possible decisions are:
  - Reject  $H_0$  (and go with  $H_a$ )
  - Fail to Reject  $H_0$

- In a criminal trial, we use the rule “innocent until proven guilty”
- Therefore, our hypotheses are:
  - $H_0$ : Innocent
  - $H_a$ : Guilty
- If we have strong evidence that the accused is not innocent, we reject  $H_0$  (innocent) and conclude  $H_a$  (guilty)
- If we do not have enough evidence to say that the accused is guilty, we do not say that the accused is “innocent”. Instead, we say that the accused is “not guilty”

Be careful setting up hypotheses:

- A statistical hypothesis test is only capable of demonstrating strong support for the alternative hypothesis  $H_a$  (by rejecting the null hypothesis  $H_0$ )
- Failing to reject  $H_0$  does not show strong support for the null hypothesis – **only a lack of strong evidence against  $H_0$** , the null hypothesis

# The $2 \times 2$ Decision Paradigm for Hypothesis Testing

| True State     | Decision     |                      |
|----------------|--------------|----------------------|
|                | Reject $H_0$ | Fail to reject $H_0$ |
| $H_0$ is true  | Type I error | Correct              |
| $H_0$ is false | Correct      | Type II error        |

## Errors in Hypothesis Testing

- The probability of a **type I error** is denoted by  $\alpha$
- The probability of a **type II error** is denoted by  $\beta$

- In a hypothesis test, our “evidence” comes in the form of a test statistic
- A test statistic incorporates a number of aspects of the sample: the **sample size**, the **point estimate**, the **standard deviation**, and the **hypothesized value**
- If we're conducting a hypothesis test about  $\mu$  (assuming we don't know  $\sigma$ ) we would use the following test statistic:

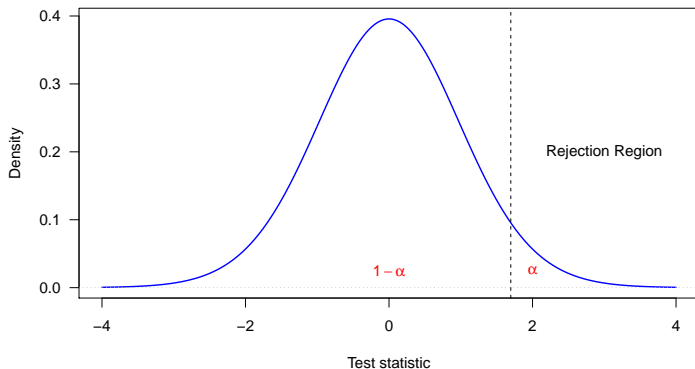
$$t^* = \frac{\bar{X} - \mu_0}{\frac{s}{\sqrt{n}}}$$

If  $\mu = \mu_0$ , we have  $t^* \sim t(df = n - 1)$

- Decision based on  $t^*$ ,  $H_a$ , and  $\alpha$ , the **significant level**, that is pre-defined by the researcher
- Two approaches:
  - **Rejection Region Method**: reject  $H_0$  if  $t^*$  is in the rejection region, otherwise fail to reject  $H_0$
  - **P-Value Method**: reject  $H_0$  if P-value is less than  $\alpha$ , otherwise fail to reject  $H_0$
- **Question**: How to determine the rejection region and how to compute P-value?

## Rejection Region Method

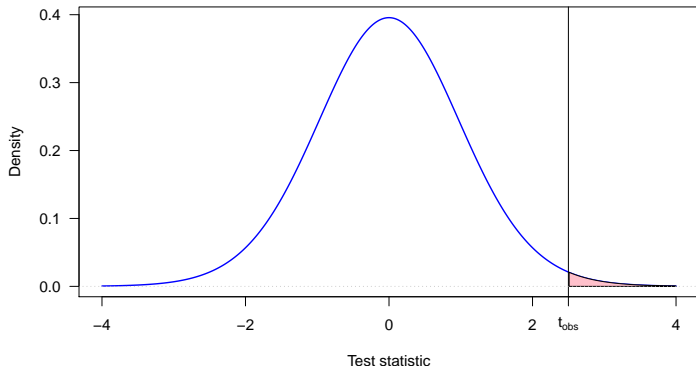
Let  $H_0 : \mu = \mu_0$  vs.  $H_a : \mu > \mu_0$  and  $\alpha = 0.05$



Under the  $H_0$ , the test statistic  $t^* = \frac{\bar{X} - \mu_0}{\frac{s}{\sqrt{n}}} \sim t(df = n - 1)$ . The cutoff of the rejection region ( $=t(0.05, n - 1)$ ) can be found from a t-table



Let  $H_0 : \mu = \mu_0$  vs.  $H_a : \mu > \mu_0$



**P-value:** the probability of getting a test statistic that is at least as extreme as the one we actually observed **if the null hypothesis is true**  $\Rightarrow \mathbb{P}(t^* \geq t_{obs})$

Use the following “generic” conclusion:

“We (do/do not) have enough statistical evidence to conclude that ( $H_a$  in words) at  $\alpha\%$  significant level.”

- Reject  $H_0 \Leftrightarrow$  do
- Fail to reject  $H_0 \Leftrightarrow$  do not

## Example (taken from The Cartoon Guide To Statistics)

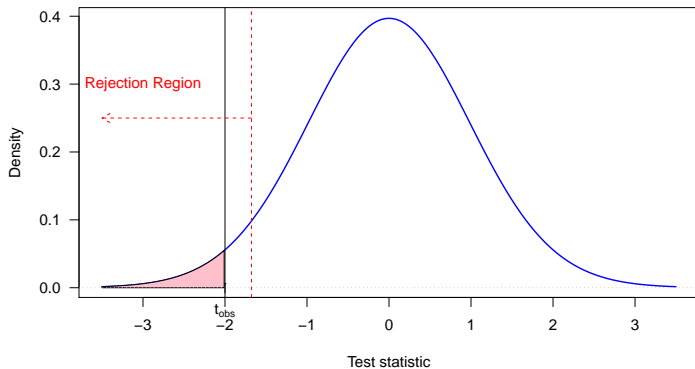
New Age Granola Inc claims that average weight of its cereal boxes is 16 oz. The Genuine Grocery Corporation will send back a shipment if the average weight is any less.

Suppose Genuine Grocery Corporation takes a random sample of 49 boxes, weight each one, and compute the sample mean  $\bar{X} = 15.90$  oz and sample standard deviation  $s = 0.35$  oz.

Perform a hypothesis test at 0.05 significant level to determine if they would reject  $H_0$ , and therefore, this shipment

- 1  $H_0 : \mu = 16$  vs.  $H_a : \mu < 16$
- 2 Test Statistic:  $t_{obs} = \frac{15.9-16}{0.35/\sqrt{49}} = -2$
- 3 **Rejection Region Method:**  $-t(0.05, 48) = -1.68 \Rightarrow$   
Rejection Region is  $(-\infty, -1.68]$ . Since  $t_{obs}$  is in rejection region, we reject  $H_0$
- 4 **P-Value Method:**  $\mathbb{P}(t^* \leq -2) = 0.0256 < \alpha = 0.05 \Rightarrow$  reject  $H_0$
- 5 **Draw a Conclusion:** We do have enough statistical evidence to conclude that the average weight is less than 16 oz at 0.05% significant level

## Cereal Weight Example Cont'd



## Example

A series of blood tests were run on a particular patient over five days. It is of interest to determine if the mean blood protein for this patient differs from 7.25, the value for healthy adults. Suppose the sample mean ( $n=20$ ) is 7.35 and sample standard deviation is 0.5. Perform a hypothesis test using significance level of 0.05

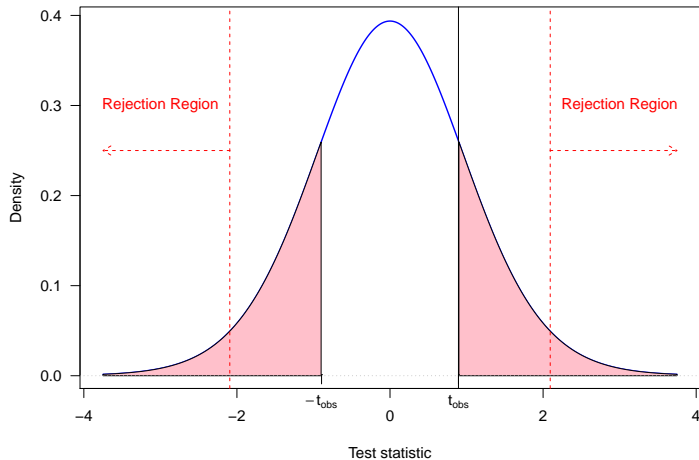
1  $H_0 : \mu = 7.25$  vs.  $H_a : \mu \neq 7.25$

2  $t_{obs} = \frac{7.35 - 7.25}{0.5/\sqrt{20}} = 0.8944$

3 P-value:  $2 \times \mathbb{P}(t^* \geq 0.8944) = 0.3823 > 0.05$

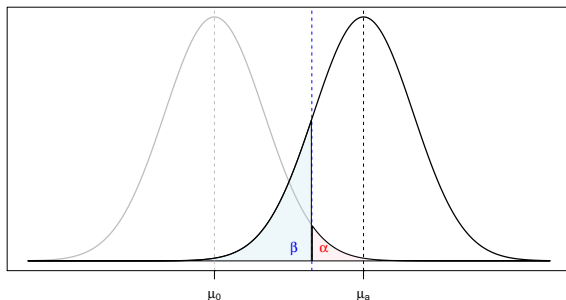
4 We do not have enough statistical evidence to conclude that the mean blood protein is different from 7.25 at 5% significant level

## Example Cont'd



## Type I & Type II Errors

- Type I error:  $\mathbb{P}(\text{Reject } H_0 | H_0 \text{ is true}) = \alpha$
- Type II error:  $\mathbb{P}(\text{Fail to reject } H_0 | H_0 \text{ is false}) = \beta$



$\alpha \downarrow \beta \uparrow$  and vice versa