

# Week 7: Tests for Differences

## Session 1

Spring 2020

# iClicker Question 1

Scenario:

- ▶ You want to know whether the growth rates of oak trees differ in two habitats.
- ▶ You have two clones each of 37 different genotypes.
- ▶ One clone of each genotype is planted in a mesic prairie, the other in an old agricultural field.
- ▶ Which test should you use?

- A two-sample t-test
- B Mann-Whitney paried test
- C Mann-Whitney U test
- D paired t-test

## iClicker Question 2

Scenario:

- ▶ You are testing whether Pennsylvania sedge (*C. pennsylvanica*) cover is greater under hardwoods or conifers.
- ▶ You have collected percent cover data from 50 hardwood and 50 conifer patches at Mount Toby.
- ▶ Which test should you use?

- A two-sample t-test
- B Mann-Whitney paried test
- C Mann-Whitney U test
- D paired t-test

## iClicker Question 3

Scenario:

-You want to know whether the growth rates of oak trees differ in two habitats.

-You think they will grow more quickly in a old agricultural field than in a mesic prairie.

► Which is the best alternative hypothesis?

A growth is faster in the prairie

B there is no difference in growth between habitats

C growth is faster in the field

D there is a difference in growth between habitats

E growth rates are the same

## iClicker Question 4

Scenario:

-You want to know whether the growth rates of oak trees differ in two habitats: an old agricultural field and mesic prairie.

► Which is the best alternative hypothesis?

A growth is faster in the prairie

B there is no difference in growth between habitats

C growth is faster in the field

D there is a difference in growth between habitats

E growth rates are the same

## iClicker Question 5

Scenario:

- ▶ You calculate a t-value, and an associated p-value, for individual length in two populations of *Daphnia*.
- ▶ Which critical significance category for the t-value would give you the best evidence that the two populations are different:

- A 1%
- B 10%
- C 5%
- D 95%
- E 0.1%

If you're feeling stuck, remember my office hours are Tuesday/Thursday 1:00 - 2:00.



# For Today

- ▶ Toward statistics
- ▶ Tests for differences

Follow-up questions from the Chapter 6 homework

- ▶ What questions do you have?



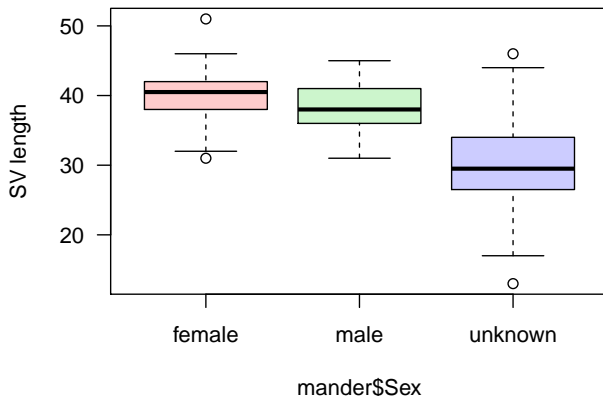
# Beyond graphs, Towards statistics

- ▶ Graphs are powerful tools that provide insight and understanding of the patterns and relationships in the data.
- ▶ Graphs alone don't give us the complete answer. We need to **quantify** the relationships we see in our plots.
- ▶ What other tools do we have to **support** our conclusions?



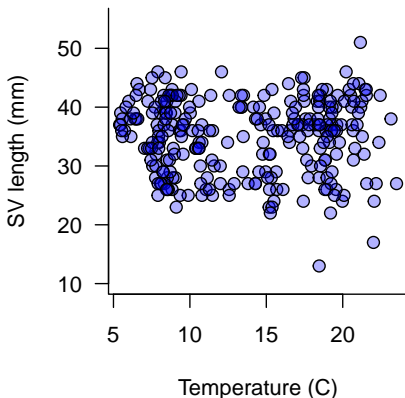
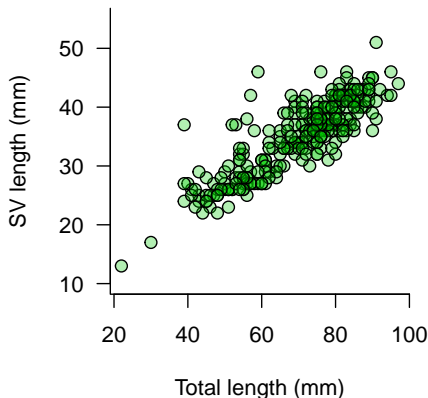
# Beyond graphs, Towards statistics

- How can we **quantify** our evidence for relationships?
  - Are differences between groups *significant*?
  - Are differences between groups *meaningful*?



# Beyond graphs, Towards statistics

- How can we **quantify** our evidence for relationships?
  - Are associations between 2 variables *significant*?
  - Are associations between 2 variables *meaningful*?



# Beyond graphs, Towards statistics

- ▶ Statistics is the tool we use to formally answer these questions!
  - ▶ Are differences *are/are not* significant?
  - ▶ Are associations *are/are not* significant?

Wait a second... what do we mean when we say **significant**?



# Let's examine some plots to gain intuition:

- Scenario: We want to know whether the size of 3-year-old bluegill (*Lepomis macrochirus*) are larger in some Massachusetts lakes than others.
- We have collected data for bluegill from Wyola Lake and the Quabbin Reservoir in Western Mass.

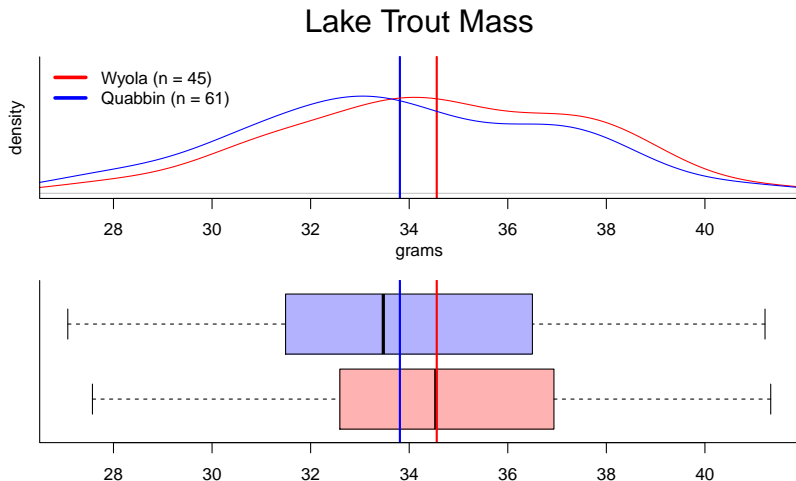


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<sup>1</sup>Image credit: New York Fish and Game Commission

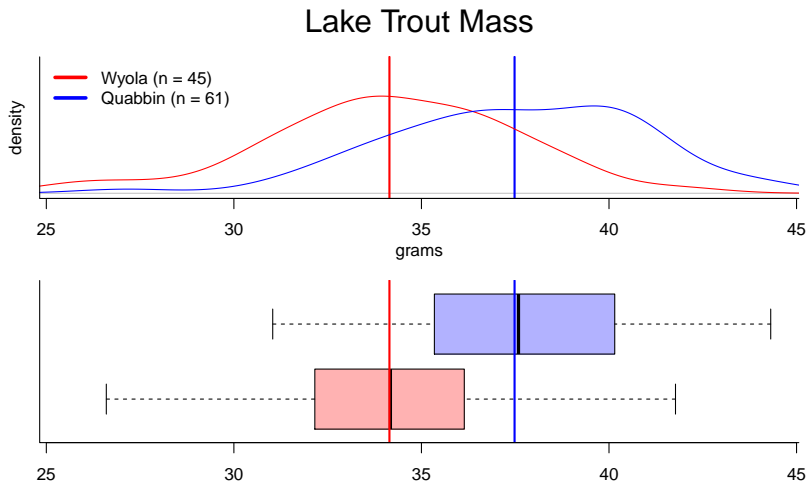
# Bluegill Data I

- ▶ Are differences between lakes *significant*?
- ▶ Are differences between lakes *meaningful*?



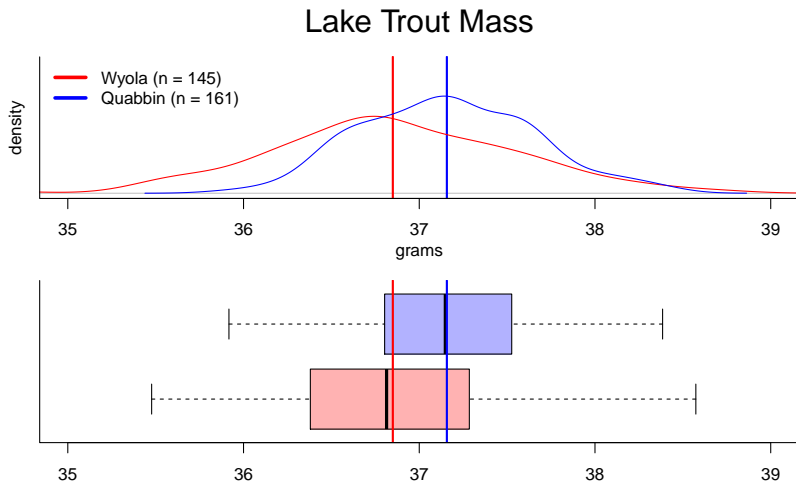
# Bluegill Data II

- ▶ Are differences between lakes *significant*?
- ▶ Are differences between lakes *meaningful*?



# Bluegill Data III

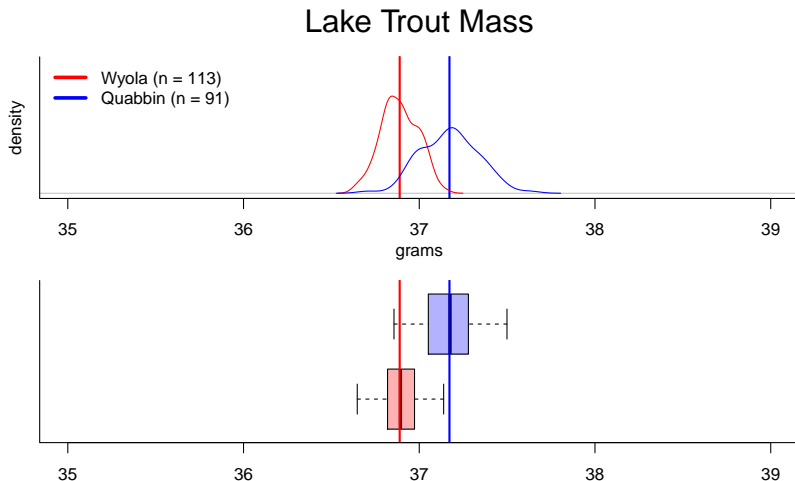
- ▶ Are differences between lakes *significant*?
- ▶ Are differences between lakes *meaningful*?





# Bluegill Data IV

- ▶ Are differences between lakes *significant*?
- ▶ Are differences between lakes *meaningful*?



# Tests for differences

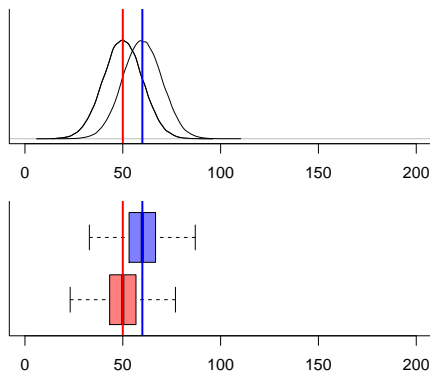
- ▶ How can we test if two groups of observations are different?
- ▶ What kinds of tests do we know about?
- ▶ Which statistic(s) do the tests test for?



# Tests for differences

Often we want to know if two of more samples are different

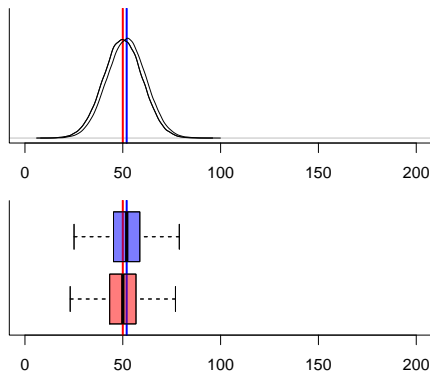
- ▶ are the sample *means* different?
- ▶ are the sample *medians* different?
- ▶ are the differences *statistically significant*?



# Tests for differences

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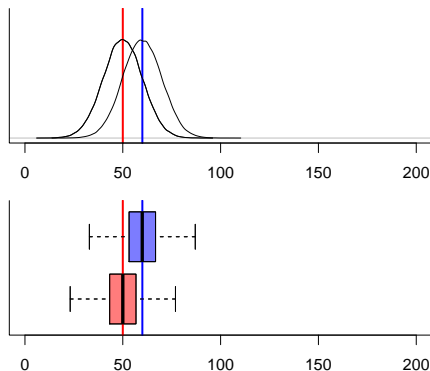
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# Tests for differences

Often we want to know if two of more samples are different

- ▶ are the sample *means* different?
- ▶ are the sample *medians* different?
- ▶ are the differences *statistically significant*?

To determine the significance of differences between **two**, we need a statistical test

- ▶ *t-test*
- ▶ *U-test*

# Tests for differences: intuition

- ▶ What information would we need to know?
- ▶ What kinds of evidence would support our conclusion?
- ▶ How do we define *different*?

- ▶ Let's draw some distributions:

## Tests for differences: intuition with R



# Differences: t-test

Purpose:

- ▶ compare the means of two samples (say  $a$  and  $b$ )

Assumptions:

- ▶ both samples normally distributed
- ▶ both samples have equal variances

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$$t = \frac{|\bar{x}_a - \bar{x}_b|}{\sqrt{\frac{s_a^2}{n_a} + \frac{s_b^2}{n_b}}}$$

- ▶  $t$ : the  $t$ -statistic
- ▶  $\bar{x}$ : sample mean
- ▶  $s$ : sample standard deviation
- ▶  $n$ : sample size

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- ▶ if  $|\bar{x}_a - \bar{x}_b|$  is large, then  $t$  is ????
- ▶ if  $\sqrt{\frac{s_a^2}{n_a} + \frac{s_b^2}{n_b}}$  is large, then  $t$  is ????

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- ▶ compare the means of two samples (say  $a$  and  $b$ )

Assumptions:

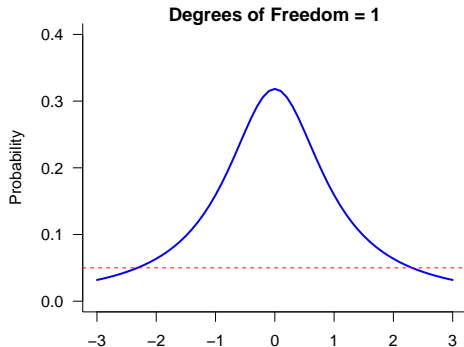
- ▶ both samples normally distributed
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- ▶ if  $|\bar{x}_a - \bar{x}_b|$  is large, then  $t$  is **large**
- ▶ if  $\sqrt{\frac{s_a^2}{n_a} + \frac{s_b^2}{n_b}}$  is large, then  $t$  is **small**

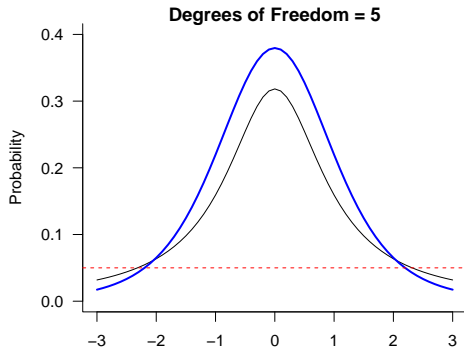
# Differences: t-test

Understanding the *t-distribution*:



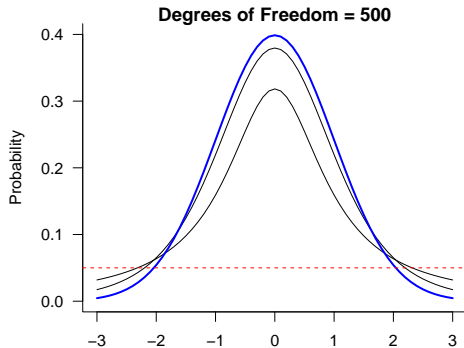
# Differences: t-test

Understanding the *t*-distribution:



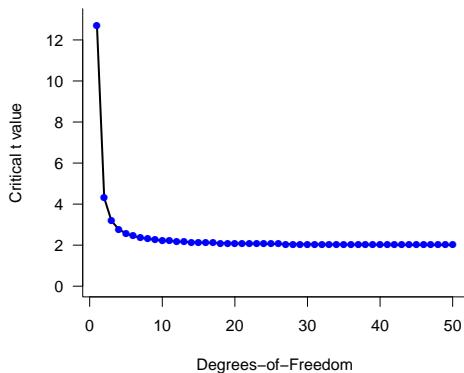
# Differences: t-test

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# Differences: t-test

Understanding the *t-distribution*:

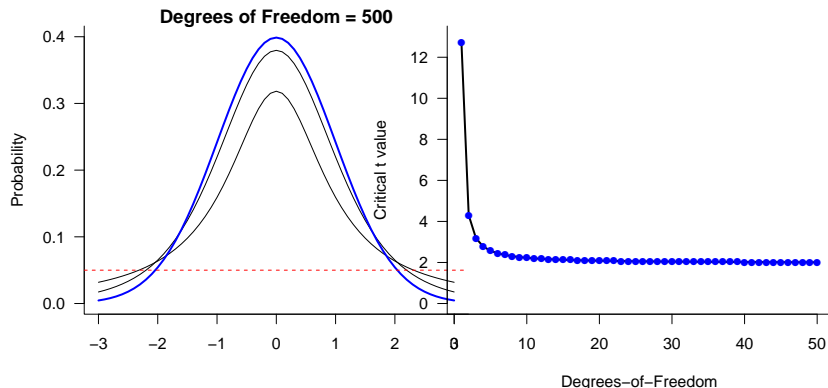




# Differences: t-test

Understanding the *t-distribution*:

- ▶ whether a difference is significant depends on:
  - ▶ the *t-statistic*
  - ▶ degrees-of-freedom ( $n_a - 1 + n_b - 1$ )
- ▶ larger *t-statistics* more likely to be significant



# Differences: t-test

Understanding the *p-value*:

- ▶ *p-value* is the probability of observing a *t-statistic* as high as we did by chance
- ▶ if *p-value* is lower than significance level (e.g. 5%):
  - ▶ difference is significant
  - ▶ reject the null hypothesis
  - ▶ accept the alternative hypothesis

# Differences: t-test

Which *t-test*?

- ▶ standard *t-test*
  - ▶ compare two independent samples
  - ▶ both normally distributed
  - ▶ equal (similar) variances
  - ▶ samples sizes can be the same or not

$$t = \frac{|\bar{x}_a - \bar{x}_b|}{\sqrt{\frac{s_a^2}{n_a} + \frac{s_b^2}{n_b}}}$$

- ▶  $t$ : the *t*-statistic
- ▶  $\bar{x}$ : sample mean
- ▶  $s$ : sample standard deviation
- ▶  $n$ : sample size

# Differences: paired t-test

Sometimes samples are not independent

- ▶ compare pairs of samples
  - ▶ e.g., before-after
  - ▶ e.g., north-south
  - ▶ e.g., left-right
- ▶ both normally distributed
- ▶ equal (similar) variances
- ▶ samples sizes *must* be the

# Differences: paired t-test

Which *t-test*?

- ▶ paired *t-test*
  - ▶ compare pairs of samples
  - ▶ both normally distributed
  - ▶ equal (similar) variances
  - ▶ samples sizes are \_\_\_\_\_ ?

$$t = \frac{\bar{D}}{\sqrt{\frac{s_D^2}{n}}}$$

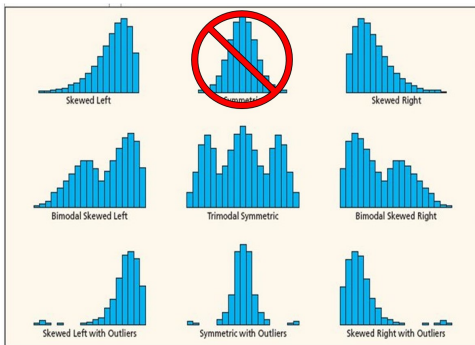
- ▶ *t*: the *t*-statistic
- ▶  $\bar{D}$ : mean of the *differences*
- ▶ *s*: standard deviation of the *differences*
- ▶ *n*: number of *paired* samples

Differences: When might the t-test be inappropriate?



# Differences: U-test

- ▶ compare two samples
- ▶ one or both *not* normally distributed
- ▶ based on *median*, *range*, and *ranks*
- ▶ rank all values as one sample, calculate group rank sums  $R$
- ▶ calculate a  $U$ -value, a measure of overlap



## Differences: U-test

- ▶ compare two samples
- ▶ both or differences *not* normally distributed
- ▶ based on *median*, *range*, and *ranks*
- ▶ rank all values as one sample, calculate group rank sums  $R$
- ▶ calculate a  $U$ -value, a measure of overlap

$$U_a = n_a \times n_b + \frac{n_a(n_a + 1)}{2} - R_a$$

$$U_b = n_b \times n_a + \frac{n_b(n_b + 1)}{2} - R_b$$

- ▶  $n_a$ : number of samples in sample  $a$
- ▶  $n_b$ : number of samples in sample  $b$
- ▶  $R_a$ : sum of the ranks of values in  $a$
- ▶  $R_b$ : sum of the ranks of values in  $b$



## Differences: U-test

- ▶ compare two samples
- ▶ both or differences *not* normally distributed
- ▶ based on *median*, *range*, and *ranks*
- ▶ rank all values as one sample, calculate group rank sums  $R$
- ▶ calculate a  $U$ -value, a measure of overlap

$$U_a = n_a \times n_b + \frac{n_a(n_a + 1)}{2} - R_a$$

$$U_b = n_b \times n_a + \frac{n_b(n_b + 1)}{2} - R_b$$

- ▶ smallest is used to find the  $p$ -value
- ▶ unlike the t-statistic, lower  $U$ -values are more likely to be significant

## Differences: Wilcoxon matched-pairs test

- ▶ both or differences *not* normally distributed
- ▶ based on ranked *differences*
  - ▶ first calculate the differences
  - ▶ second rank the differences
  - ▶ 0's not ranked
- ▶ sum and compare +ve and -ve ranks

$$W^+ = \sum R^+$$

$$W^- = \sum R^-$$

- ▶  $W^+$ : the Wilcoxon test statistic for positive differences
- ▶  $W^-$ : the Wilcoxon test statistic for negative differences
- ▶  $R^+$ : the sum of the ranks of positive differences
- ▶  $R^-$ : the sum of the ranks of negative differences

# Differences: Wilcoxon matched-pairs test

- ▶ pairs or differences *not* normally distributed
- ▶ based on ranked *differences*
  - ▶ first calculate the differences
  - ▶ second rank the differences
  - ▶ 0's not ranked
- ▶ sum and compare +ve and -ve ranks

$$W^+ = \sum R^+$$

$$W^- = \sum R^-$$

- ▶ smallest is used to find the  $p$ -value
- ▶ lower  $W$ -values are more likely to be significant

# Group Assignment

Using the whale count data, complete the group assignment on Moodle.