



# One Sample T-Tests



Teaching Demo

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# Overview

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- Hypothesis tests overview
  - Understanding p-values
- One sample T-tests
- Example problem from T-test's origins
  - Interpreting results



# Hypothesis Testing: Overview

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# Hypothesis Test Premise

- Suppose we have  $n$  samples of a random variable  $X_1, X_2, X_3 \dots X_n$
- We have a hypothesis  $H$  about the model that generates our sample
  - Ex:  $X$  drawn from a normal distribution with mean  $\mu = 0$  and standard deviation  $\sigma = 1$
- If we calculate a *test statistic*  $t$  of our sample, we can evaluate likelihood of a statistic like  $t$  being generated if  $H$  is true
  - Ex: *sample mean*,  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$  is a good test statistic if we want to test a hypothesis about distribution mean  $\mu$



# Hypothesis Test Premise

- We cannot prove  $H$ , only reject it or fail to reject it
- Establish a *null hypothesis*  $H_0$ , that we initially assume to be true
  - Ex.  $\mu = 0$
- Establish an *alternative hypothesis*  $H_a$  that contradicts  $H_0$ 
  - $\mu \neq 0$ ,  $\mu > 0$ , or  $\mu < 0$  all valid depending on what alternative we're testing (1 tail vs 2 tail)
- Given a *significance level*  $\alpha$ , we can compute a *rejection region*, which tells us a range of test statistic values that have a likelihood below  $\alpha$  of being observed when  $H_0$  is true
- If our test statistic is in the rejection region, then we can reject  $H_0$  and presume (not prove!) that  $H_a$  is true



# Steps for Conducting a Hypothesis Test

1. Identify the parameter of interest  $\theta$  and describe it in the problem context.
2. Determine the null value  $\hat{\theta}$  and state the null hypothesis  $H_0$ .
3. State the appropriate alternative hypothesis  $H_a$ .
4. Give the formula for the computed value of the test statistic  $t$ .
5. State the rejection region for the selected significance level  $\alpha$ .
6. Compute any necessary sample quantities, substitute into the formula for the test statistic value, and compute that value.
7. Decide if  $H_0$  should be rejected and state this conclusion in the problem context.



# p-values from Hypothesis Tests

- Instead of pre-deciding  $\alpha$  and calculating rejection region, a *p-value* can be calculated which directly indicates the strength of our evidence against the null hypothesis  $H_0$
- A p-value indicates the likelihood that a test statistic more contradictory to  $H_0$  as  $t$  is observed, when  $H_0$  is true
- If  $p < \alpha$ , then our test statistic is in the rejection region for significance level  $\alpha$



# Student's T-Test







# One Sample T-test

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- Hypothesis test for the mean of a distribution
  
- Used in the case of small sample sizes
  - Z-test more common for  $n > 30$

# Historical Context

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- Developed by William Sealy Gosset (1876-1937)
- Head Experimental Brewer at Guinness
- Published anonymously under the name “Student” due to trade secret restrictions
- Interested in hypothesis testing for small sample sizes
  - Farming barley, brewing stout, took lots of time to get samples





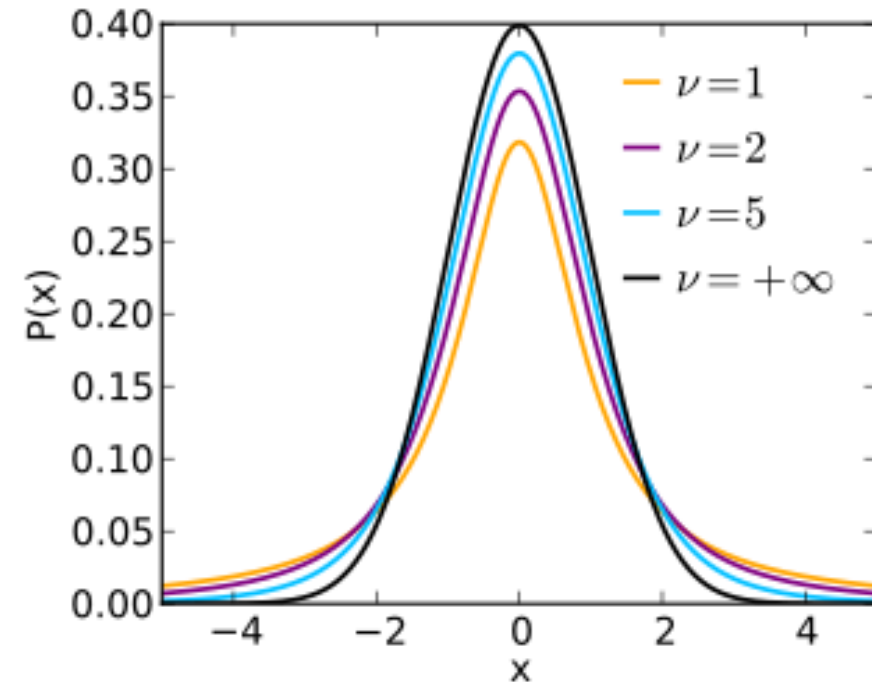
# Assumptions/Conditions for a T-test

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- Assume  $X$  is drawn from a normal distribution
- Assume samples drawn from  $X$  are independent
- No required knowledge of distribution variance, uses sample variance in calculations instead

# “Student’s” T distribution

- Under stated assumptions, sample mean test statistic follows a *T distribution* when null hypothesis of  $\mu = 0$  is true
- Shape of T distribution depends on *degrees of freedom*  $\nu$ 
  - T distribution approaches normal distribution as  $\nu$  approaches infinity
- For one sample t-test with  $n$  observations, utilize T distribution with  $n - 1$  degrees of freedom to calculate rejection regions/p-values





# Conducting a One Sample T-test

- Given a sample mean  $\bar{X}$ , sample standard deviation  $s$ , and a hypothesis distribution mean  $\mu_0$ , the test statistic directly compared against the T distribution is:

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

- Calculating rejection regions from T distributions by hand is *not easy*, lookup tables will give rejection regions for common values of  $\alpha$ ,  $\nu$
- Existing packages that can do this for you
  - Excel: TDIST( $t$ ,  $\nu$ , # of tails) directly calculates p-value
  - Python: `scipy.stats.ttest_1samp(sample,  $\mu_0$ )` calculates  $t$  and p-value



# Example: Barley Crops



# Scenario

- We work for Guinness, making (non-alcoholic) beer from barley farmed on-site
- Barley produces multiple grains per “head”
- The average grain of barley is 42 mg
- Want to determine if our barley’s mean grain mass is below average, to see if we should try planting a different variety for better yields
  - Test to significance level of  $\alpha = 0.05$





# Scenario

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- Collected samples of 20 heads of barley from this year's crop
- Data found in a .csv file at: [ari-smith-research.github.io/TeachingDemo](https://ari-smith-research.github.io/TeachingDemo)





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# Next Class

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- Comparing two populations/samples
- Paired 2 sample T tests: a simple variation on 1 sample
- Unpaired 2 sample T tests
  - Example: comparing two candidate crops

