ECON 0150 | Economic Data Analysis The economist's data analysis pipeline.

Part 4.1 | Extending The t-Test

One-Sample t-Test: Example (using sample stats)

How surprising would it be for the average wait time to be 10 minutes?

One-sample t-test:

- $H_0: \mu = 10$
- $H_1 : \mu \neq 10$
- n: 29
- t-stat: t = 2.401
- *p-value*: 0.0230

```
1 # Imports
2 import numpy as np
3 from scipy import stats
```

```
1  # Sample Data
2  sample_mu = 10.864
3  pop_mu = 10  # null hypothesis
4  std_dev = 1.971
5  n = 30
```

```
1 # Calculate t-statistic
2 t_stat = (sample_mu - pop_mu) / (std_dev / np.sqrt(n))
```

```
1 # Calculate p-value
2 p_value = 2 * (1 - stats.t.cdf(abs(t_stat), df=n-1))
```

> so there is a 2.3% chance of seeing data this extreme given a true mean of 10

One-Sample t-Test: Example (using sample data)

How surprising would it be for the average wait time to be 10 minutes?

> we can perform this very simply from raw data

```
1 # Imports
2 import scipy.stats as stats
3 import numpy as np
```

> load the observed data

> define the null hypothesis

```
1 # Null Hypothesis
2 null = 10
```

> perform the test

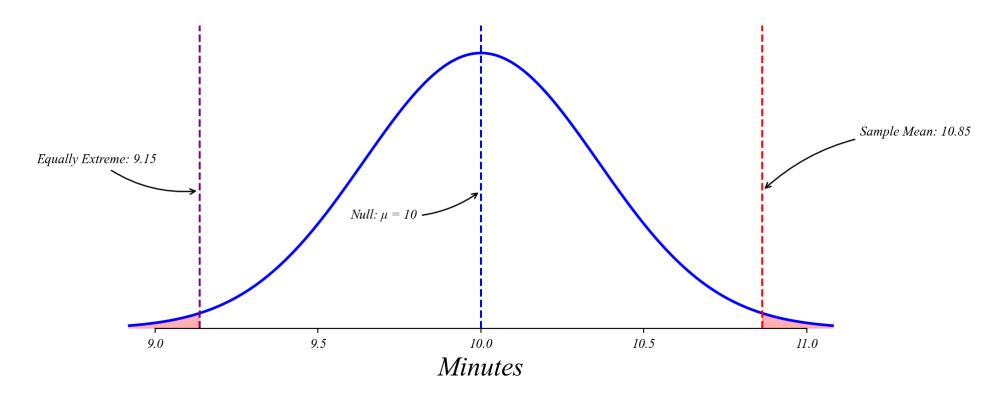
```
1 # Calculate t-statistic and p-value
2 t_stat, p_value = stats.ttest_1samp(data, null)
```

> so there is a 2.3% chance of seeing data this extreme given a true mean of 10

One-Sample t-Test: Example

How surprising would it be for the average wait time to be 10 minutes?

> so there is a 2.3% chance of seeing data this extreme given a true mean of 10



- > but why check both tails?
- > why also consider 9.15, which is equally far from 10 but on the opposite side?

One-Tail vs Two-Tail

... there are at least two key reasons for a two-tail test.

- 1. Scientific Integrity: Since we're testing the hypothesis " $\mu = 10$ " against " $\mu \neq 10$ ", we should be equally open to evidence in either direction.
- 2. **Statistical Reasoning**: If the true mean is 10, our sampling distribution is centered there, and random samples could fall on either side.

t-Test: Differences in Means

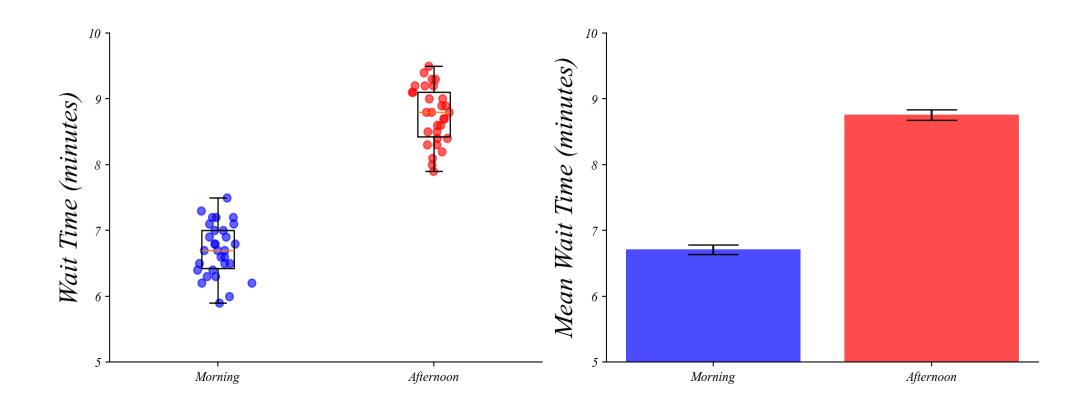
Instead of asking whether $\mu = 10$ lets ask whether $\mu_1 = \mu_2$

> the core question: "are two means different from each other?"

Common scenarios:

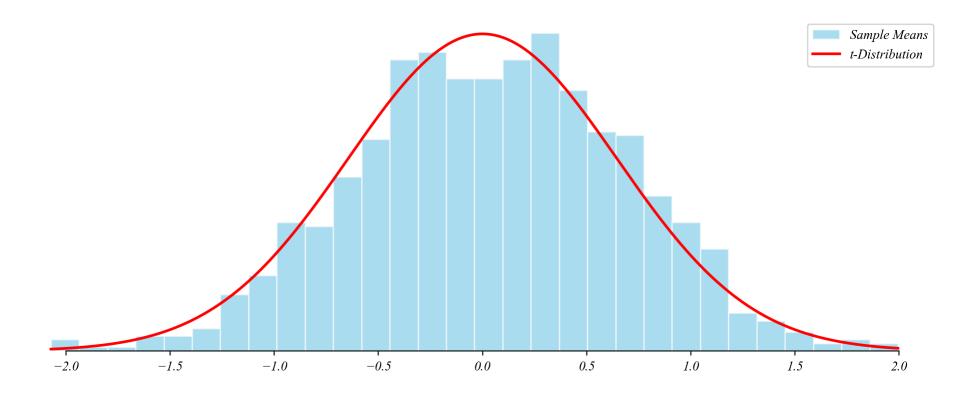
- Wait times at different times of day
- Appointment lengths with different doctors
- Wages across different groups
- Treatment effects in experimental settings

Two-Sample t-Test: Example Are wait times different in the morning vs the afternoon?



- > is there a way to turn this into a one-sample t-Test that we're familiar with?
- > what if we took the differences between sample menas?

> I ran many sample means from two distributions with equal means



- > the distribution of differences centers on zero
- > but what is the standard error of the difference between two samples?

The Math: Combining Standard Errors What is the standard error of the difference in sample means?

For a single sample mean:

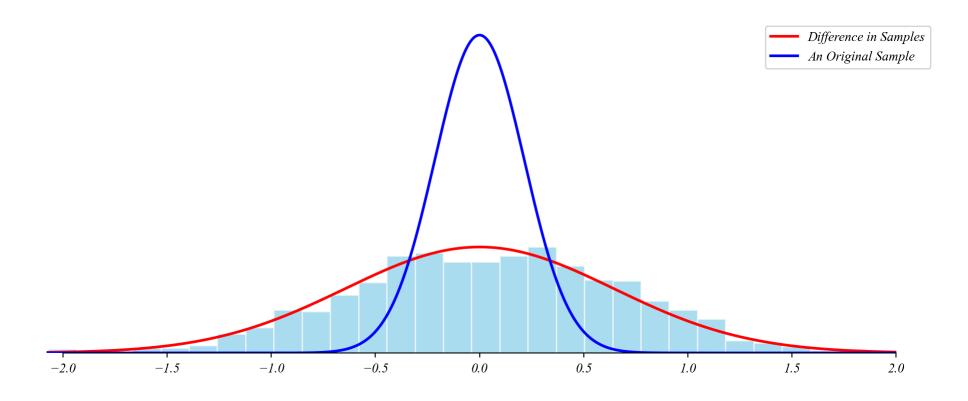
$$SE_{\bar{x}} = \sqrt{\frac{S^2}{n}}$$

For the difference in sample means:

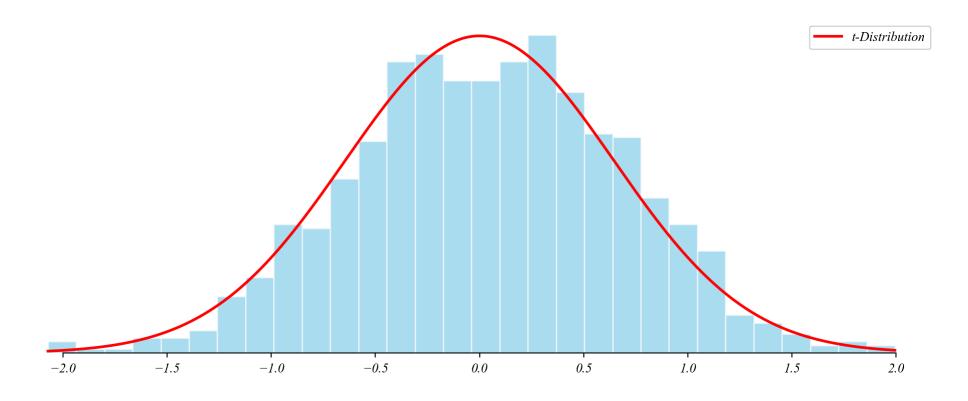
$$SE_{\bar{x}_1 - \bar{x}_2} = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

- > two sources of sampling variation combine, but not additively
- > if variances and sample sizes are equal, then $SE_{\bar{x}_1 \bar{x}_2} = \sqrt{\frac{2s^2}{n}} = \frac{s\sqrt{2}}{\sqrt{n}}$

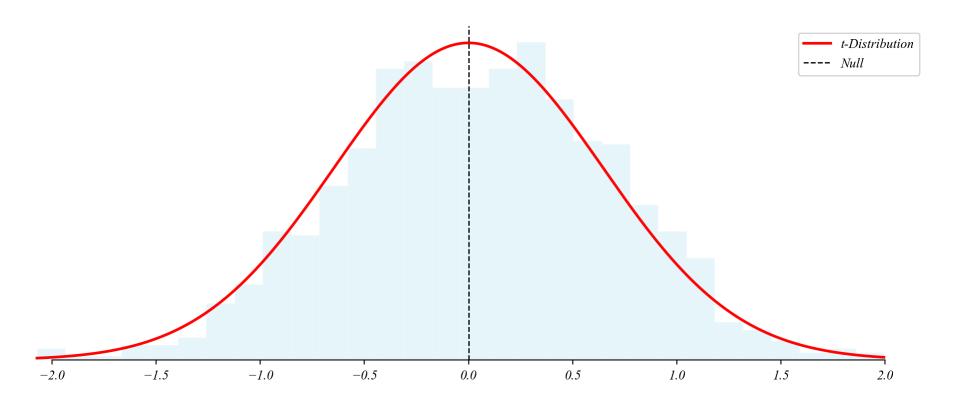
> so the distribution of the differences will be wider than either sample



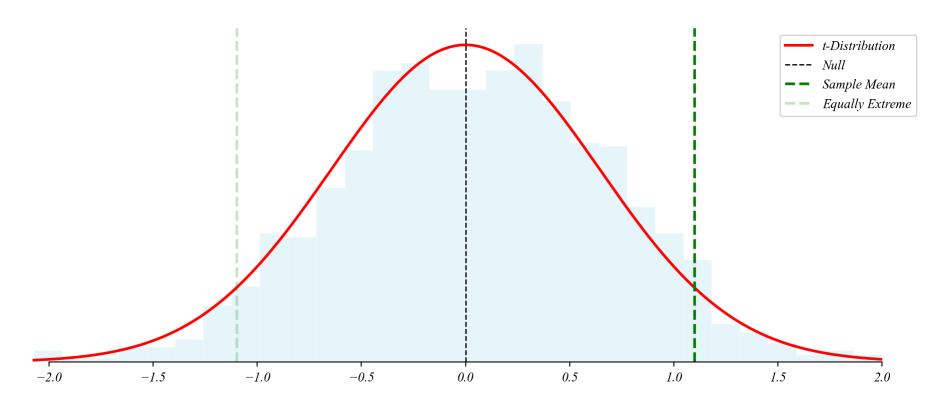
> the null hypothesis comparing two means is typically: $\mu_1 - \mu_2 = 0$



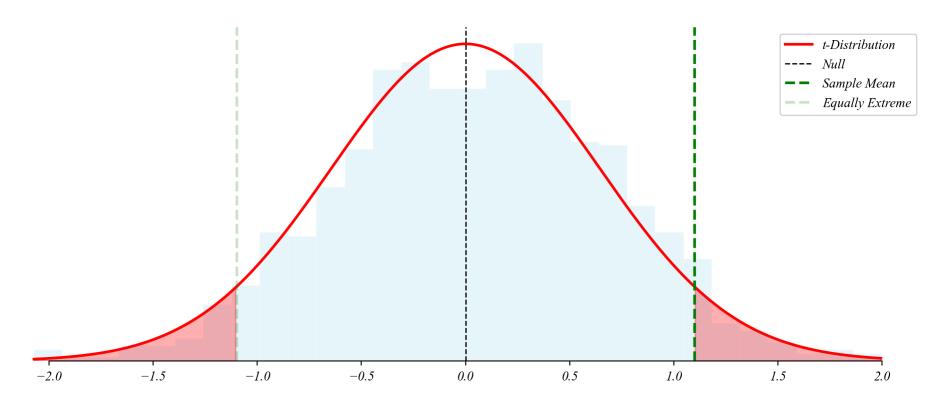
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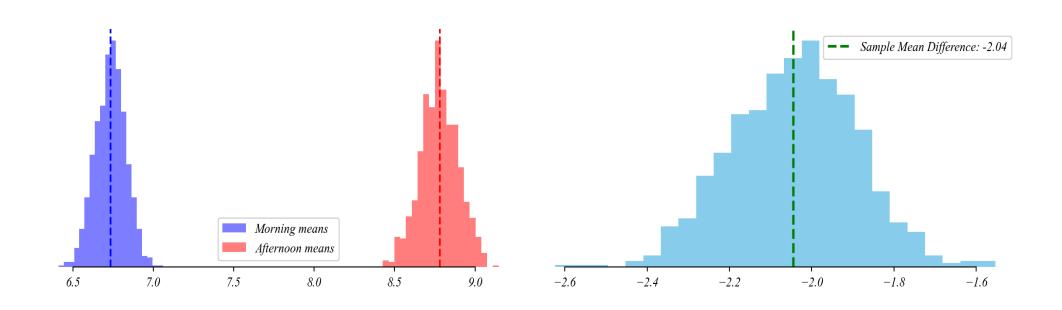
> if the observed difference is $\bar{x}_1 - \bar{x}_2 = 1.1...$

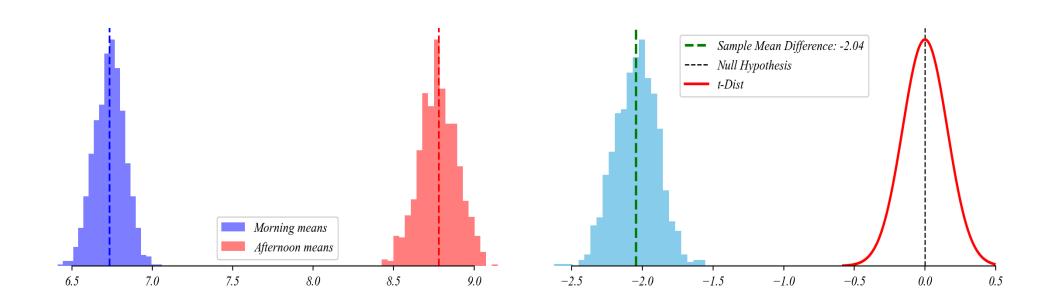


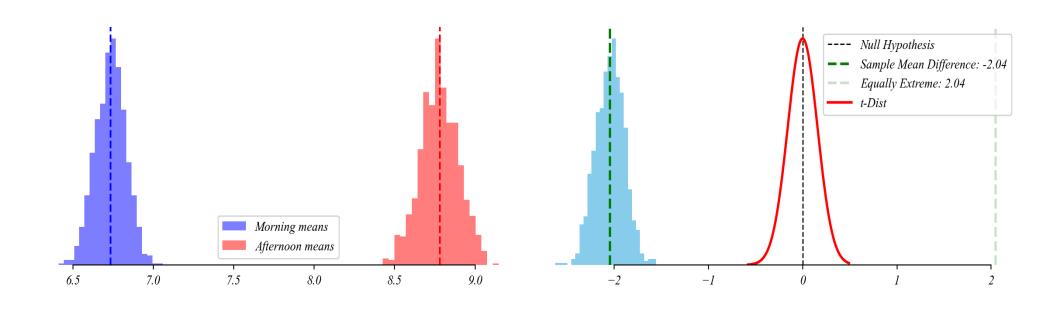
> if the observed difference is $\bar{x}_1 - \bar{x}_2 = 1.1...$



> like before, a p-value quantifies how surprising would be a big difference







- > the p-value tells us how often we'd see differences this extreme by chance
- > here it looks like the area is basically zero

Two-Sample t-test: Step by Step

Example: Morning vs. Afternoon Wait Times

> load the data

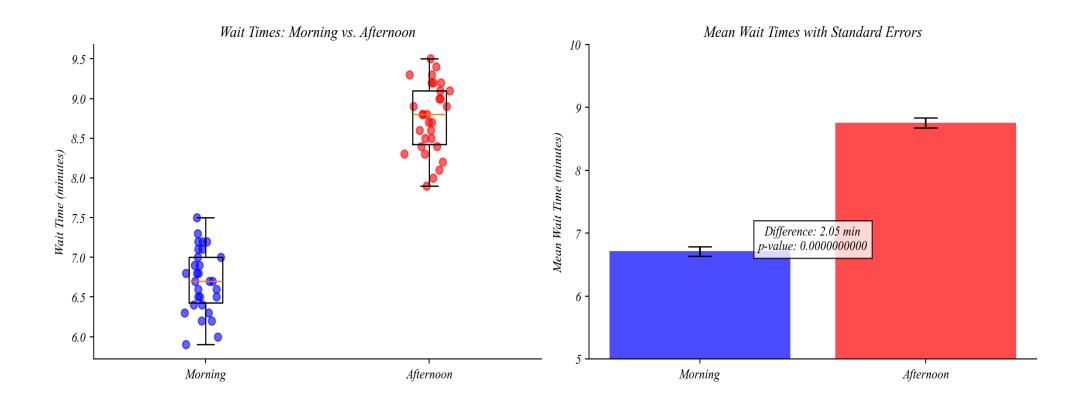
```
1 morning = [
2     7.2, 6.8, 5.9, 7.5, 6.5, 6.0, 7.1, 6.3, 6.7, 6.4,
3     6.9, 7.2, 6.2, 6.8, 7.0, 6.5, 7.3, 6.6, 7.1, 6.7,
4     6.3, 6.9, 6.4, 7.0, 6.6, 6.2, 6.8, 7.2, 6.5, 6.7
5 ]
6
7 afternoon = [
8     8.5, 9.2, 8.1, 8.8, 9.5, 8.3, 7.9, 9.0, 8.7, 9.3,
9     8.6, 9.1, 8.4, 8.9, 9.4, 8.2, 9.0, 8.5, 8.8, 9.2,
10     8.7, 9.1, 8.3, 8.9, 9.3, 8.0, 8.6, 9.2, 8.4, 8.8
11 ]
```

> perform two-sample t-test using scipy.stats

```
1 t_stat, p_value = stats.ttest_ind(morning, afternoon, equal_var=False)
```

- > small p-value = surprising difference (reject null hypothesis of equal means)
- > large p-value = difference could easily happen by chance (fail to reject null)

Visualizing the Difference Morning vs. Afternoon Wait Times



- > wait times are longer in the afternoon than the morning (p<0.001)
- > reject the null that the means are equal



One-sample t-test:

- Tests the sample mean against a specific null value
- Next time: regression with only an intercept

Two-sample t-test:

- Tests if the difference in sample means is zero
- Next time: regression with an intercept and one dummy variable

Common Applications in Economics Two-sample t-tests are one of the most common tests in economic research!

Labor Economics:

- Wage gaps between different demographic groups
- Employment effects of policy changes

Development Economics:

- Impact of interventions on economic outcomes
- Differences between treatment and control groups

Financial Economics:

- Comparing returns across different time periods
- Testing market efficiency

Looking Forward

Connecting to regression and extending to multiple groups

Next time:

- Connecting to regression
- Comparing more than two groups (ANOVA)

Coming soon:

- Multiple regression
- Controlling for confounding variables
- Interaction effects
- > all built on the same fundamental statistical framework