

20.109
Laboratory Fundamentals in
Biological Engineering

Module 1
Nucleic Acid Engineering
Lecture 8

Today

Donut Day

Finish Phylogenetics

Microbiome – other considerations

Basic Epidemiology

Assessing confidence

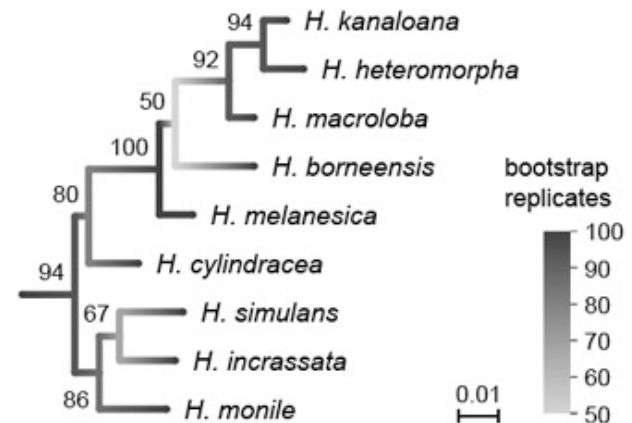
- Trees obtained by phylogenetics are subject to **error** like all other scientific hypotheses
- A tree will be generated regardless of whether there is a phylogenetic signal
- Need to **quantify** how strongly data supports each of the relationships in the tree
- What is the extent to which characters within a matrix **contradict** each other?

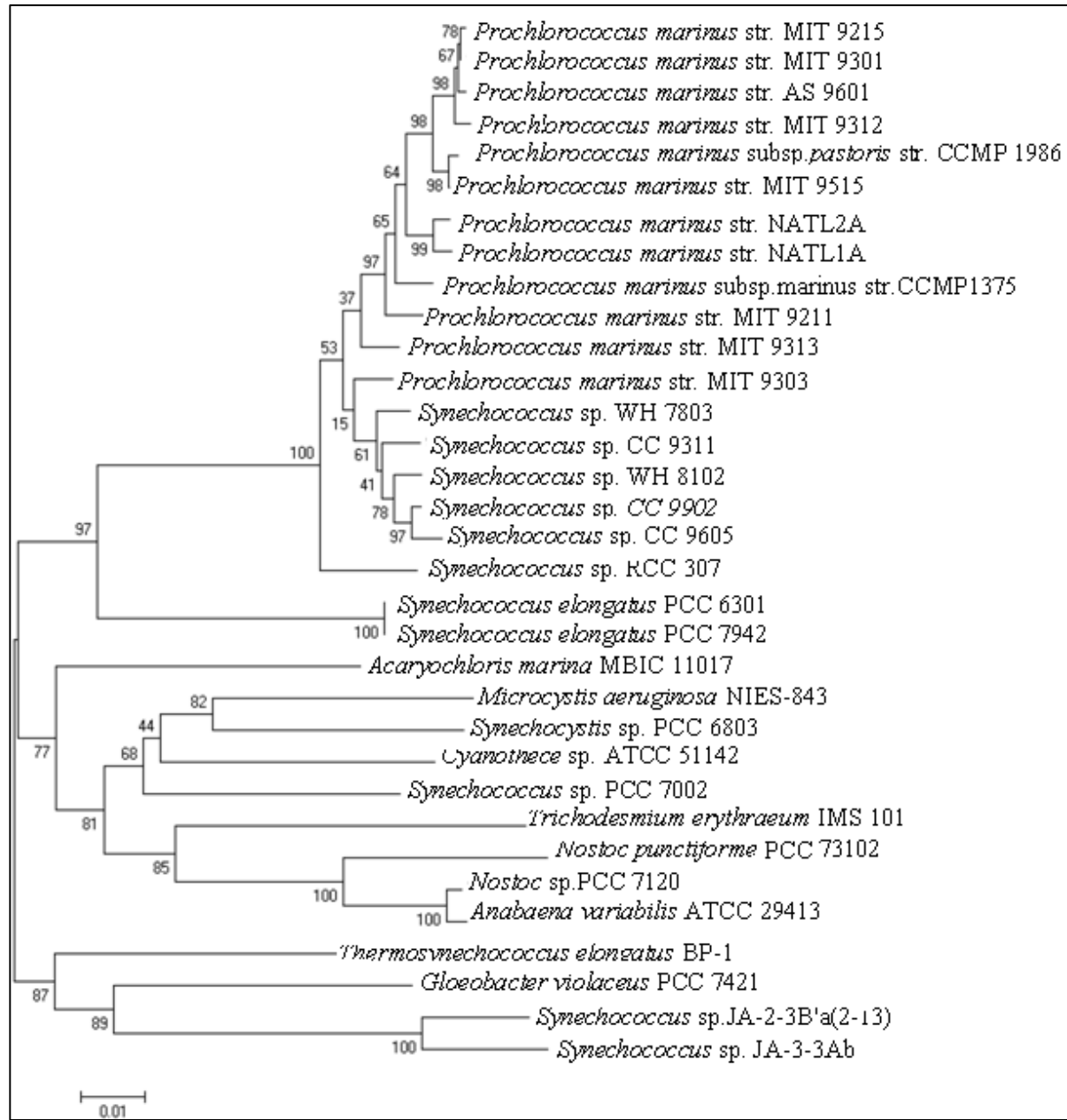
Bootstrapping

- Typically tackled with a statistical test called **bootstrapping**
- Assesses chances of recovering a particular clade again if we randomly re-sample our data
- Data matrix is sampled with replacement to produce **pseudo-replicate** datasets
- Measures which parts of the tree are weakly supported with a low bootstrap %

Bootstrap cut-offs

- Exact interpretation of bootstrap % is elusive
- Higher is better but what is a reasonable cut-off? 70%?
- Warning: bootstrapping predicts whether the same result would occur if more data were collected not whether the result is correct





Galaxy <http://ops...20109.apc...> FastUniFrac FastUniFrac FastUniFrac Intestinal microbiota a... Intestinal Microbiota a... Kiebsidele...
 unifrac.colorado.edu/root human microbiome
 Most Visited Save to Mendeley Add to ReadCube Pandora Radio Getting Started Latest Headlines Google MIT - Massachus... Institute of Arctic... Dictionary and Th... University of Alas... Events organiz... Diigoet
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Galaxy Analyze Data Workflow Shared Data Visualization Help User Using 9.6 MB

Tools
 search tools
 Get Data
 FastUniFrac
 Workflows
 All workflows

POWERED BY PYCOAGENT

FastUniFrac

powered by QIIME

ATTENTION: We have recently discovered some issues with the error report system. If you need to report an error/bug, please use the built-in system and forward the bug report that you will receive to MicrobiomeHelp@colorado.edu. Sorry for any inconvenience.

Fast UniFrac is a new version of **UniFrac** that is specifically designed to handle very large datasets. Like **UniFrac**, **Fast UniFrac** provides a suite of tools for the comparison of microbial communities using phylogenetic information. It takes as input a single phylogenetic tree that contains sequences derived from at least three different environmental samples, a file mapping ids used in the tree to a set of unique sample ids (same format as prior version 'environment file', and an (optional) category mapping file describing additional relationships between samples and subcategories for visualizations. For example, in a given set of gut samples, you might define subcategories for different diets, different physical locations/dates, different species, and/or different treatments like antibiotics or high fat. For sample data click [here](#). For citation, click [here](#).

Both the UniFrac distance metric and the P test can be used to make comparisons. Both of these techniques bypass the need to choose operational taxonomic units (OTUs) based on sequence divergence prior to analysis.

Fast UniFrac allows you to:

- Determine if the samples in the input phylogenetic tree have significantly different microbial communities.
- Cluster samples to determine whether there are environmental factors (such as temperature, pH, or salinity) that group communities together.
- Determine whether system under study was sampled sufficiently to support cluster nodes.
- Easily visualize the differences between samples graphically, with support for three dimensional exploration of datasets and with multiple subcategory coloring.

Please enter your email and password to continue. After you register you will be able to analyze up to **100000** unique sequences, up to **200** samples, and perform significance test based on up to **1000** tree permutations.

If you wish to analyze much larger datasets than the defaults, please contact us and we will be happy to try to accommodate you.

Fast UniFrac tutorial

Introduction

This tutorial takes you through the steps of analyzing data in the Fast UniFrac web application. The purpose of this tutorial is to show you how to use the interface to find the important variables for describing phylogenetic variation among your samples: in this case, to test what types of physical or chemical factors are most important for structuring bacterial diversity. The dataset used in this tutorial includes 50 of the 464 samples analyzed in Ley, RE, Lozupone, CA, Hamady, M, Knight, R and JI Gordon. (2008). Worlds within worlds: evolution of the vertebrate gut microbiota. Nat. Rev. Microbiol. 6(10): 776-88 ([PubMed](#)). It includes sequences from 16S ribosomal RNA surveys of diverse free-living bacterial assemblages and the guts of diverse mammals and termites. At the end of this tutorial, you should be fully equipped to test hypotheses about your own sequences.

Also included in this tutorial are other example files you may use to explore some of the other features of Fast UniFrac.

Example data files

To use Fast UniFrac, you need three files: a tree file, a sample id mapping file, and a category mapping file. The tree file contains a phylogenetic tree, in Newick format. The sample id mapping file contains a table showing how many times each taxon (from the tree) occurred in each of your samples. The category mapping file contains additional metadata about the samples, and is a table relating each sample to parameters you have measured such as temperature, pH, etc. In general, people usually prepare the two mapping files using Excel, although it is important to save them as plain text format and not as Excel documents.

You can either generate your own tree file, or use one of the reference trees. The PhyloChip reference tree matches the probes on the PhyloChip and is useful for analyzing PhyloChip data; the Greengenes reference tree is from the Greengenes core set and is a phylogenetically diverse and representative set of bacteria. These trees are built using 16S rRNA, although you can use trees built from any molecule, not just the 16S, or even trees constructed from morphological or other data.

History
 Unnamed history
 9.6 MB
 24: UniFrac Significance on data 8, data 20, and data 10
 23: UniFrac Significance on data 8, data 20, and data 10
 22: UniFrac Significance on data 8, data 20, and data 10
 3.3 KB
 format: html, database: ?
 HTML file
 21: UniFrac Significance on data 8, data 20, and data 10
 4 lines
 format: txt, database: ?
 #unweighted unifrac significance test

sample 1	sample 2	p val
274	298	0.312 0.936
274	312	0.06 0.18
298	312	0.084 0.012

 20: ID file.txt
 19: Sample Distance Matrix on data 8, data 9, and data 10
 2.0 KB
 format: html, database: ?
 HTML file
 18: Sample Distance Matrix on data 8, data 9, and data 10
 4 lines
 format: txt, database: ?

Galaxy File Edit Capture Window Help http://o. 109.ape Fastunifrac Fastunifrac Fastunifrac Intestinal micro... Intestinal Micro... Metabipha oxyto... Tue 10:21 AM Jonathan Runstadler

unifrac.colorado.edu/root evenness alpha diversity

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Galaxy Analyze Data Workflow Shared Data Visualization Help User Using 9.6 MB

Tools

FastUniFrac

- Cluster samples** Uses the UniFrac metric to cluster the samples based on phylogenetic lineages they contain.
- Jackknife Sample Clusters** Performs statistical resampling and will allow you to see how confident you should be in the sample clustering results.
- PCoA** Uses the UniFrac metric to perform principal coordinates analysis on your samples, allowing you to see whether different types of samples are separated in different dimensions.
- P Test Significance** Tells you which pairs of samples are significantly different using the P Test.
- Sample counts** Tells you how many sequences are in each sample.
- Sample Distance Matrix** Shows you the UniFrac distances between each pair of samples and is used as input for sample clustering and PCoA.
- UniFrac Significance** Tells you which pairs of samples are significantly different using the UniFrac significance test.

UniFrac Significance (version 1.0)

Select reference tree:
22: Unifrac Significa..and data 10

Select sample ID mapping file:
22: Unifrac Significa..and data 10

Select category mapping file:
22: Unifrac Significa..and data 10

Number of permutations:
50

Use abundance weights:

Type of test:
All samples together

Execute

Calculating the UniFrac metric: The majority of options in the FastUniFrac interface make comparisons based on the UniFrac metric. The UniFrac metric measures the difference between two samples in terms of the branch length that is unique to one sample or the other. In the tree on the right (panel C below), the division between the two samples (labeled red and blue) occurs very early in the tree, so that all of the branch length is unique to one sample or the other. This results in the maximum UniFrac distance possible, 1.0. In the tree on the left, every sequence in the first samples has a very similar counterpart in the other samples, and all of the branch length in the tree comes from nodes that have descendants in both samples. The results in the minimum UniFrac distance of 0.0. In the middle example, there is about as much branch length unique to each sample (red or blue) as is shared between samples (purple), so the UniFrac distance would be about 0.5.

A. Identical sequence sets: all seqs in red + blue set. 100% branch length shared (purple).	B. Related sequence sets: seqs in red have relatives in blue. ~50% branch length shared.	C. Unrelated sequence sets: seqs in red have no close relatives in blue. 0% branch length shared.
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History

Unnamed history
9.6 MB

X24: UniFrac Significance on data 8, data 20, and data 10

X23: UniFrac Significance on data 8, data 20, and data 10

22: UniFrac Significance on data 8, data 20, and data 10
3.3 KB
format: html, database: ?

HTML file

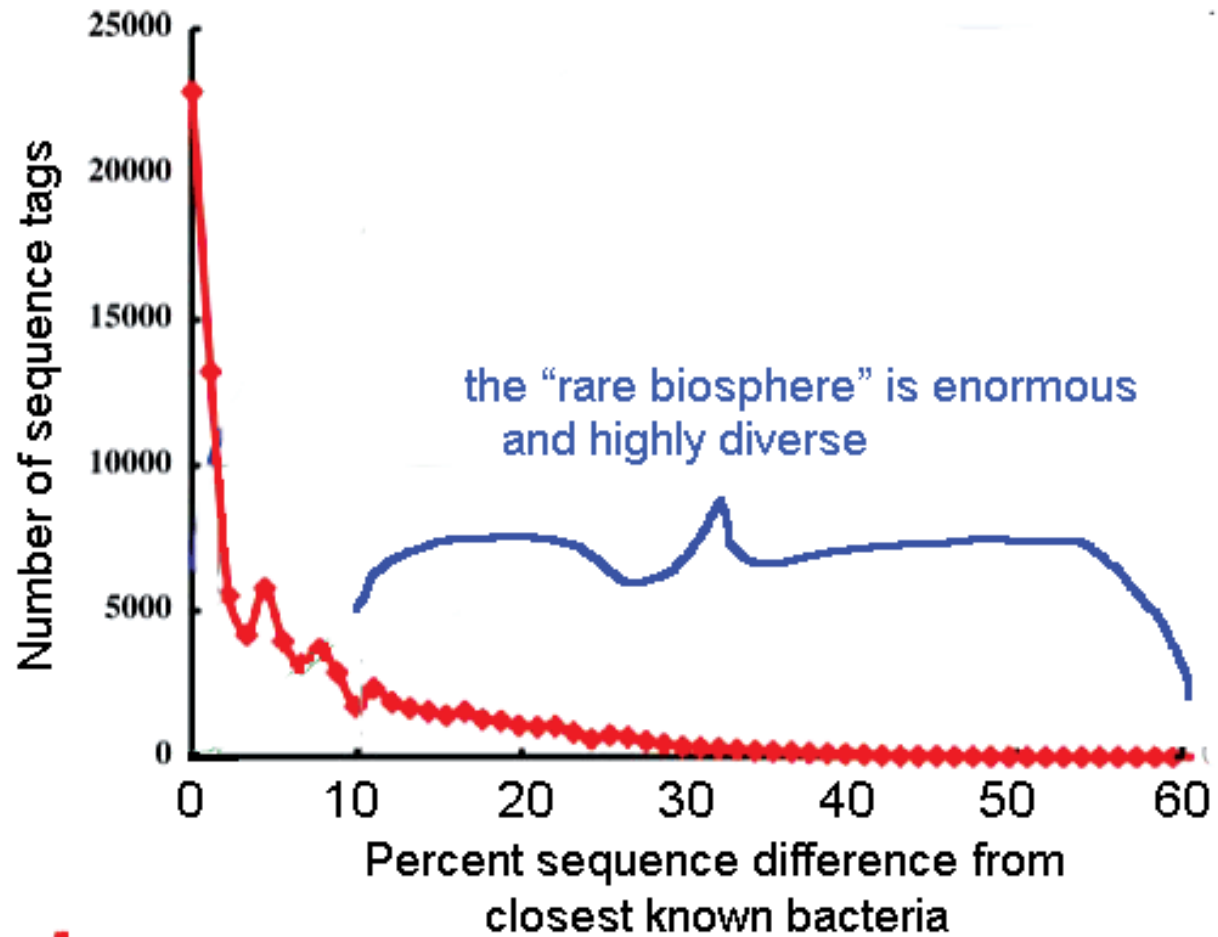
21: UniFrac Significance on data 8, data 20, and data 10
4 lines
format: txt, database: ?

```
#unweighted unifrac significance test
sample 1    sample 2    p value
274    290    0.312    0.936
274    312    0.06    0.18
290    312    0.004    0.012
```

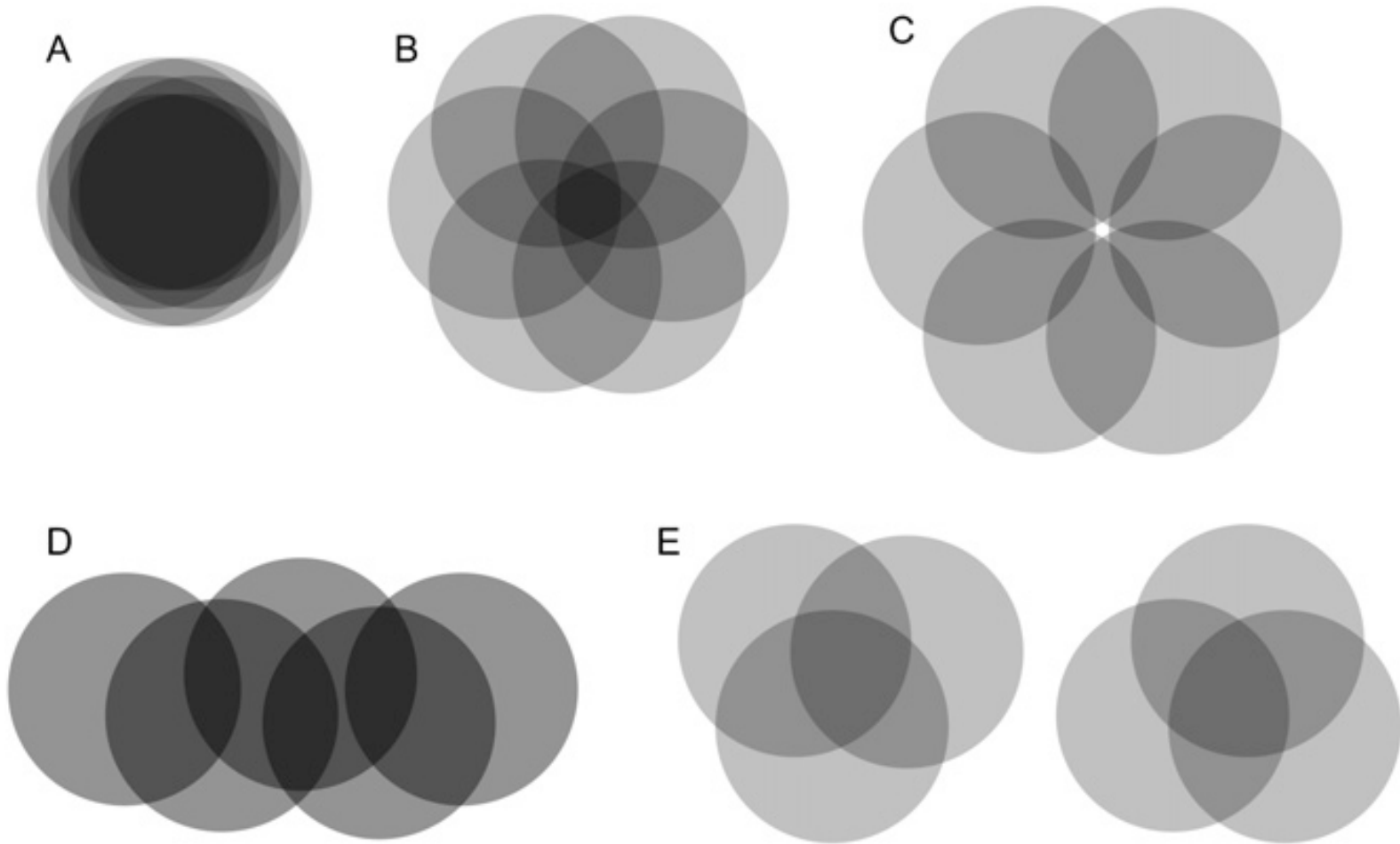

Back to the core questions

- Structure of the microbiome?
- Function of the microbiome?
- How can it be changed?

Rare biosphere

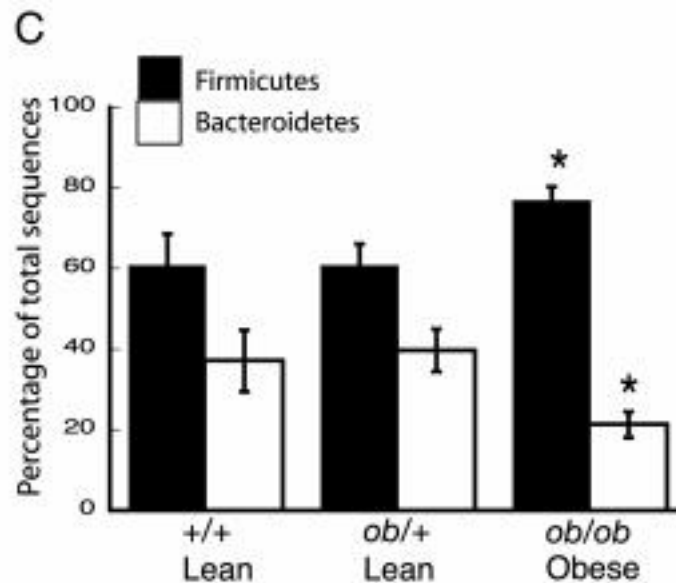


Models of a core microbiome

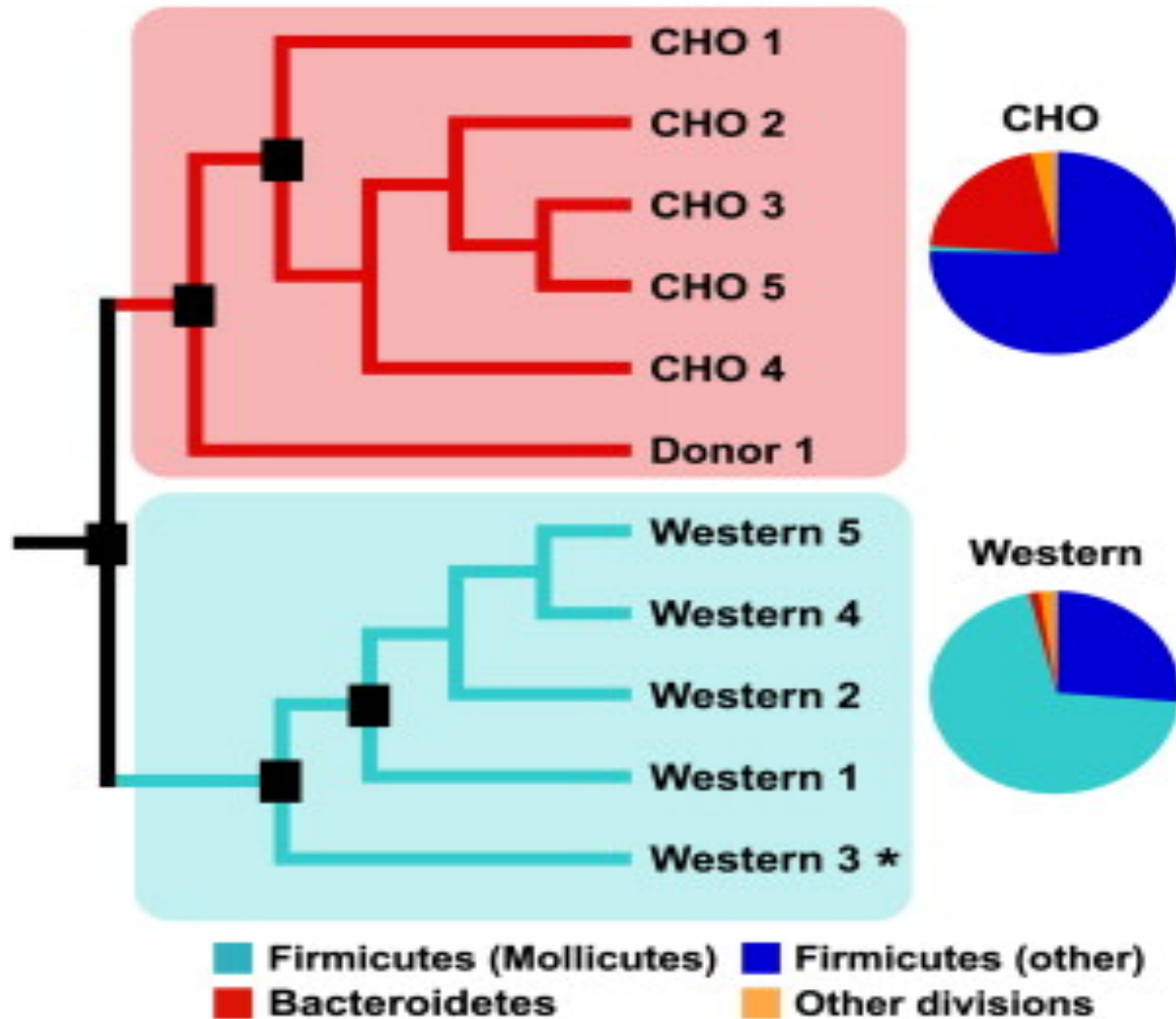


Does diet affect microbial composition?

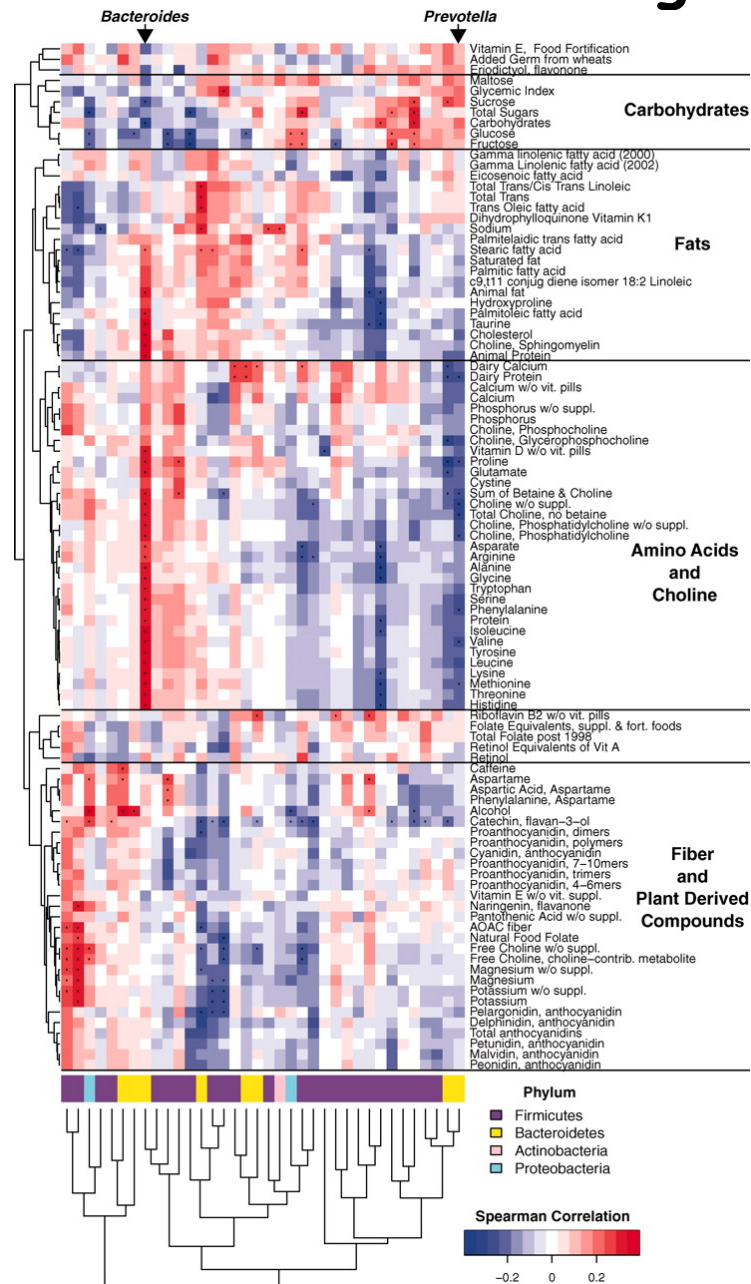
- Genetically Obese mice harbor a significantly different community than lean conventional mice



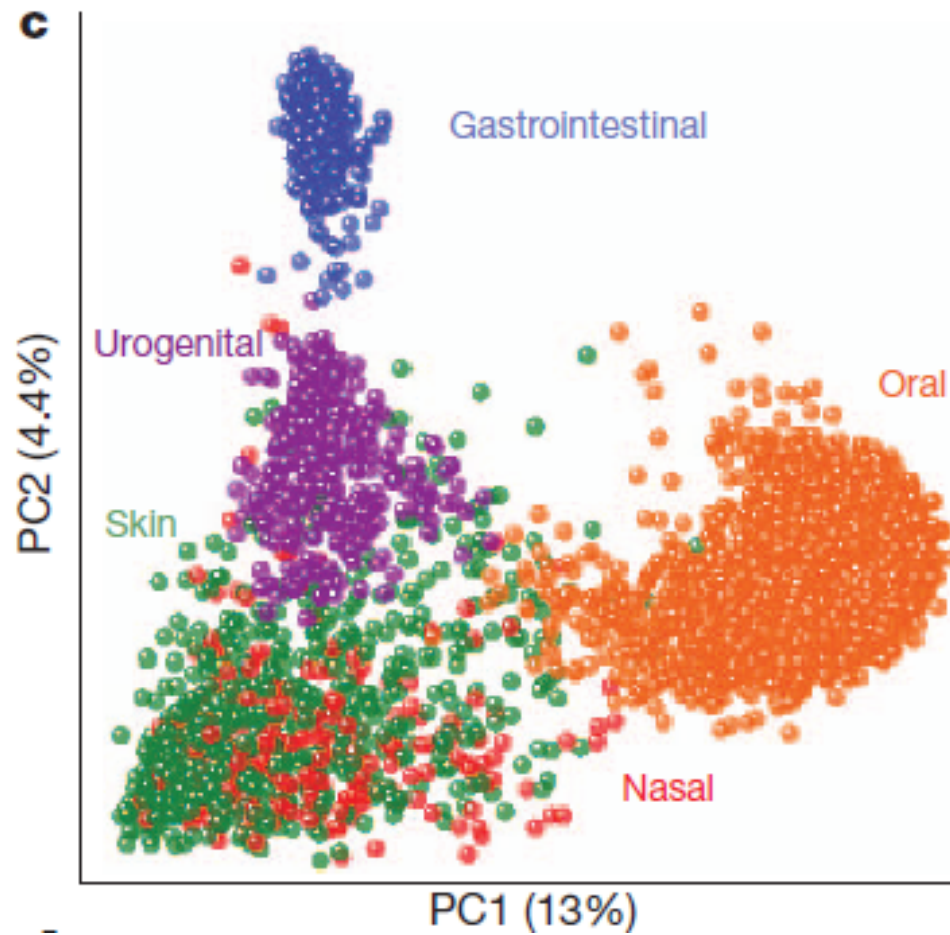
Diet affects microbial composition



Correlation of diet and gut microbial taxa



The Human Microbiome

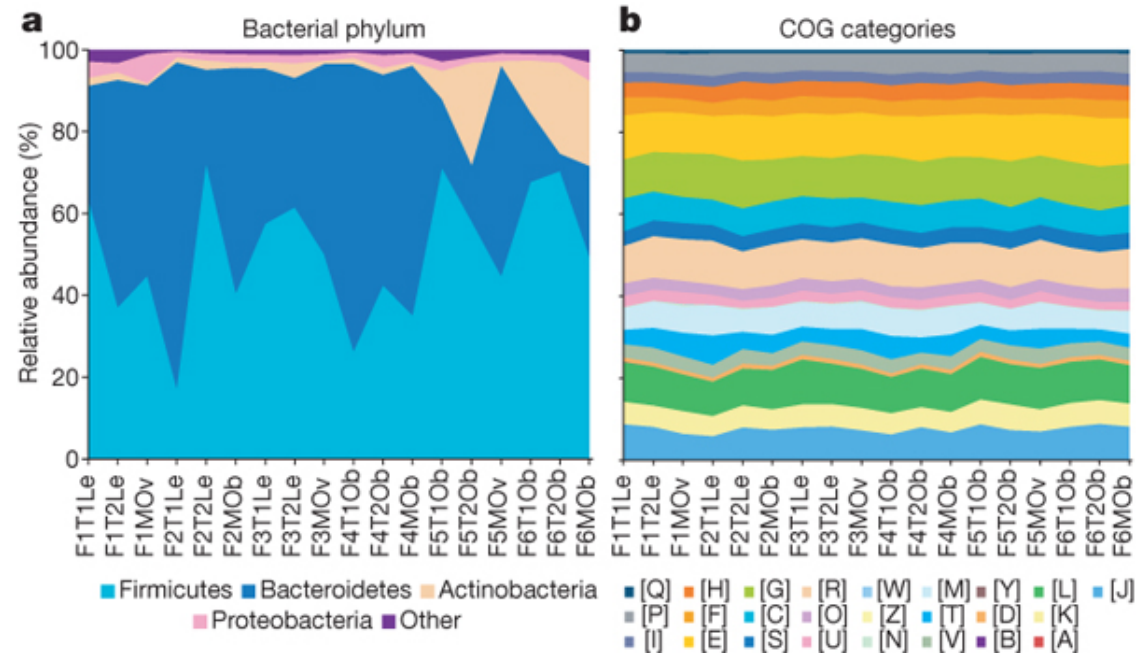




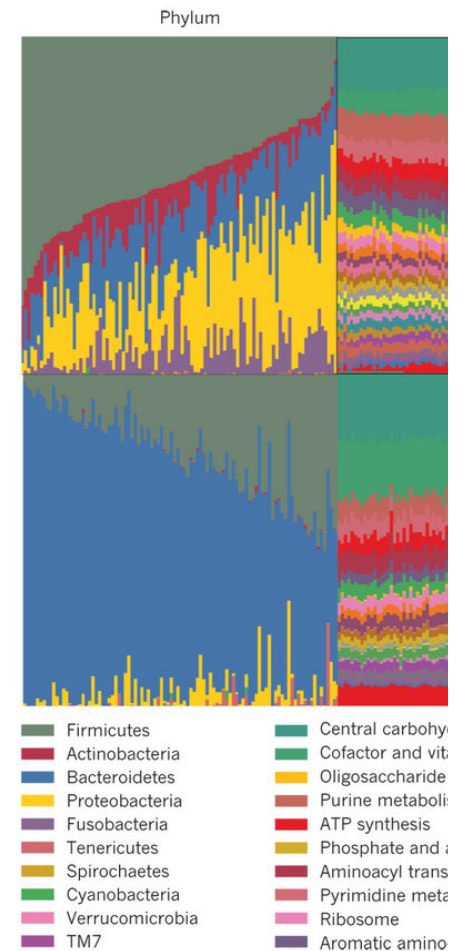
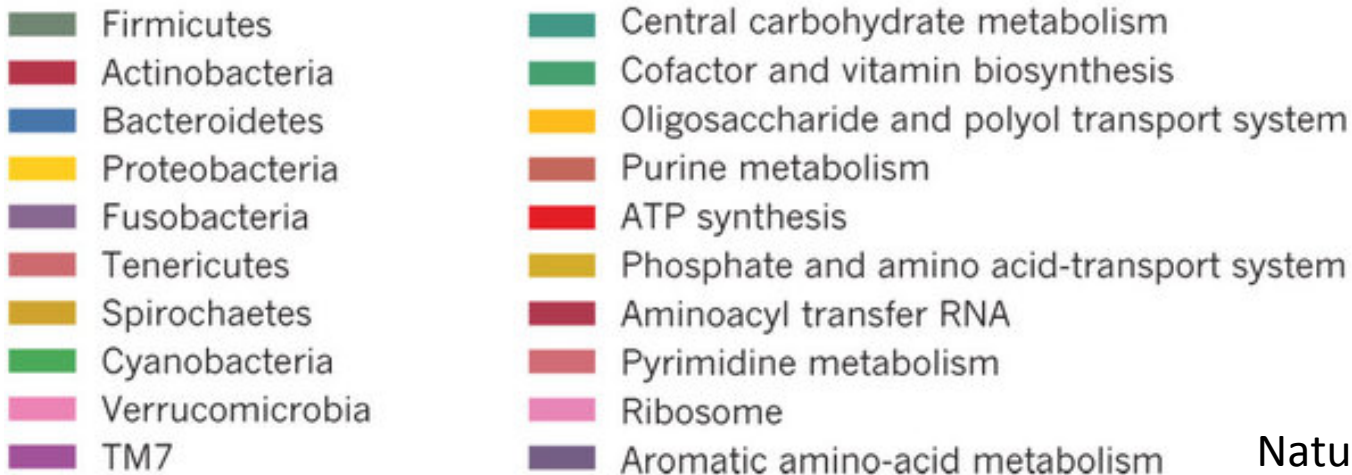
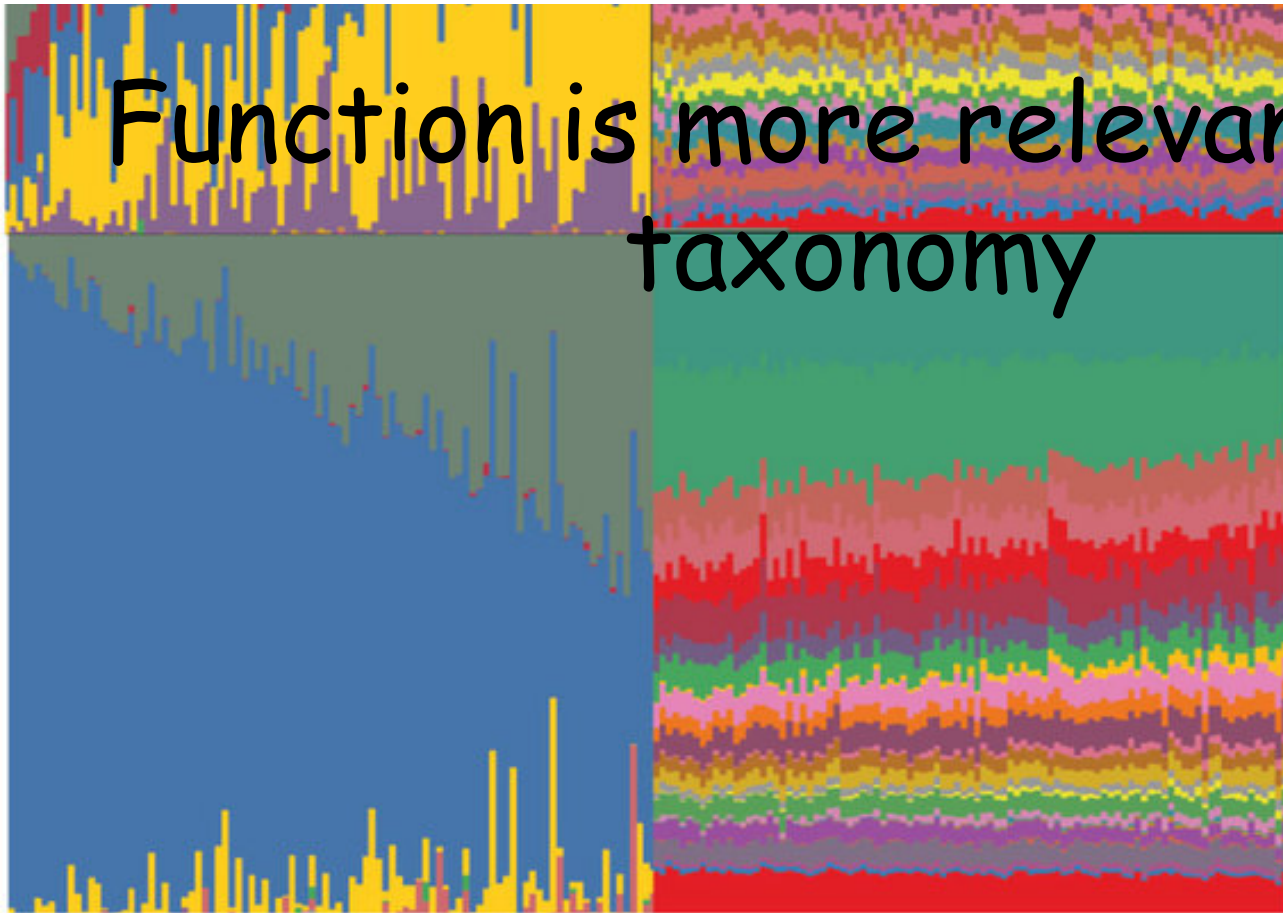
If taxonomy is not conserved,
what does that mean for
function?

- Functional core?
- Interchangeable parts?

Comparison of taxonomic and functional variations



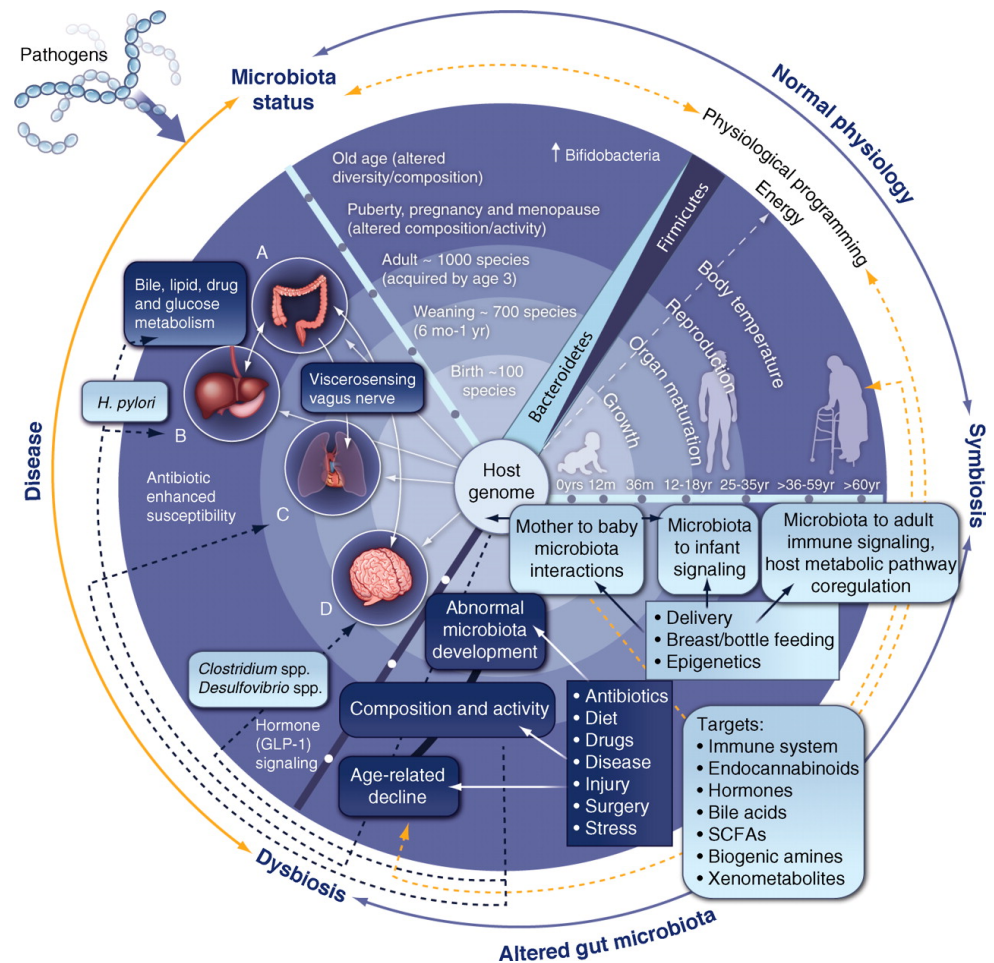
Function is more relevant than taxonomy



Nature 486 (2012)

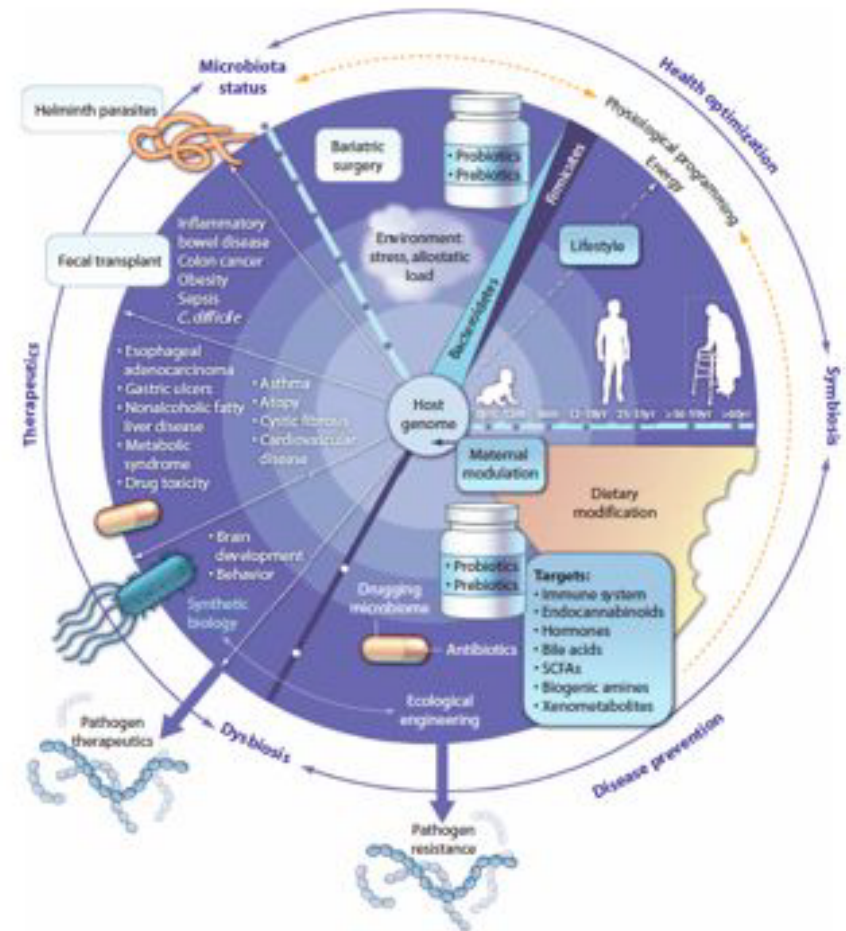
Host-gut microbiota metabolic interactions

- After birth ~ 100 microbial spp.
- Env., nutrition, influence later development
- Microbiota influences normal development, physiology, immune system, etc, at all life stages
- Dysbiosis involved in a # of diseases:
 - IBD, IBS, colon cancer
 - ulcers, fatty liver, obesity
 - asthma, hypertension
 - Mood and behavior (GLP-1)



Host-gut microbiota metabolic interactions

Is engineered homeostasis achievable?
- *C. difficile* transplants



Do you trust the microbiome?

5 questions:

- 1) Can experiments detect differences that matter?
- 2) Do studies show causation or just correlation?
- 3) What is the mechanism?
- 4) How much do experiments reflect reality?
- 5) Could anything else explain the results?

Evaluation of a diagnostic test

- Sensitivity
- Specificity

Calculating sensitivity and specificity

		True disease	
		+	-
Test	+	a 10	b 2
	-	c 5	d 83

$$\text{Sensitivity} = \frac{a}{a+c}$$
$$\frac{10}{15}$$

$$\frac{d}{b+d} = \text{Specificity}$$
$$\frac{83}{85}$$

Test Accuracy

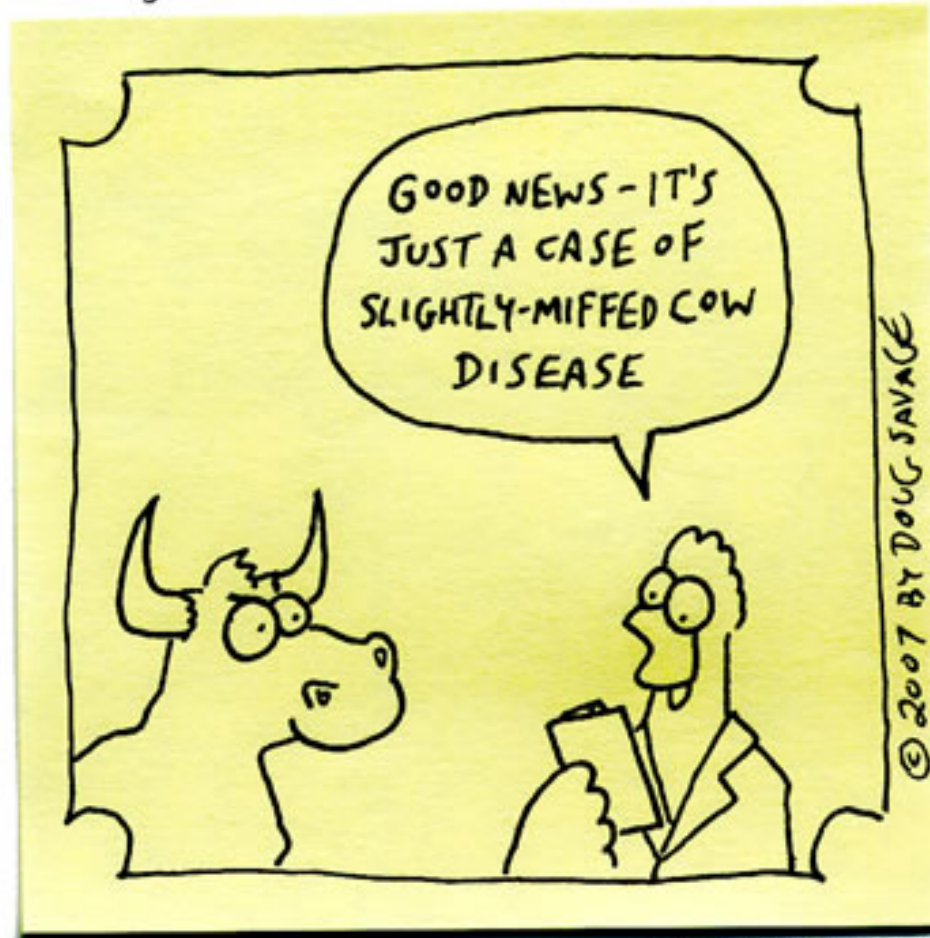
		True disease	
		+	-
Test	+	a 20	b 2
	-	c 10	d 68

$$\text{Accuracy} = \frac{88}{100} = 88\%$$

$$\text{Prevalence} = \frac{30}{100} = 30\%$$


Savage Chickens


by Doug Savage



Sensitivity and Specificity and Predictive values

		True disease	
		+	-
Test	+	a 20	b 2
	-	c 10	d 68

Positive Predictive Value = $\frac{a}{a+b} = \frac{20}{22} = \frac{10}{12}$ 

Negative Predictive Value = $\frac{d}{c+d} = \frac{68}{78} = \frac{83}{88}$ 

Sensitivity = 67%

98% = Specificity

Likelihood ratios - diagnostic utility of a test

		True disease		
		+	-	
Test	+	a 20	b 2	Likelihood Ratio for a Positive Test = $\frac{a/a+c}{1-(d/b+d)}$ 33.5 = $\frac{20/30}{1-(68/70)}$
	-	c 10	d 68	Likelihood Ratio for a Negative Test = $\frac{1-(a/a+c)}{d/b+d}$ 0.33 = $\frac{1-(20/30)}{68/70}$

Sensitivity = 67% 98% = Specificity

Comparing tests?

- When is a test with high sensitivity most useful?
- When is a test with high specificity most useful?