

# EGM101 – Skills Toolbox

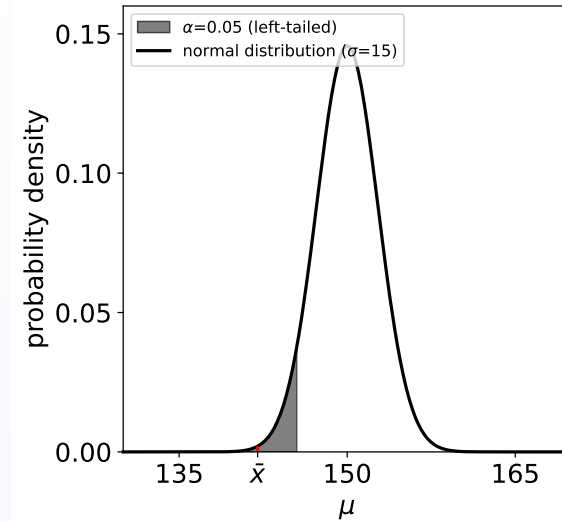
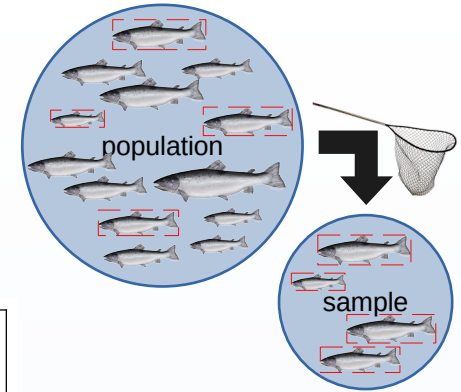
Week 8, Part 3: Parametric tests

# What is a parametric test?

- **Parametric test:** assume population can be modeled by a probability distribution with fixed *parameters*
  - e.g., normal distribution
  - Example: Z-test
  - Mostly going to focus on Student's *t*-distribution
- **Non-parametric test:** don't assume a probability distribution for the population

population:  $\mu = 150$  cm  
 $\sigma = 15$  cm

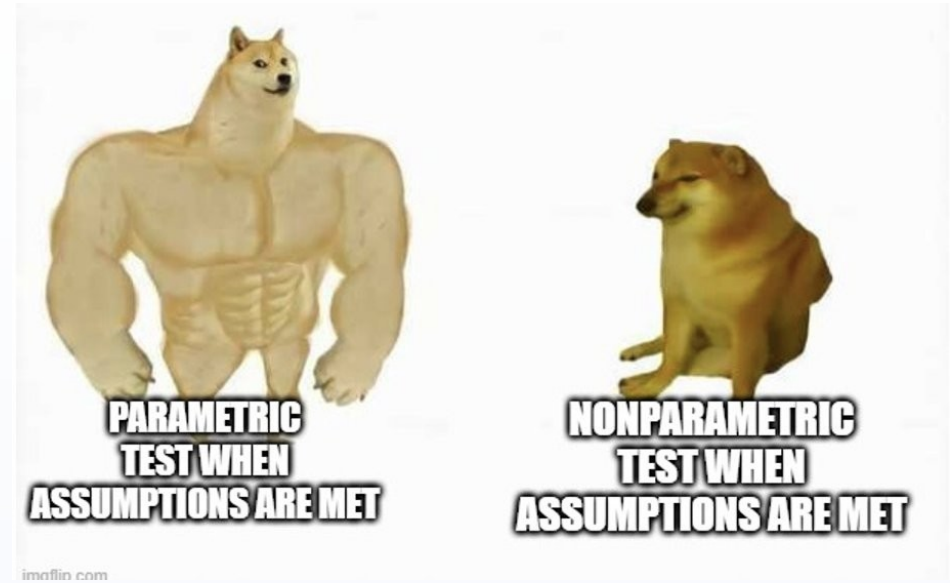
sample ( $n = 30$ ):  $\bar{x} = 142$  cm  
 $s = 20$  cm



$$\begin{aligned}
 z &= \frac{\bar{x} - \mu}{\sigma_{\bar{x}}} \\
 &= \frac{142 - 150}{15/\sqrt{30}} \\
 &= -2.921
 \end{aligned}$$

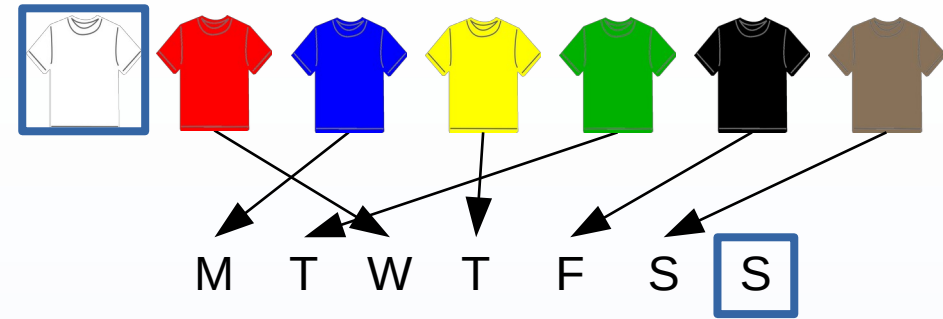
# When do I use parametric tests?

- When we are working with normally-distributed, continuous data
  - Remember the Central Limit Theorem!
  - If in doubt, check the data!
- When our observations are independent
- When the variance of the samples is approximately the same
- In general, prefer parametric tests because they tend to have more **power**



# Degrees of Freedom

- Remember:  $\bar{x}$ ,  $s$  are *estimates* of  $\mu$ ,  $\sigma$
- Degrees of freedom** ( $\nu$ ): the number of *independent* pieces of information that are used to calculate an estimate
  - Example: wearing a different t-shirt every day during the week
  - Only 6 *independent* choices available
- Alternative definition: the number of observations that are free to vary
- Example: calculating sample mean
  - For a given value of  $\bar{x}$ , only  $n - 1$  values are “free” to vary
  - In practice, for mean:  $\nu = n - 1$



$$\bar{x} = 6$$

|   |   |    |   |   |    |    |   |   |   |
|---|---|----|---|---|----|----|---|---|---|
| 2 | 1 | 13 | 3 | 7 | 10 | 10 | 7 | 4 | 3 |
|---|---|----|---|---|----|----|---|---|---|

$$\sum_{i=1}^9 x_i = 57 \rightarrow x_{10} = 3$$

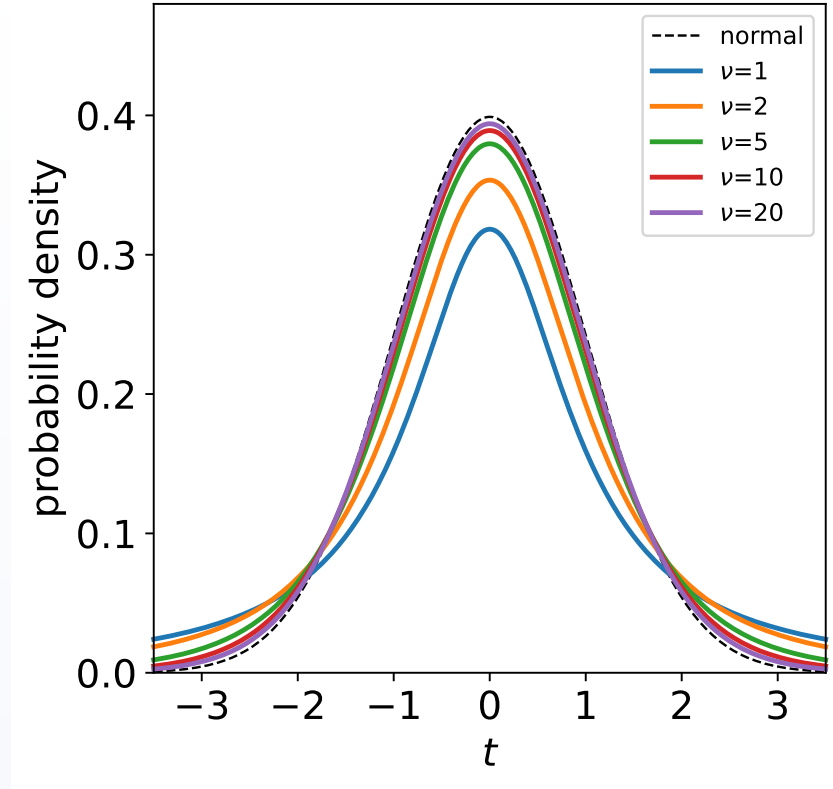
$$\bar{x} = 10$$

|   |   |    |    |    |    |    |   |   |   |
|---|---|----|----|----|----|----|---|---|---|
| 2 | 7 | 19 | 18 | 12 | 10 | 16 | 3 | 4 | 9 |
|---|---|----|----|----|----|----|---|---|---|

$$\sum_{i=1}^9 x_i = 91 \rightarrow x_{10} = 9$$

# Student's $t$ -Distribution

- Similar to normal distribution, but takes degrees of freedom ( $\nu$ ) into account
  - Shorter peak, larger tails
- As  $\nu$  increases, distribution approaches normal distribution
  - Above  $\nu=20$  or so, essentially the same as the normal distribution
- Use this distribution when:
  - We have a small sample size
  - When we don't know population standard deviation (most of the time)

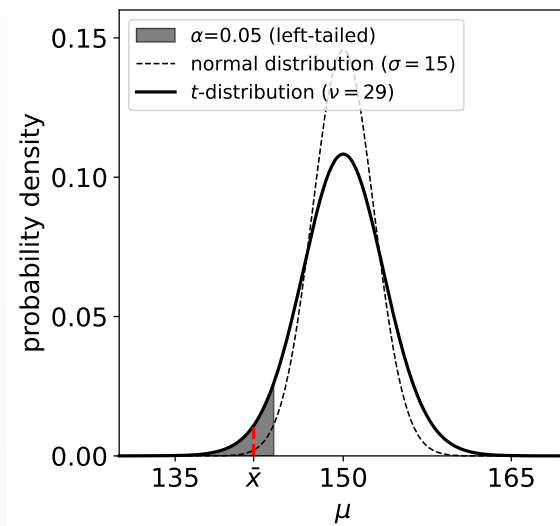
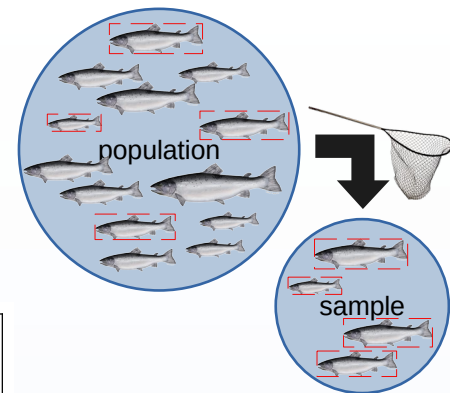


# Student's $t$ -Test for a single mean

- Proceed exactly like  $z$ -test
  - Instead of normal distribution, use  $t$ -distribution with  $\nu = n-1$
- Example: salmon length
- Hypotheses:
  - $H_0: \mu = 150$  cm
  - $H_a: \mu < 150$  cm
- Critical  $t$  for  $\nu = 29$ ,  $\alpha = 0.05$ :  $-1.699$ 
  - $t < t_c$  ( $p < \alpha$ )  $\rightarrow$  reject  $H_0$
  - Based on the evidence,  $\mu$  is most likely less than 150 cm.

population:  $\mu = 150$  cm

sample ( $n = 30$ ):  $\bar{x} = 142$  cm  
 $s = 20$  cm



$$\begin{aligned}
 t &= \frac{\bar{x} - \mu}{s_{\bar{x}}} \quad \text{standard error of the mean} \\
 &= \frac{142 - 150}{20/\sqrt{30}} \\
 &= -2.19 \\
 p &= 0.018
 \end{aligned}$$

# Two samples

- With two samples:

- Comparison is between the difference of means
- Divided by estimate of the standard error of the mean,  $s_{\bar{x}}$
- Depends on whether samples are paired or not

- Paired samples:

- Samples are dependent
- Each measurement in one sample is paired with a measurement in the other sample
- Useful for observing changes over time

- Unpaired samples:

- Samples are independent
- Each individual in one sample is independent of every individual in the other sample
- Useful for comparing different populations

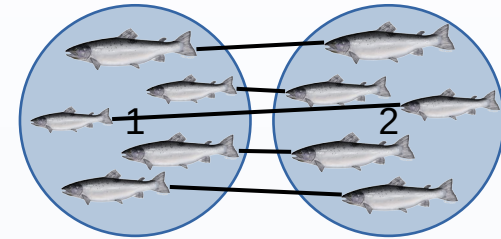
$$t = \frac{\overline{x}_1 - \overline{x}_2 - (\mu_1 - \mu_2)}{s_{\bar{x}}}$$

sample difference

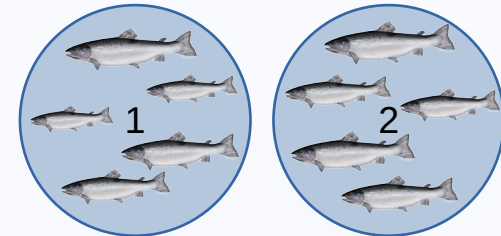
hypothesized difference

estimate of the standard error of the mean

paired:



unpaired:



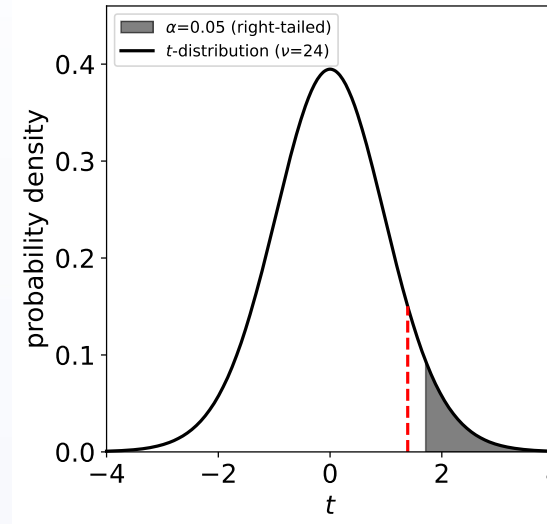
# Paired samples

- Example: salmon weighed before, after migration ( $n=25$ )
  - Before: 3.5 kg; After: 3.25 kg;
  - $s_d = 0.9$  kg
- Hypotheses:
  - $H_0: \mu_1 = \mu_2$
  - $H_a: \mu_1 > \mu_2$
- At  $\alpha = 0.05$ , is there a decrease in the population mean?
  - Right-tailed ( $\mu_1 > \mu_2$ , or  $\mu_1 - \mu_2 > 0$ )
  - Degrees of freedom:  $\nu = n - 1 = 24$
  - Critical value,  $t_c = 1.711$
- Because  $t < t_c$  ( $p > \alpha$ ), we fail to reject  $H_0$

$$t = \frac{\overline{x}_1 - \overline{x}_2 - (\mu_1 - \mu_2)}{s_d / \sqrt{n}}$$

sample difference      hypothesized difference

standard deviation of differences



$$\overline{x}_1 = 3.5 \quad \overline{x}_2 = 3.25$$

$$s_d = 0.9$$

$$t = \frac{\overline{x}_1 - \overline{x}_2}{s_d / \sqrt{n}}$$

$$= \frac{3.5 - 3.25}{0.9 / \sqrt{25}} = 1.389$$

$$p = 0.089$$

# Unpaired samples

- For unpaired samples, form of  $s_{\bar{x}}$  depends on population variances
- If variances are “similar”:
  - pooled variance,  $s_p^2$
  - $v = n_1 + n_2 - 2$
- If variances aren’t “similar”:
  - “Welch’s t-test”
  - $v$ : it’s complicated...
- Note: don’t memorize the formulas, but understand when to use which form

$$\frac{1}{2} < \frac{s_1}{s_2} < 2:$$

$$t = \frac{\bar{x}_1 - \bar{x}_2 - (\mu_1 - \mu_2)}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \quad s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$$

$$s_1 > 2s_2 \quad \text{or} \quad s_2 > 2s_1:$$

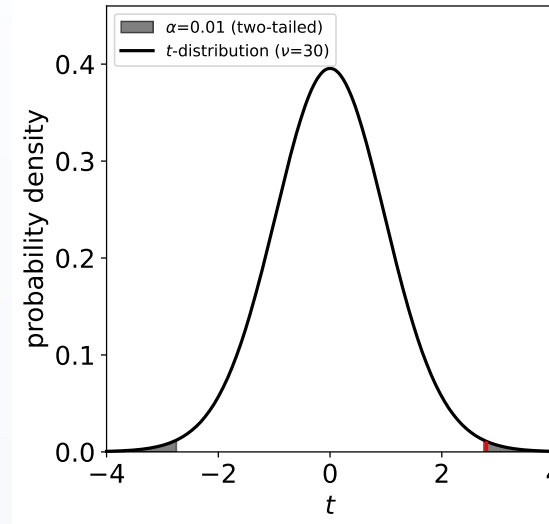
$$t = \frac{\bar{x}_1 - \bar{x}_2 - (\mu_1 - \mu_2)}{s_{\bar{\Delta}}} \quad s_{\bar{\Delta}} = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

# Independent samples

- Example: effect of wastewater discharge on fish health
  - Upstream ( $\mu_1$ ): sample of 18 fish
  - Downstream ( $\mu_2$ ): sample of 14 fish
- Hypotheses:
  - $H_0: \mu_1 = \mu_2$
  - $H_a: \mu_1 \neq \mu_2$
- At  $\alpha = 0.01$ , is there a difference in length between the two populations?
  - Two-tailed (any difference)
  - Degrees of freedom:  $\nu = n_1 + n_2 - 2$
- Because  $t > t_c$  ( $p < \alpha$ ), we reject  $H_0$

$$t = \frac{\overbrace{\bar{x}_1 - \bar{x}_2}^{\text{sample difference}} - \underbrace{(\mu_1 - \mu_2)}_{\text{hypothesized difference}}}{\underbrace{s_p}_{\text{pooled standard deviation}} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

$$s_p = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}}$$



|                | upstream | downstream |
|----------------|----------|------------|
| $\bar{x}$ (cm) | 23.97    | 20.67      |
| $s$ (cm)       | 3.10     | 3.56       |
| $n$            | 18       | 14         |

$$t = 2.800 \quad p = 0.004$$

$$t_c = 2.749$$

- Parametric tests: when we assume something about the distribution of the population
- Most of the time, we end up using Student's  $t$ -distribution
- Calculation varies depending on:
  - How many samples (one or two)
  - Independence of samples
  - Population variances
- Remember: focus more on why, rather than the formulas

- Illowsky and Dean, Chapters 9, 10
- Caswell, Chapter 15
- Weiss, Chapters 9, 10
- What are degrees of freedom in statistics? [[Minitab Blog](#)]
- Nonparametric tests vs Parametric tests [[Jim Frost](#)]
- One-tailed and two-tailed tests [[Khan Academy](#)]
- Z-statistics vs T-statistics [[Khan Academy](#)]