

Introduction to Statistics

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Topics

1 Hypothesis Test for μ when σ is Unknown

Objectives

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- Carry out a one-mean t test for a population mean μ when the population standard deviation σ is unknown.

Hypothesis Test for μ when σ is Unknown (9.5)

Introduction to the One-Mean t Test

- Usually the population standard deviation σ **isn't known**, so we **can't** use the *one-mean z test* for μ .

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Introduction to the One-Mean t Test

- Usually the population standard deviation σ **isn't known**, so we **can't** use the *one-mean z test* for μ .

Instead, we **estimate** σ by the sample standard deviation s , and then use the **one-mean t test** described ahead.

The test is used when we have a **random sample** from a **population**.

We'll use the sample to decide if the population mean μ is different from some **claimed value** μ_0 .

- The **null hypothesis** is:

Null Hypothesis:

$$H_0 : \mu = \mu_0$$

- The **alternative hypothesis** will depend on what we're trying to **substantiate**:

Alternative Hypothesis: The alternative hypothesis will be one of

1. $H_a : \mu > \mu_0$ (**one-sided, upper-tailed test**)
2. $H_a : \mu < \mu_0$ (**one-sided, lower-tailed test**)
3. $H_a : \mu \neq \mu_0$ (**two-sided, two-tailed test**)

depending on what we're trying to substantiate in our study.

- The **test statistic** for the **one-mean t test for μ** is

One-Mean T Test Statistic:

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

When

$$H_0 : \mu = \mu_0$$

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is true, the **sampling distribution** of the test statistic t is a t distribution with $n - 1$ df.

(The *one-mean t test* is valid if either the sample is from a **normal** population or the sample size n is **large** ($n \geq 30$).)

- t measures (approximately) how many **standard errors** \bar{x} is **away from** the claimed value μ_0 .

1. Values of t **close to zero** provide almost ***no* evidence against** the **null hypothesis**

$$H_0 : \mu = \mu_0.$$

2. **Positive** values of t provide **evidence against the null hypothesis in favor of**

$$H_a : \mu > \mu_0.$$

3. **Negative** values of t provide **evidence against the null hypothesis in favor of**

$$H_a : \mu < \mu_0.$$

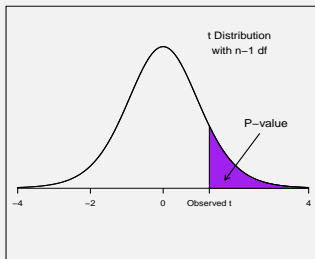
4. **Positive and negative** values of t provide **evidence against the null hypothesis in favor of**

$$H_a : \mu \neq \mu_0.$$

- The ***p-value*** is the probability that just by chance (under H_0) we'd get a test statistic value as far from zero, in the direction predicted by H_a , as the observed value.

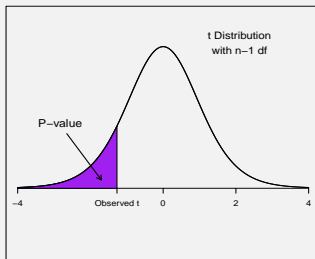
1. **P-value** = Area to the **right** of the observed t if the alternative hypothesis is $H_a : \mu > \mu_0$.

P-Value for Upper-Tailed t Test



2. **P-value** = Area to the **left** of the observed t if the alternative hypothesis is $H_a : \mu < \mu_0$.

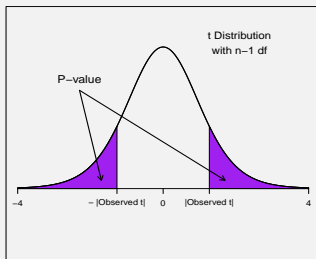
P-Value for Lower-Tailed t Test



Values of t

3. **P-value** = Area to the **left** of $-|t|$ and **right** of $|t|$ if the alternative hypothesis is $H_a : \mu \neq \mu_0$.

P-Value for Two-Tailed t Test



- As always, after choosing a **level of significance** α , we **reject H_0** if **p-value** $< \alpha$, otherwise fail reject H_0 .

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The two the come after it illustrate **one-tailed tests**.

Example

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The machine will need to be adjusted if the boxes are systematically being **under-filled** or **over-filled**.

To decide if the boxes are being **under-filled or overfilled**, the engineer will test the **hypotheses**

$$H_0 : \mu = 16$$

$$H_a : \mu \neq 16$$

where μ is the true (unknown) population mean weight.

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A random sample of **ten** boxes gives

$$\bar{x} = 16.6 \quad \text{and} \quad s = 0.9.$$

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Thus the **sample mean** weight, $\bar{x} = 16.6$, is about **2.11 standard errors above 16** ounces.

The **p-value** is the **probability** that we'd get a t value this far away from zero (in either direction) by chance **if the population mean μ was 16**.

From the **two tail** areas of the t **distribution** with $n - 1 = 9$ **df**, to the **right** of **2.11** and **left** of **-2.11**,

$$\mathbf{p\text{-value}} = 2 \times 0.033 = \mathbf{0.066}.$$

(The value 0.033 was obtained from the **table** showing **areas to the right of t** .)

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Thus we'd get a result like the one we got **6.6%** of the time **even if the population mean μ was 16 ounces**.

Using a **level of significance** $\alpha = 0.05$, the **decision rule** is

Reject H_0 if p-value < 0.05 .

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The result that the engineer got (by taking a random sample) can be explained by chance variation (sampling error).

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Ten men were given a calcium supplement for 12 weeks. Blood pressure was measured both **before** and **after** the 12-week period.

The data are the **changes** in blood pressure for the 10 men:

-7 4 -18 -17 3 5 -1 -10 -11 2

A **negative** value means the **blood pressure decreased**.

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Hint: You should get **-1.81**.

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- c) Determine the **p-value**.

Hint: You should get **0.053**.

d) State the **conclusion** using a **level of significance** $\alpha = 0.05$, in which case the **decision criterion** is

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e) **Interpret** the result: Is there **statistically significant** evidence that calcium **lowers blood pressure**?

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- e) **Interpret** the result: Is there **statistically significant** evidence that calcium **lowers blood pressure**?
- f) If instead we had used a **significance level** $\alpha = 0.10$, would the conclusion have been different? Explain

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To investigate the **claim** that **sugar increases the decay rate of teeth**, **25** adults were examined and then given a sugar solution to supplement all their meals.

After one year, the mean and standard deviation of the number of newly decayed teeth for the group were

$$\bar{x} = 0.60 \quad \text{and} \quad s = 0.45.$$

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- c) Determine the **p-value**.

Hint: You should get **0.002**.

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If you don't have a particular direction in mind for H_a before examining the data, use the two-tailed test.

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- The next example shows how **data snooping** can lead an **artificially small p-value** and in turn to **mistakenly rejecting H_0** when you shouldn't have.

Example

Suppose in the cereal-box weight study from the earlier example that the engineer "**cheats**" (**data snoops**) and decides, **after** noticing that $\bar{x} = 16.6$ is **greater** than **16**, to do an **upper-tailed test** of

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What would the **p-value** be?

Answer: **0.033** (just the **upper** tail area).

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Answer: **Yes**. In this case, the **null hypothesis** would (mistakenly) be **rejected**.