

# ADJUSTING FOR MULTIPLE TESTING IN MICROBIOME DATA ANALYSIS

Research Group: Statistical Diversity Lab (new!)

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# GOAL OF SCIENCE

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- Find "truth"
  - Find an interesting result that stands up to replication
- Get papers published?



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# APPROACHES

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## Exploratory

- hypothesis generating
- "I wonder if [diet] affects [the microbiome], and if so, how..."

## Confirmatory

- hypothesis confirming
- "Does the abundance of [firmicutes] change with diet?"



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# HYPOTHESIS TESTING

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- Tests that use data to assess the "statistical significance" of a result
- Definition: the *p-value* of a test is the probability of observing a more extreme result than we did if the null hypothesis were true.
- Idea: If our results were extreme under the assumption of the null hypothesis, then maybe the hypothesis isn't supported by the data.



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# TESTING MULTIPLE HYPOTHESES

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- Extreme things happen occasionally: the longer you look, the more likely they are
- If you have no signal in your experiment, what's the probability of a p-value  $< 0.01$ ?



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# TESTING MULTIPLE HYPOTHESES

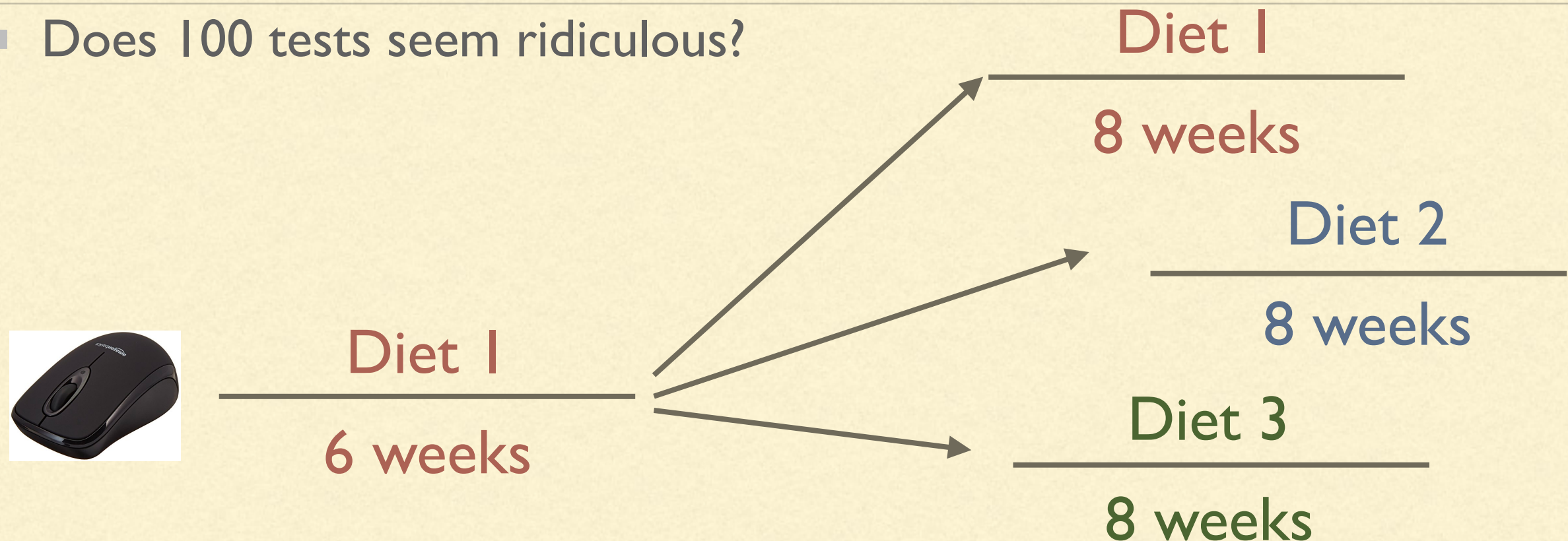
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- If you have no signal in your experiment, the probability of a p-value  $< 0.01$  is 0.01
- If you do 2 hypothesis tests, the probability that one or more has p-value  $< 0.01$  is 1.99%
- 3 tests = 3%; 10 tests = 10%, 20 tests = 18%, 100 tests = 63%



# TESTING MULTIPLE HYPOTHESES

- Does 100 tests seem ridiculous?



- 128 families + Simpson + Shannon:  $130 \times 4 = 520$  tests = 99.5% chance of finding something at 1% level *when there is no difference*
- 10 phyla + 20 families + Simp. + Shann.:  $32 \times 4 = 128$  tests = 73% chance



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# TESTING MULTIPLE HYPOTHESES

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- Two strategies for correcting for multiple comparisons
  - Family wise error rate control
  - False discovery rate control
- Both either decrease your significance threshold OR increase your p-value
- Both start with a protocol: a list of every hypothesis you're interested in



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# PROTOCOL

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- ~~"Characterize the taxonomic composition of the gut microbiome before, during and after the dietary intervention..."~~
- Write down a list of every single hypothesis you're interested in
  - Confirmatory study: maybe 10, definitely < 20
  - Exploratory study: could be hundreds, thousands



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# FAMILY WISE ERROR RATE CONTROL

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- Idea: Want to guard against any false positives
  - Very, very strict standard
- Many, many methods exist!
- Simplest method is Bonferroni: new p-values are

$$p_{\text{new}} = \min(\# \text{ tests} \times p_{\text{old}}, 1)$$



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# FALSE DISCOVERY RATE CONTROL

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- Idea: False positives inevitable, try to limit the number
  - Less strict than FWERC
- Many, many methods exist!
- Simplest method is Benjamini-Hochberg



# FALSE DISCOVERY RATE CONTROL

- Benjamini-Hochberg: Control percentage of false discoveries at 10%
- Order p-values  $p_1 \leq p_2 \leq \dots \leq p_m$
- $p_1, \dots, p_j$  are significant for largest  $j$  such that  $p_j \leq 0.1 \times j/m$

RANK	1	2	3	4	5	6	7	8	9	10
p-value	0.0008	0.0009	0.165	0.205	0.396	0.450	0.641	0.781	0.9	0.993
0.1 j/m	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09	0.1



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# EXAMPLE METHODS SECTION

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- Main: The large number of inferential tests performed in this article necessitated a multiple comparisons adjustment. Details of the adjustment procedure are available in Supplementary Statistical Methods. The procedure implies that a significance level of  $\alpha = 0.0244$  should be used to assess significant hypotheses. Throughout the article, only hypotheses that meet this threshold are described as significant.
- Supp: ...to control false discovery rate (FDR) at 5%, we employ the Benjamini-Hochberg procedure on all 38 hypotheses investigated. This count includes hypotheses that were investigated prior to preparing the manuscript...



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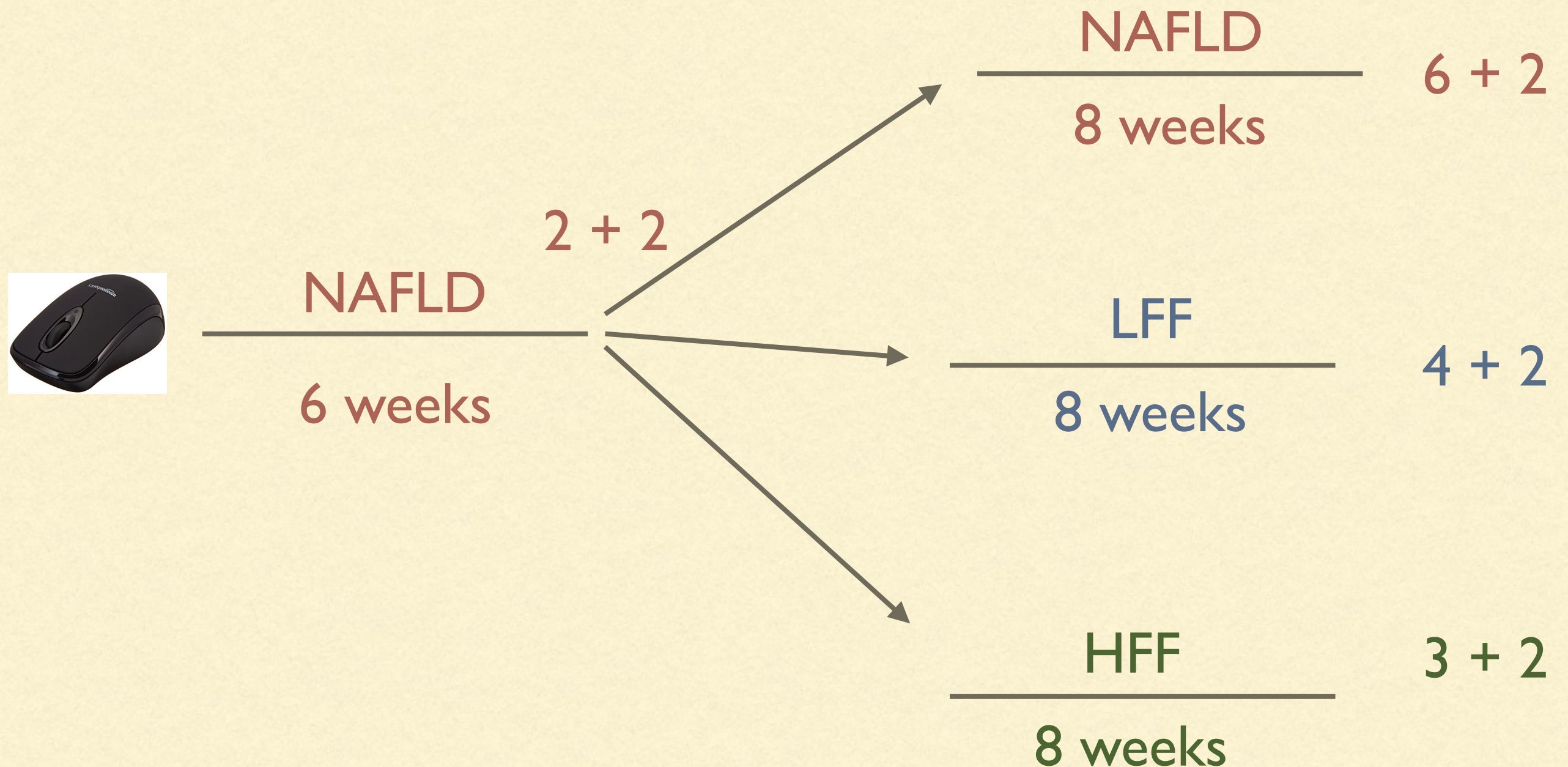
# DEMONSTRATION

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- DePaolo Lab study on non-alcoholic fatty liver disease (NAFLD)
- Mice fed a non-alcoholic fatty liver disease (NAFLD) inducing diet for 6 weeks
- Then assigned for 8 weeks to either a NAFLD diet, a HFF diet or a LFF diet
- 2 cohorts
- Ileum & fecal matter
- *Amy's analysis: "We found a significant alteration of Firmicutes in both fecal and ileum samples ( $p = 0.01$  and  $p = 0.04$ ), and a significant alteration of Verrucomicrobia in both fecal and ileum samples ( $p = 0.02$ ,  $p = 0.03$ ). We also found changes in Actinobacteria abundance in fecal matter ( $p = 0.01$ )."*

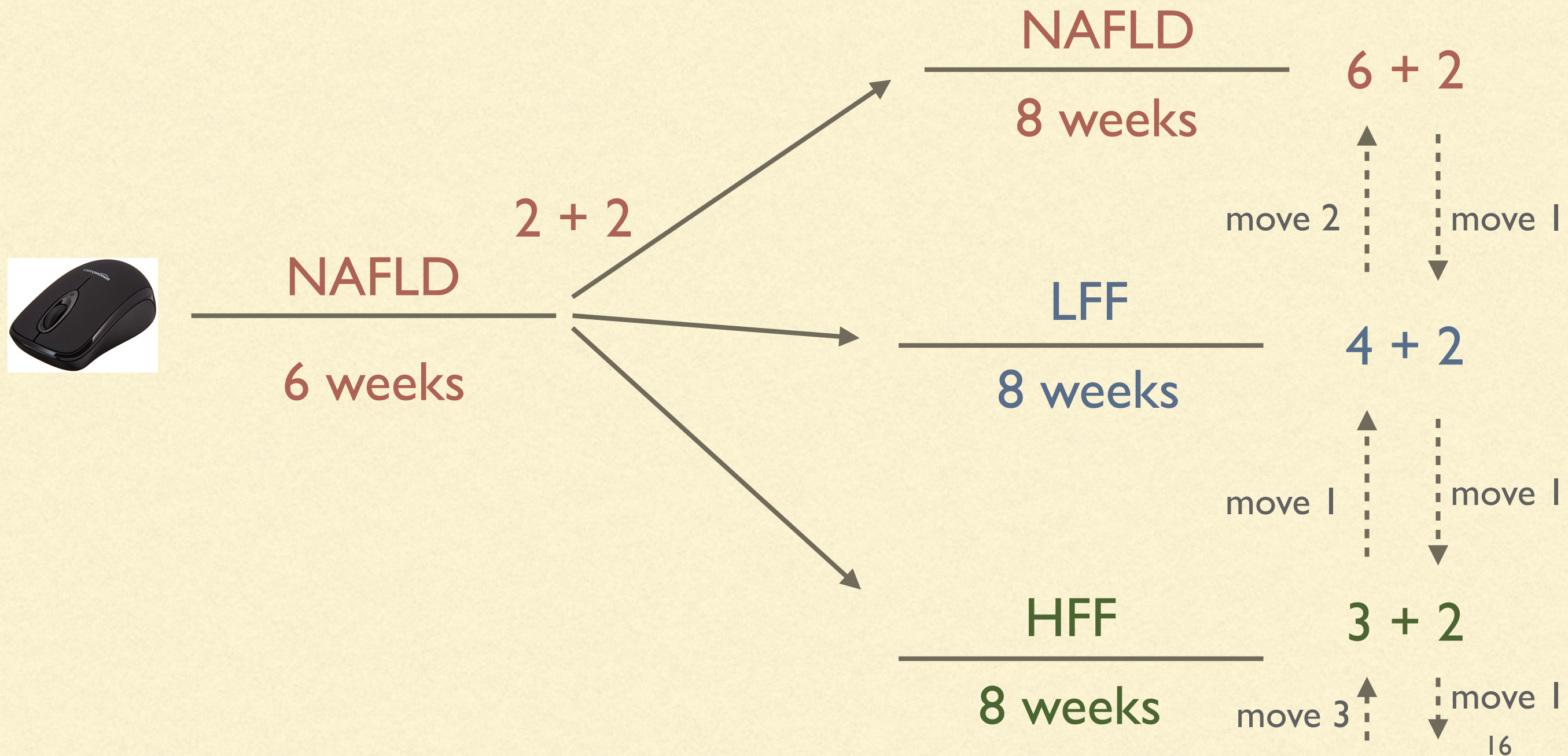


# DEPAOLO LAB STUDY





# AMY'S TWEAK





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# ANALYZING SCRAMBLED DATA

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- First fit logistic regression to all phyla abundances with diet and cohort as fixed effects... nothing
- Repeat at family level... nothing
- Then fit logistic regression with only diet... nothing
- Then fit linear model with only diet...

*“We found a significant alteration of Firmicutes in both fecal and ileum samples ( $p = 0.01$  and  $p = 0.04$ ), and a significant alteration of Verrucomicrobia in both fecal and ileum samples ( $p = 0.02$ ,  $p = 0.03$ ). We also found changes in Actinobacteria abundance in fecal matter ( $p = 0.01$ ).”*



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# KEEP IN MIND

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- If you don't know every statistical analysis that was involved in preparing a paper, you cannot make an informed decision about the "significance" of its results
- Fishing: Many tests conducted but only "interesting" results reported
- Withholds information necessary to adjust for multiple comparisons



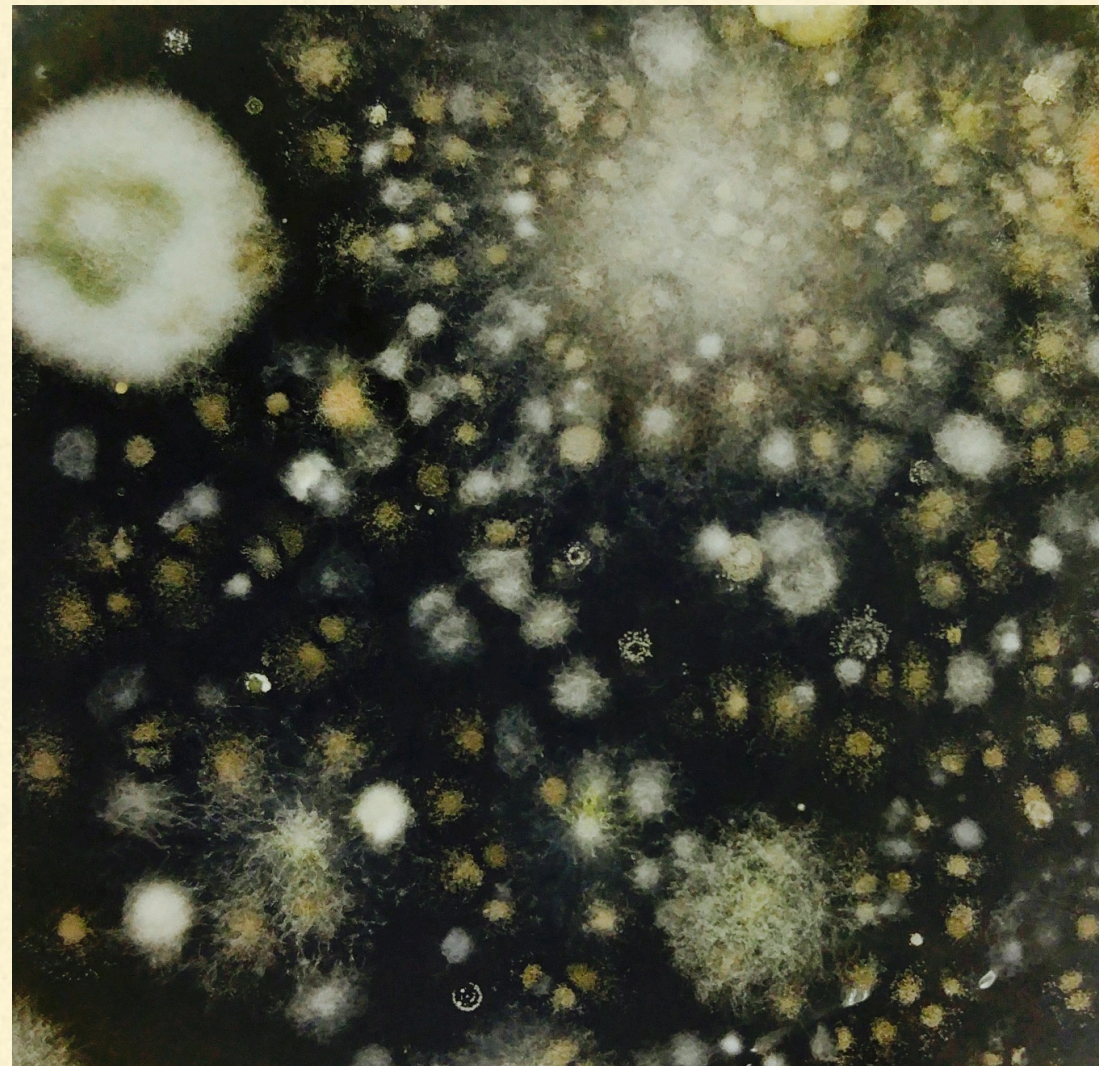
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# RESOURCES

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- Departments of Biostatistics and Statistics @ UW provide consulting services
  - [stat.washington.edu/consulting/](http://stat.washington.edu/consulting/)
- The new Statistical Diversity Lab @ UW
  - <http://faculty.washington.edu/adwillis/>
  - new site coming soon...
- Any STAT/BIOST class in any university...





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