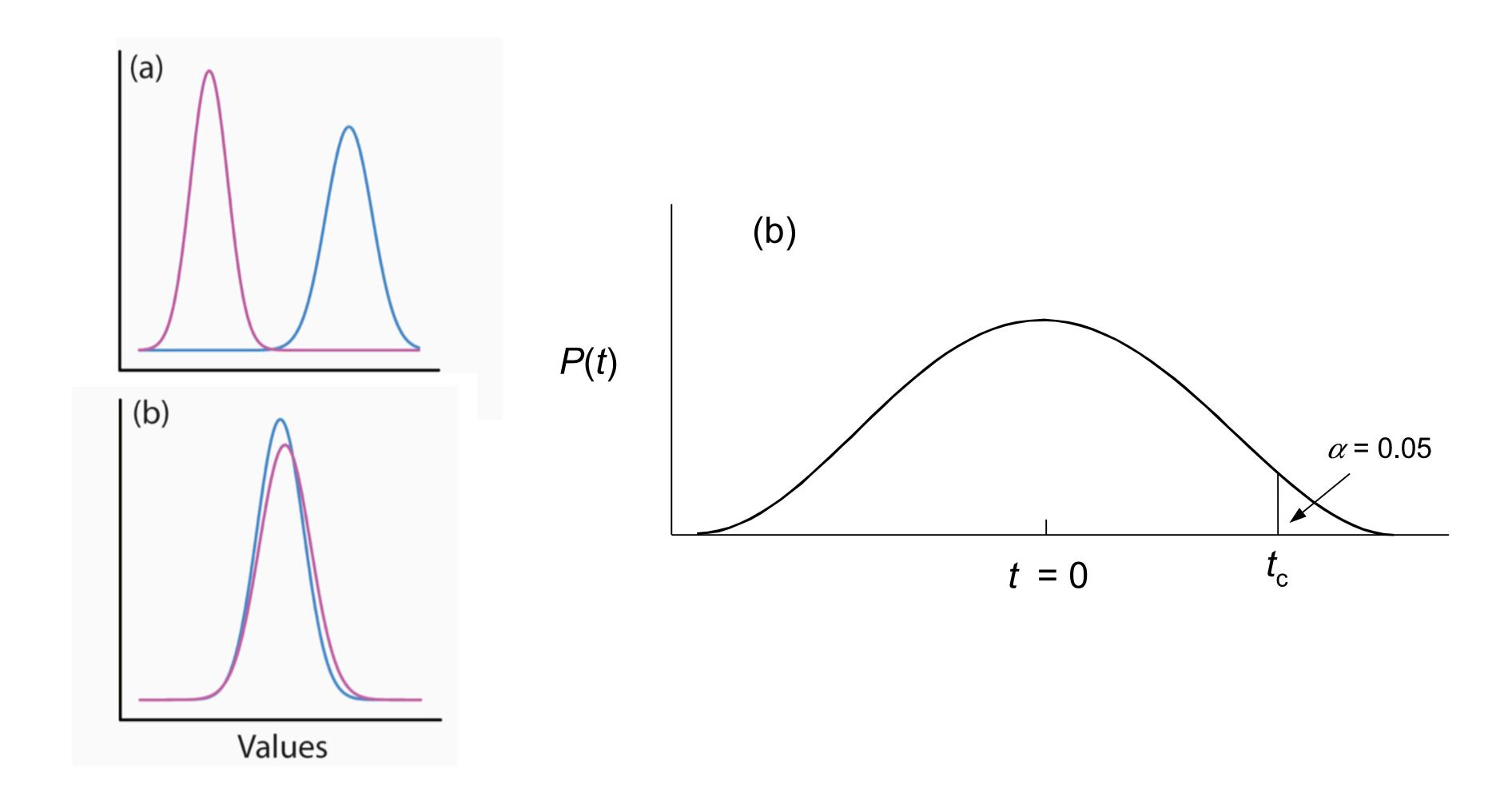
Foundational Statistics Comparing two means: the *t*-test and nonparametric alternatives



What p-values don't tell us

1. Whether the observed effect is meaningful in the context of your study system.

2. Whether your data meet the assumptions of the statistical test.

3. The level of absolute plausibility of your hypothesis. (the results of your hypothesis tests have to be evaluated in the context of prior and similar work)

Comparing 2 means

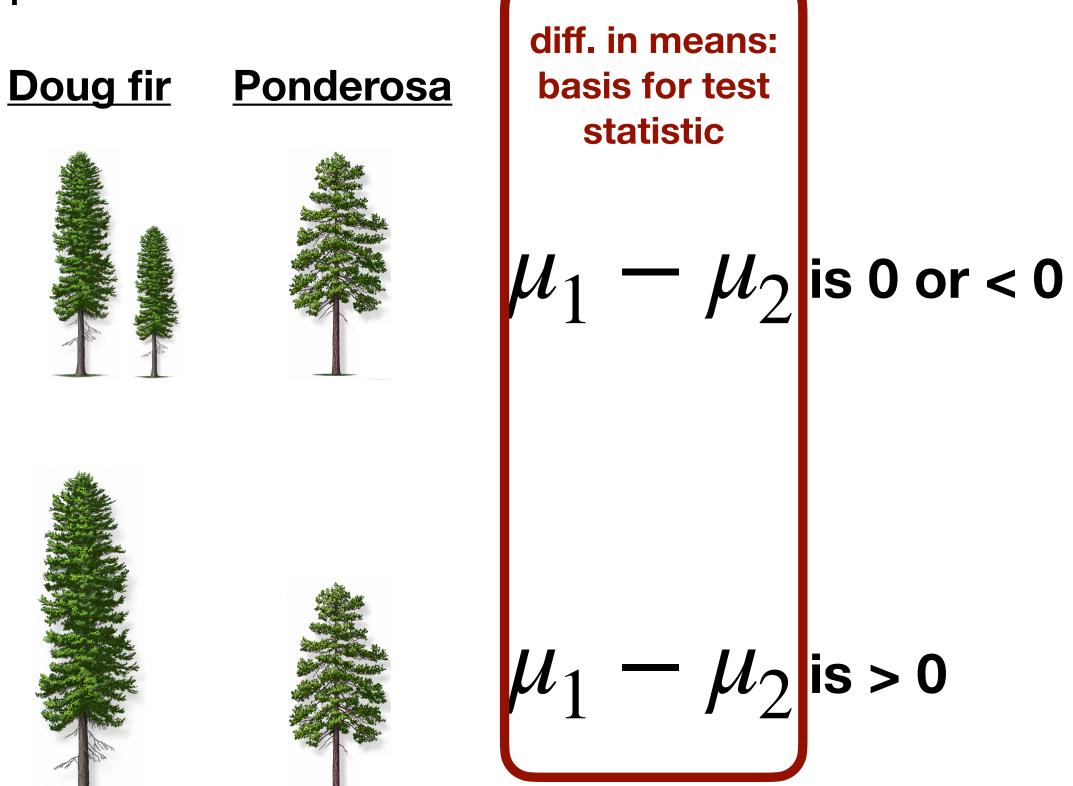
Null Hypothesis (H_0): Douglas fir trees are **not taller**, on average, than Ponderosa pine trees.

Alternative Hypothesis (H_A): Douglass fir trees are taller, on

average, than Ponderosa pine trees.

$$H_0: \mu_1 \leq \mu_2$$

$$H_A: \mu_1 > \mu_2$$



The t-statistic and its sampling distribution

$$H_0: \mu_1 \leq \mu_2 \qquad \qquad H_A: \mu_1 > \mu_2$$
 Under $H_0: t \gg 0$ Under $H_A: t \gg 0$

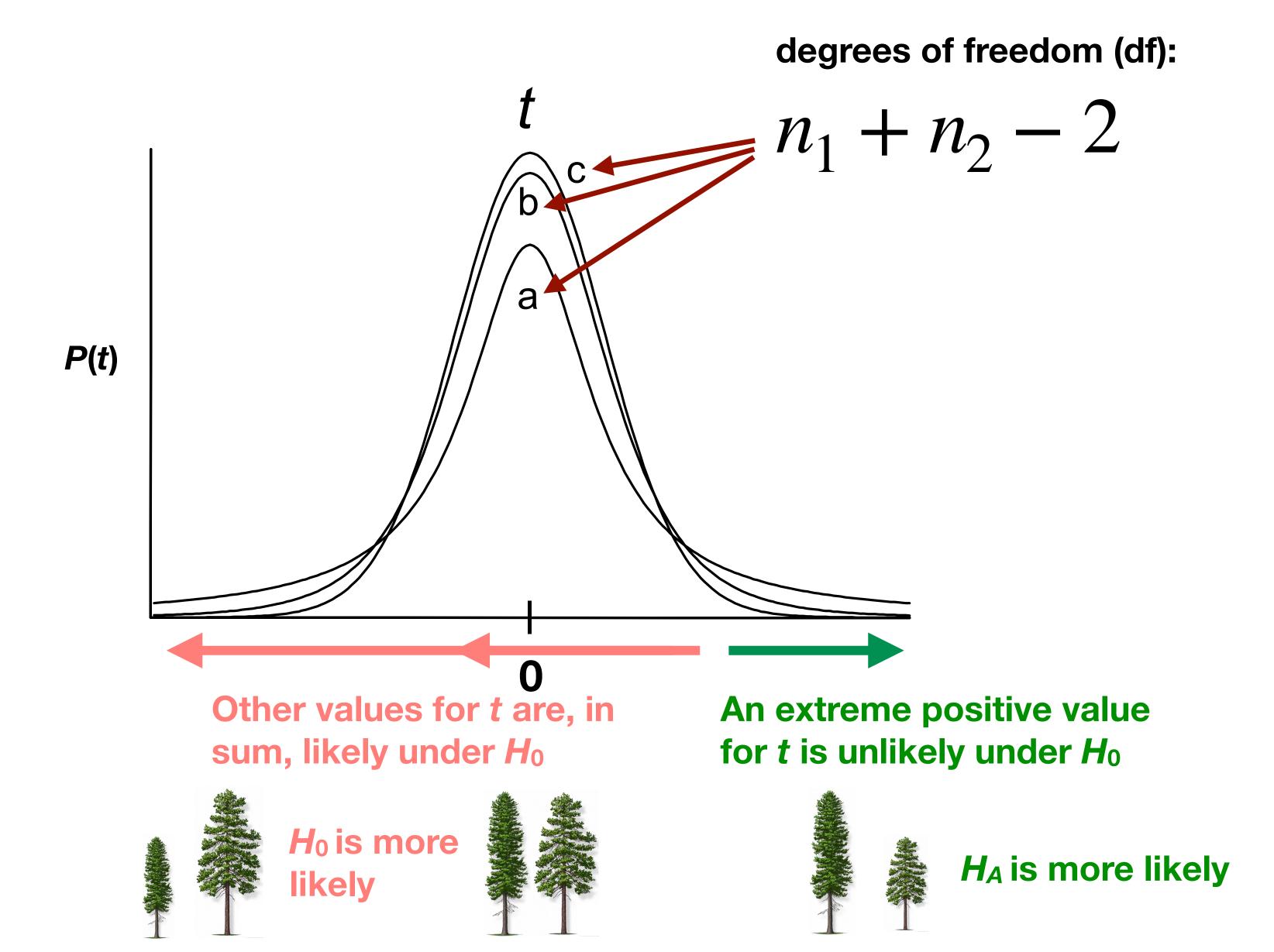
We can use the t-statistic to test these hypotheses

$$t = \frac{(\overline{y}_1 - \overline{y}_2)}{S_{\overline{y}_1 - \overline{y}_2}}$$

$$s_{\bar{y}_1 - \bar{y}_2} = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}} (\frac{1}{n_1} + \frac{1}{n_2})$$
 (Calculation for the std. error of mean difference)

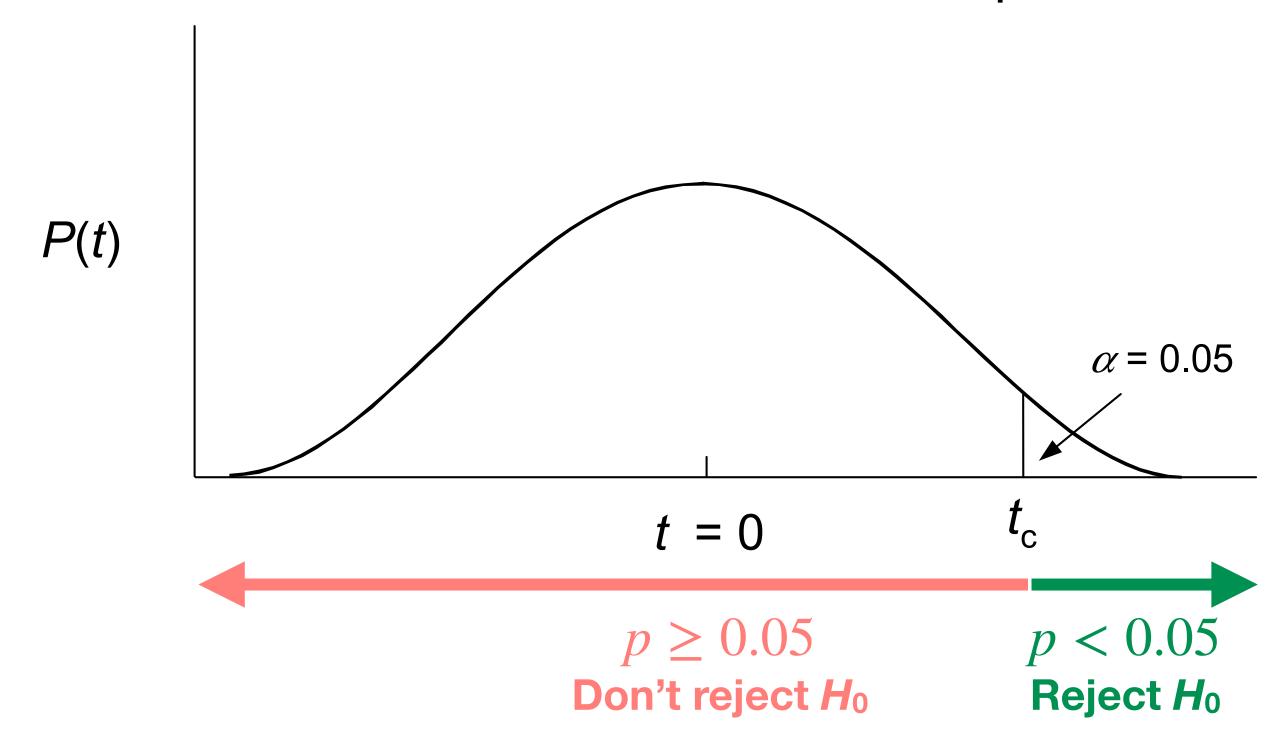
Let's think about the sampling process for our t-statistic

The *t*-statistic and *t*-distribution



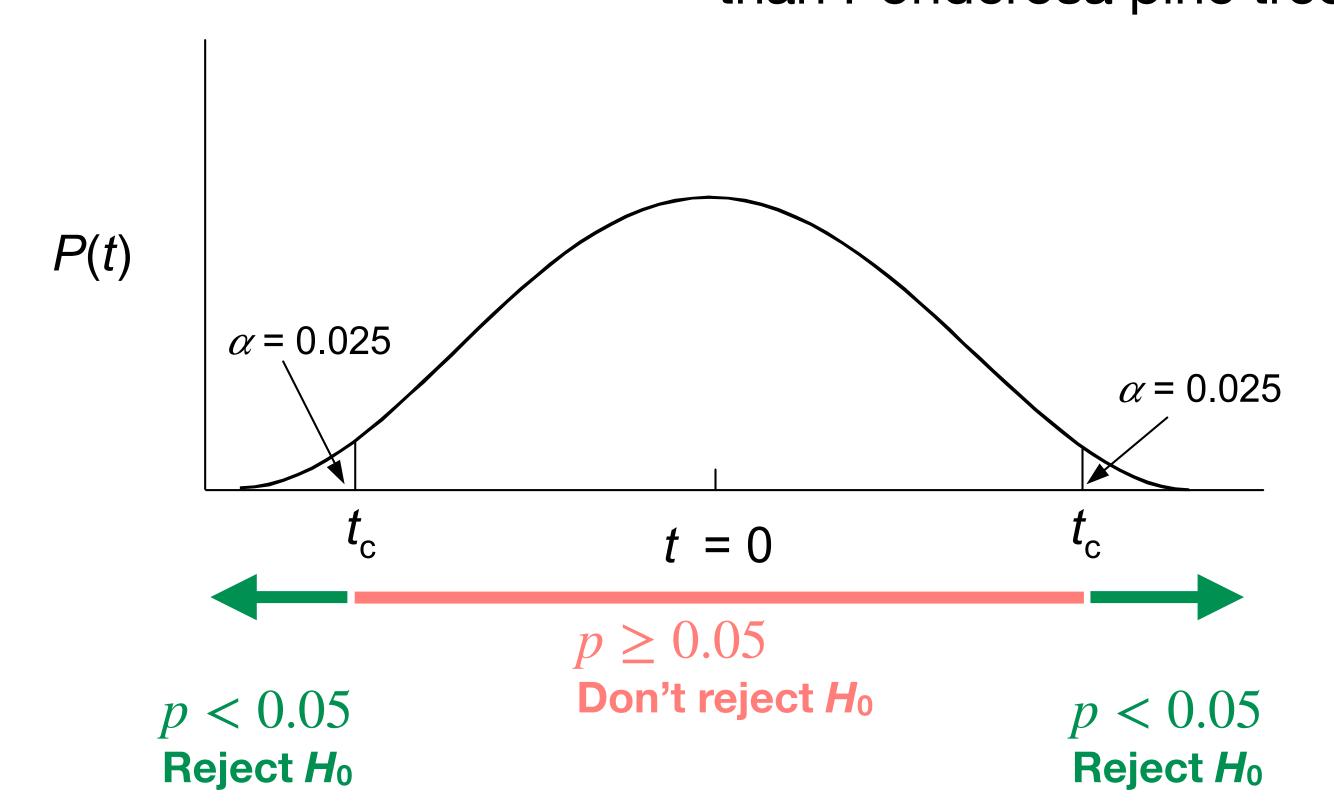
The *t*-statistic and two-sample *t*-test (**one-sided** test)

$$H_0: \mu_1 \leq \mu_2 \qquad H_A: \mu_1 > \mu_2$$
 Douglass fir trees are **taller** than Ponderosa pine trees



The *t*-statistic and two-sample *t*-test (**two-sided** test)

$$H_0: \mu_1 = \mu_2 \qquad H_A: \mu_1 \neq \mu_2$$
 Douglass fir trees are **taller or shorter** than Ponderosa pine trees



Assumptions for parametric *t*-tests that your sample(s) should meet

- The theoretical *t*-distributions for each degree of freedom were calculated for variables that meet the following criteria:
- 1) normally distributed in both populations
- 2) have approximately equal variances between populations
- 3) samples are random, composed of independent observations
- Non-normality?: can use nonparametric tests such as Mann-Whitney-Wilcoxon, or randomization / resampling tests

Randomization to generate a "null" distribution for *t*

Example: The number of successful broods of pseudoscorpion females that were mated twice to either a single male (SM) or two different males (DM).

SM: 65442342420201260233

DM: 2026434427416364

Calculate the observed *t*-statistic: $t = \frac{(\overline{y}_1 - \overline{y}_2)}{S_{\overline{y}_1 - \overline{y}_2}} = -1.39$

Steps of the randomization test:

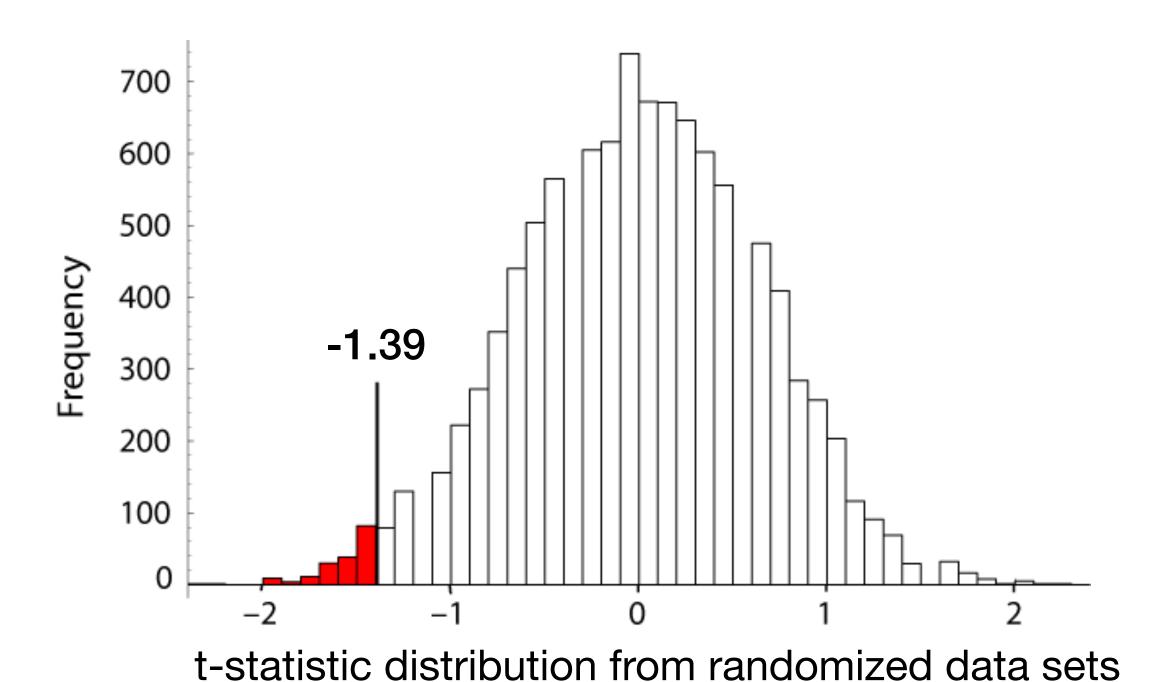
1. Create a randomized data set in which the measurements are randomly reassigned (shuffled) to the two groups (color coding indicates original treatment assigned)

SM: 40742221640334624600

DM: 2233243442146625

- 2. Calculate the *t*-statistic for the randomized sample: $t = \frac{(\overline{y}_1 \overline{y}_2)}{S_{\overline{y}_1 \overline{y}_2}} = -0.485$
- 3. Repeat steps 1 and 2 many times to get an empirical null distribution for t

Randomization to generate a "null" distribution for *t*



The null distribution of the test statistic from 10,000 replications of the randomization process. (-1.39 is the observed value from the data)

Of these 10,000 randomizations, only 176 had a test statistic equal to or less than the observed value, -1.39, so p = 176/10000 = 0.0176

Reject H₀ for a one-sided test: Singly-mated (SM) females have fewer successful broods than multiply mated (DM) females

Resampling (bootstrapping) to generate a "null" distribution for *t*

- 1. Use the computer to take two random samples of the data from all observations (with replacement, same sample sizes as 2 groups).
- 2. Calculate a "t statistic" using the sample pair from step 1.
- 3. Repeat steps 1 and 2 a very large number of times (≥1000)
- 4. Calculate the proportion of bootstrap replicates with a value at least as extreme as your actual t from the two observed samples.

This is the p-value for a one-sided test