

**Fine-tuning the association  
between pediatric arthritis and  
*Faecalibacterium prausnitzii***

**Matthew Stoll MD**

**Wonderful World of Technology**

**March 23, 2017**

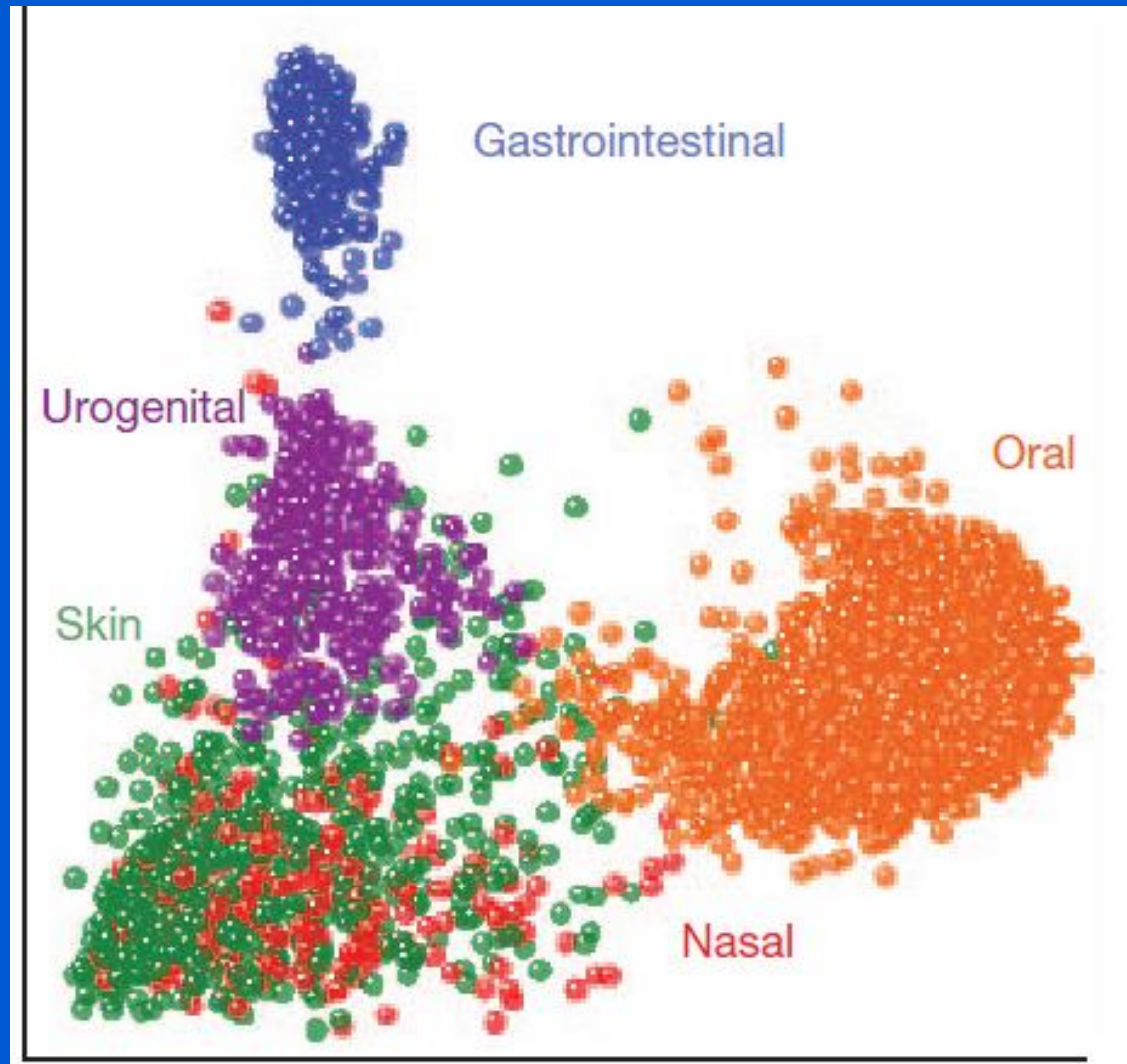
# Microbiota and human diseases

- Inflammatory bowel disease
- Arthritis
  - Juvenile idiopathic arthritis
  - Rheumatoid arthritis
  - Spondyloarthritis
- Type I diabetes
- Obesity / metabolic syndrome

# Outline

- **Where to sample**
- Sequencing
- Data analysis
- Next steps

# Bacteria composition differs by site

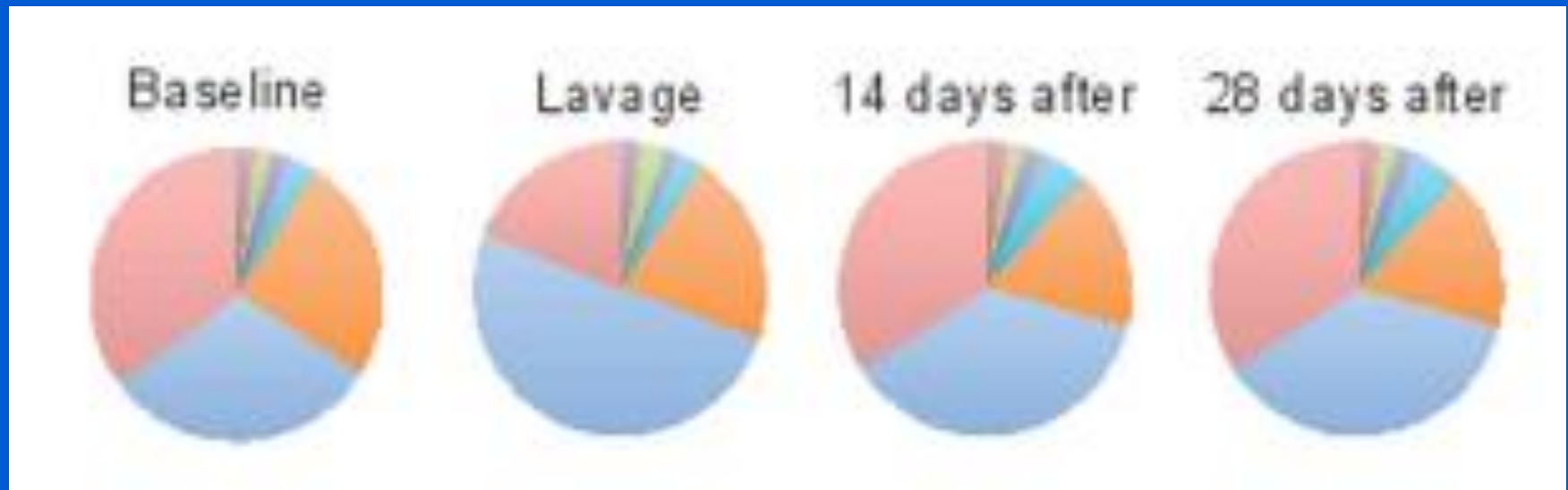
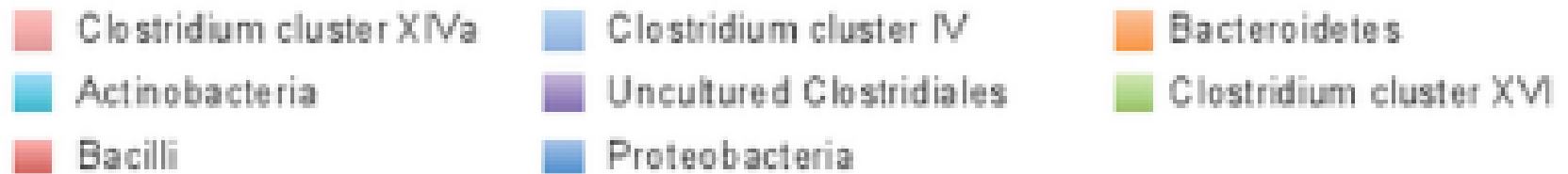


HMP Consortium, *Nature* 2012;486:207

# Heterogeneity within habitats

- Gut
- Skin
- Mouth

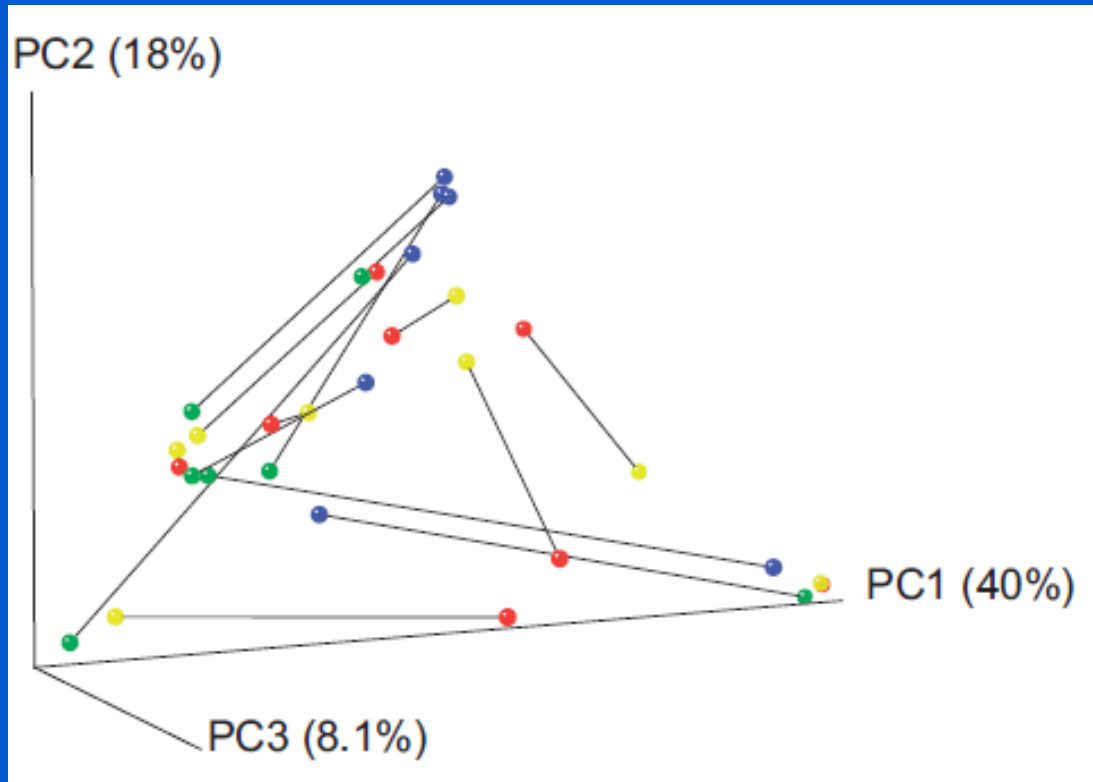
# Bowel prep affects fecal microbiota



Jalanka, *Gut* 2015;64:1562

# And the mucosal microbiota

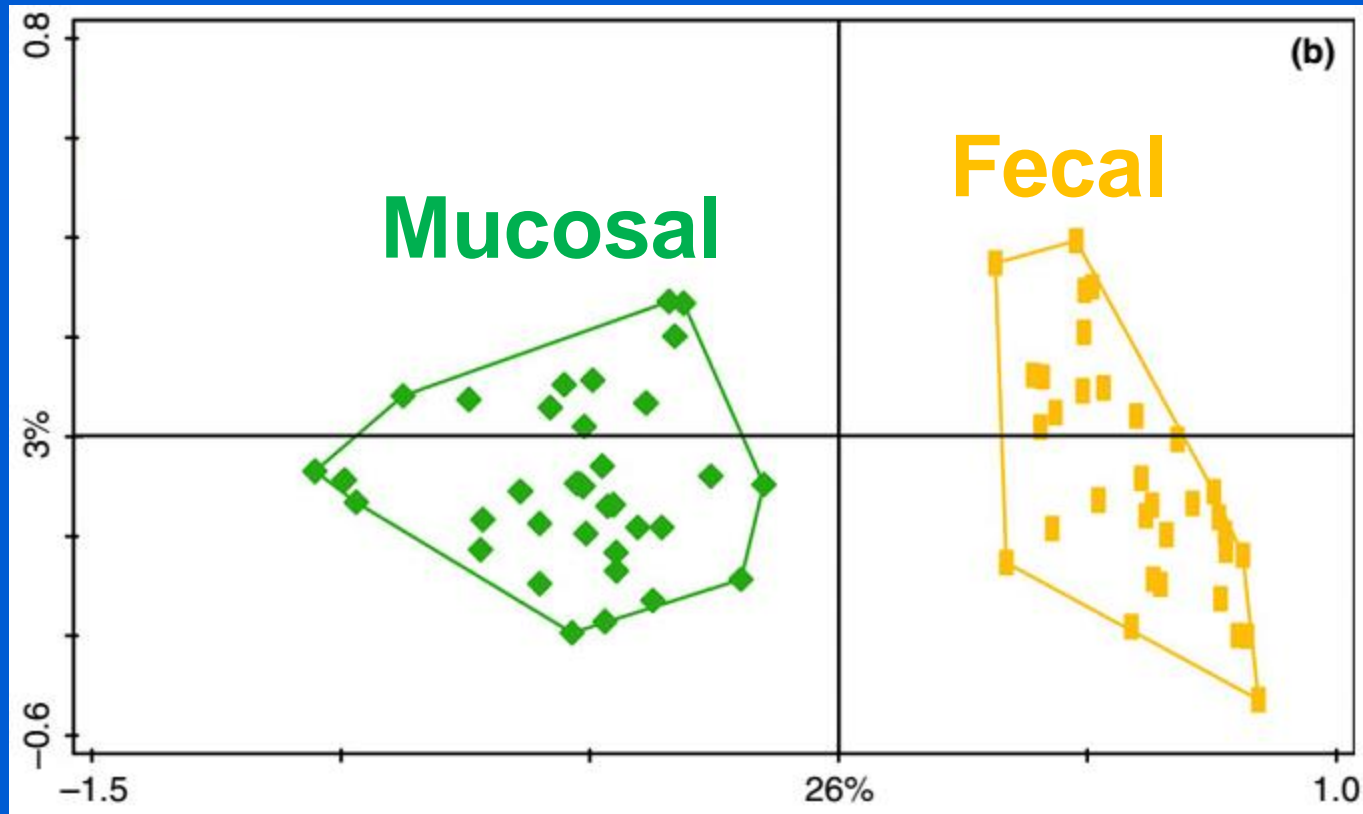
Sigmoid biopsies before and after bowel preparation  
Patients with IBD



Red-yellow: Biopsy samples  
Blue-green: Fecal samples

# Fecal vs mucosal microbiota

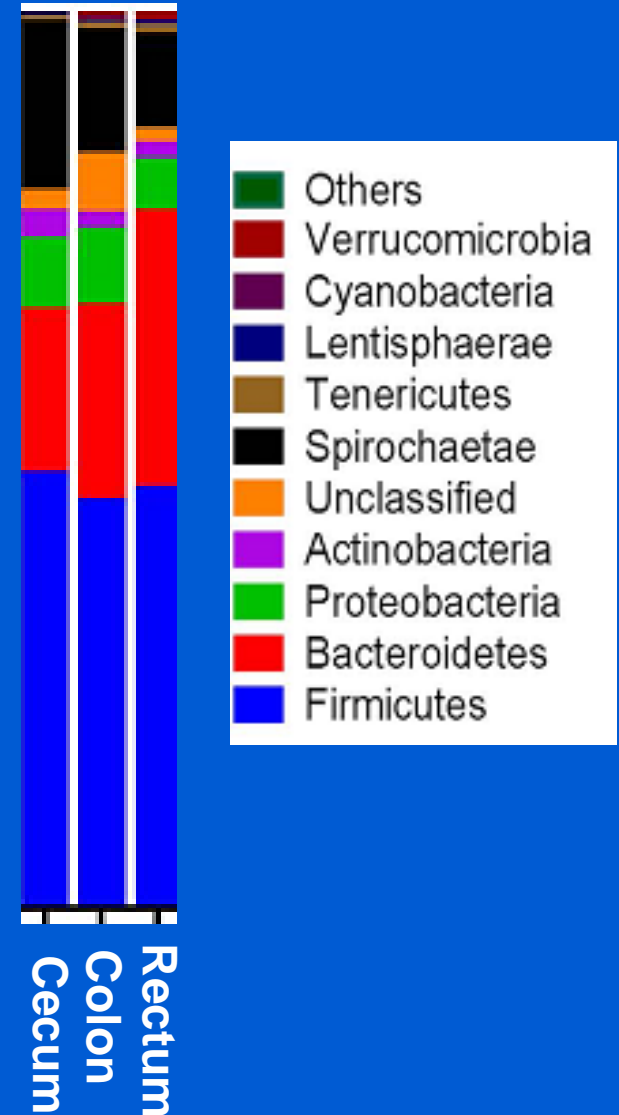
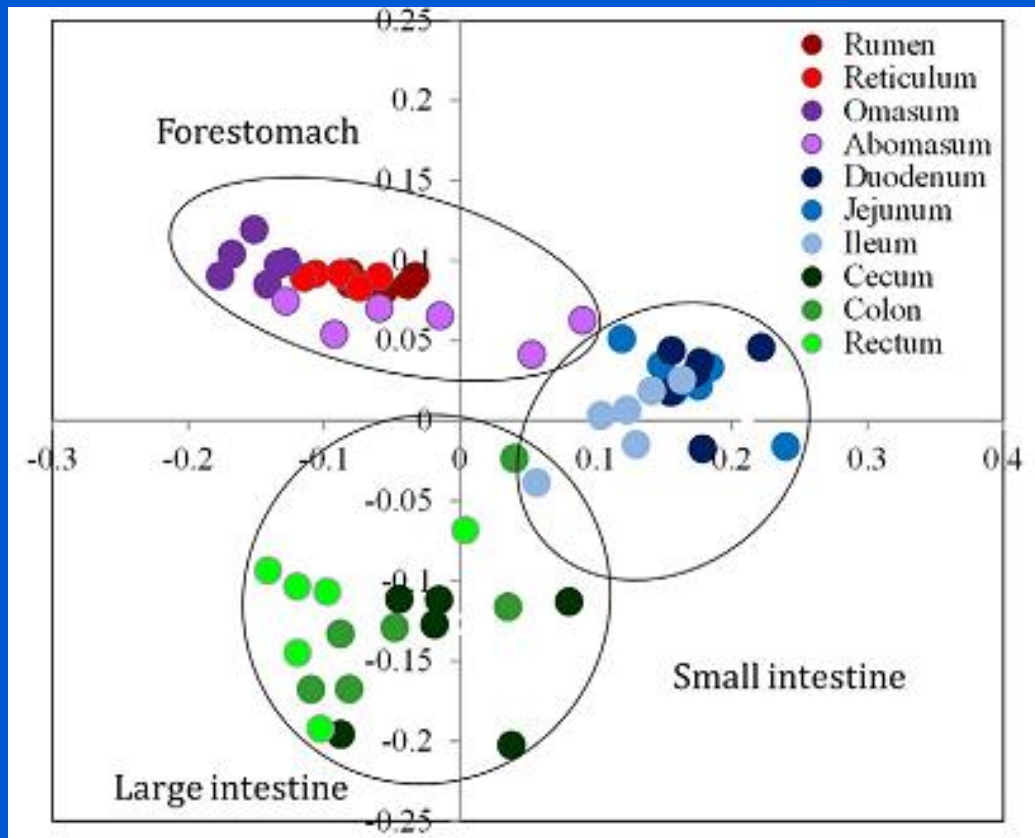
Sigmoid biopsies; no bowel prep



Rangel, *Aliment Pharmacol Ther* 2015;42:1211



# Spatial heterogeneity in intestines



Mao, *Scientific Reports* 2015;5:16116

# Skin has topographically distinct microbiota



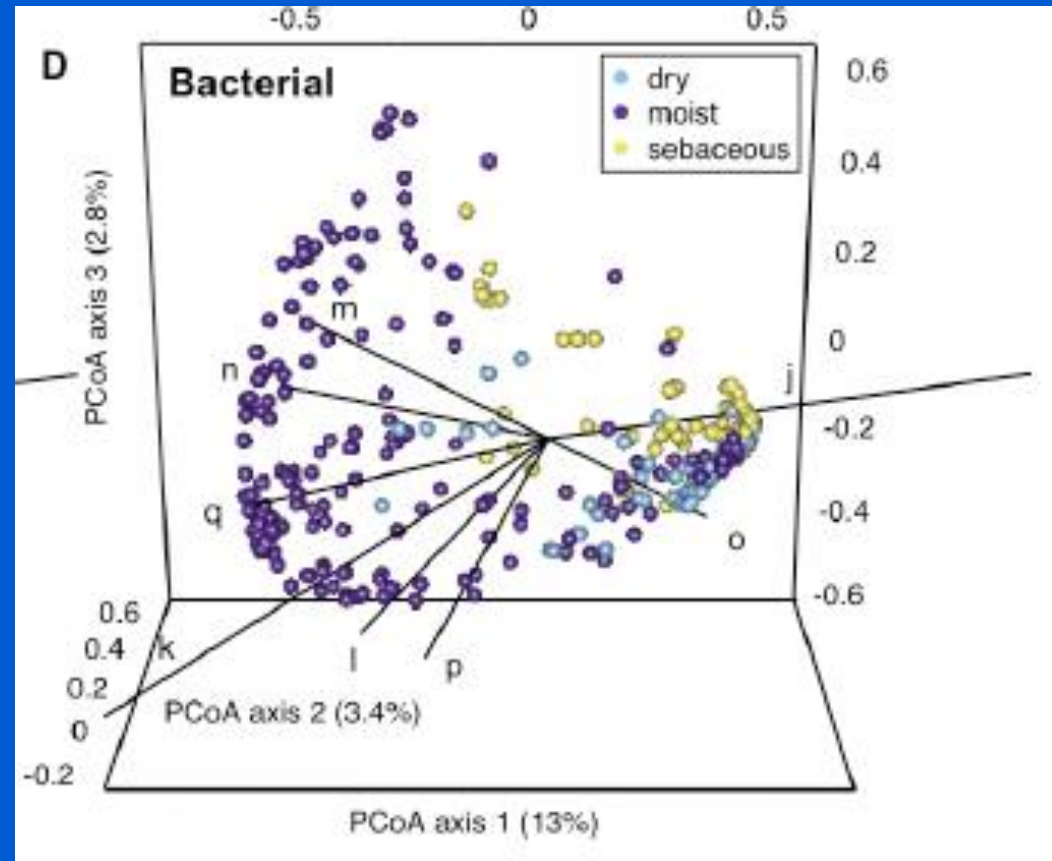
Heel



Elbow



Back



Findley, *Nature* 2013;498:367

# Mouth also has topographically distinct microbiota



Buccal mucosa  
Keratinized gingiva  
Hard palate



Throat  
Tonsils  
Tongue Dorsum  
Saliva



Supragingival plaque  
Subgingival plaque

# DNA preparation

- Various kits used for DNA purification
  - MoBio preferred by HMP
  - We use Zymo for fecal collection
- Key is that you need conditions harsh enough to lyse the microbes

# Outline

- Where to sample
- **Sequencing**
- Data analysis
- Next steps

# Type of sequencing

- Amplification of 16S ribosomal DNA
- Whole genome sequencing

## Whole genome sequencing

**More information**

More information on function

~1000 / sample

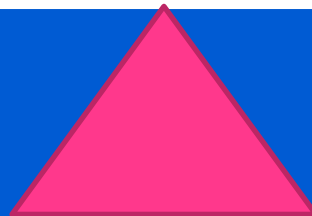
Higher informatics cost

## 16S Sequencing

**Lower cost**

Taxonomy only

\$50 / sample



# Sequencing prep at UAB

## Starting with purified DNA

### 16S (Peter Eipers PhD)

- PCR of V4 16S region
- Special primers
  - Barcode at one end
  - Adaptor at other

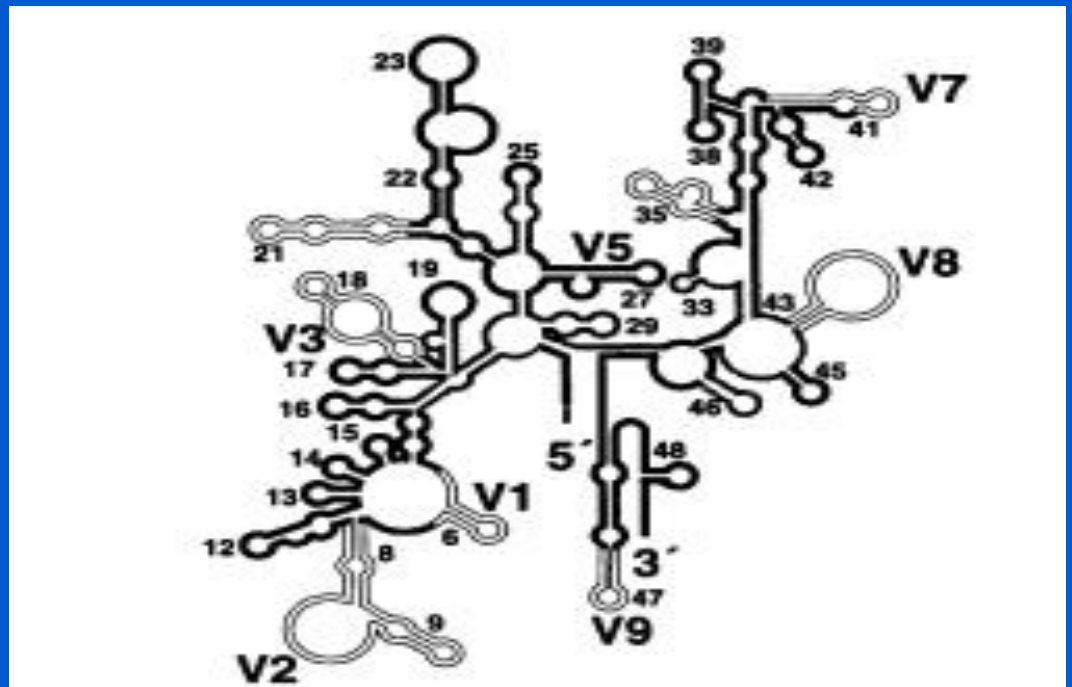
### WGS (Mike Crowley PhD)

- Shear DNA
- Ligate adaptors to each end
  - Includes barcodes
- Short PCR
- Optional size selection

# Key elements of 16S rDNA

- 9 conserved and 9 variable regions
  - Interspersed
- Primers bind to conserved regions
- Amplify variable sequences

Tortoli, *Clin Microbiol Rev* 2003;16:319





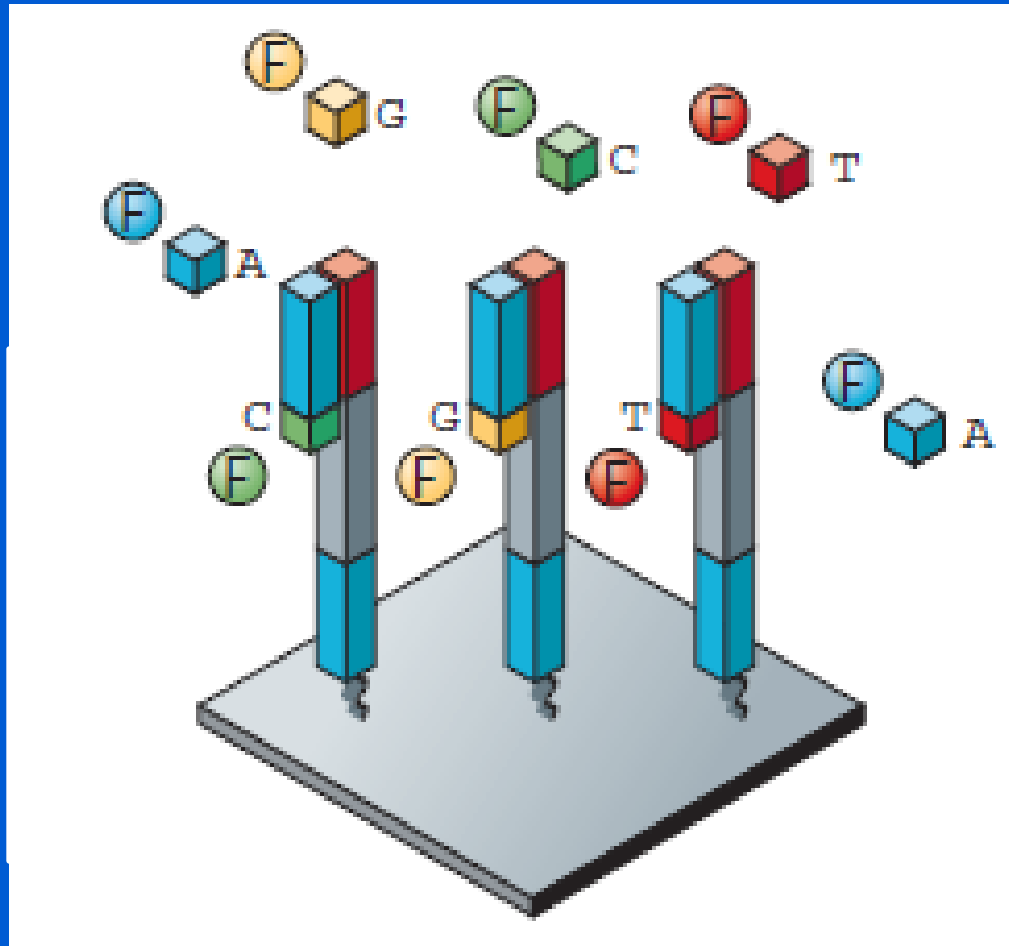
# Foundation of 16S analysis

Position	Consensus composition	Purple bacteria				Gram-positive bacteria		Cyanobacteria	Green sulfur bacteria	Spirochetes		Bacteroides		Planctomyces	Deinococcus	Green non-sulfur bacteria
		$\alpha$	$\beta$	$\gamma$	$\delta$	Lo	High			Spirochetes	Leptospiras	Bacteroides	Flavobacteria <sup>a</sup>			
47	C	● <sup>b</sup>	●	●	●	●	●	●	●	U <sup>c</sup>	U	●	●	G	●	●
48	Y	●	●	●	●	●	●	●	●	●	●	●	●	A	●	●
50	A	●	U	●	●	●	●	●	●	U	U	●	●	U	●	●
52	Y	●	●	●	●	●	●	●	●	A	A	●	●	G	G	●
53	A	●	●	●	●	●	●	●	●	G	●	●	●	G	●	G
353	A	●	●	●	●	●	●	●	●	●	●	●	●	U	●	●
570	G	●	●	●	●	●	●	●	●	●	●	U	U	U	●	●
698	G	U	●	●	Y	●	●	●	●	●	●	●	●	●	●	●
812	G	●	C	● <sup>c<sup>d</sup></sup>	●	●	●	●	●	●	C	●	● <sup>c</sup>	●	C	●
906	G	Ag	Ag	Ag	Ag	● <sup>a</sup>	A	●	●	●	●	●	●	●	●	A
933	G	●	●	●	●	●	●	●	● <sup>a</sup>	●	●	●	●	A	●	●
955	U	●	Au	●	●	●	AC	●	●	●	●	●	●	C	●	●
976	G	●	A	●	●	●	●	●	●	● <sup>a</sup>	A	●	●	●	●	●
983:1	— <sup>e</sup>	—	—	—	—	—	—	—	—	—	—	—	—	U	—	—
995	C	●	●	●	●	●	●	●	(A) <sup>f</sup>	●	●	A	A	●	●	●

Carl Woese, *Microbiol Rev* 1987;51:221

# Cyclic reversible termination

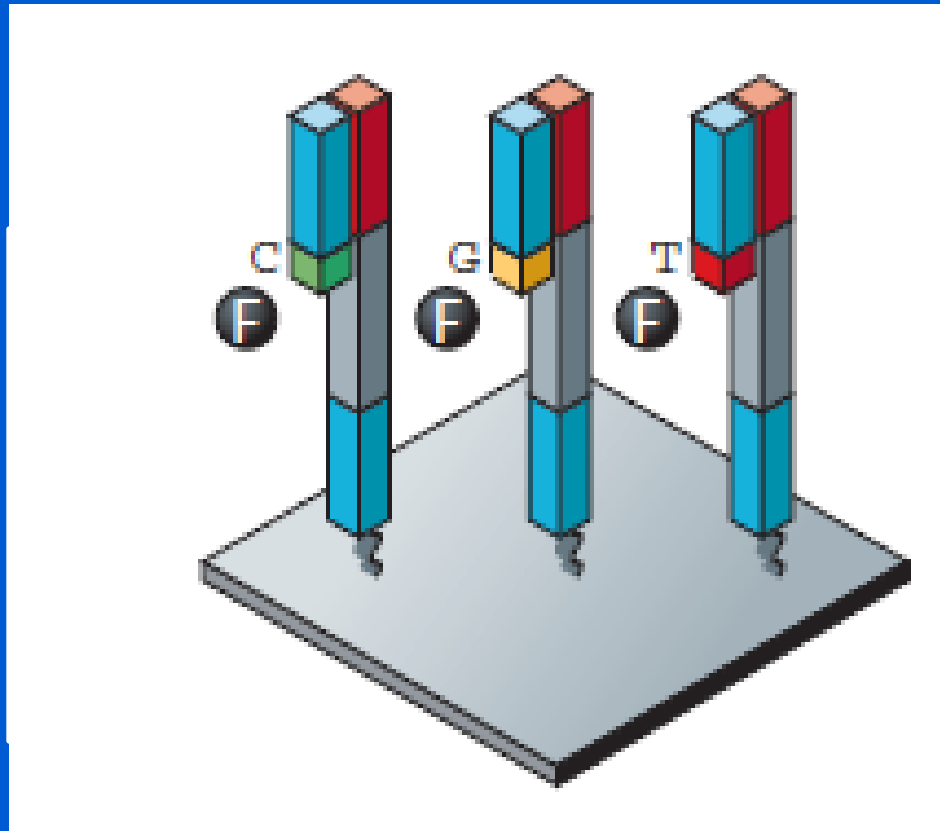
Metzker, *Nature Rev Genet* 2010;11:31



Incorporate all four nucleotides, each with different dye

# Cyclic reversible termination

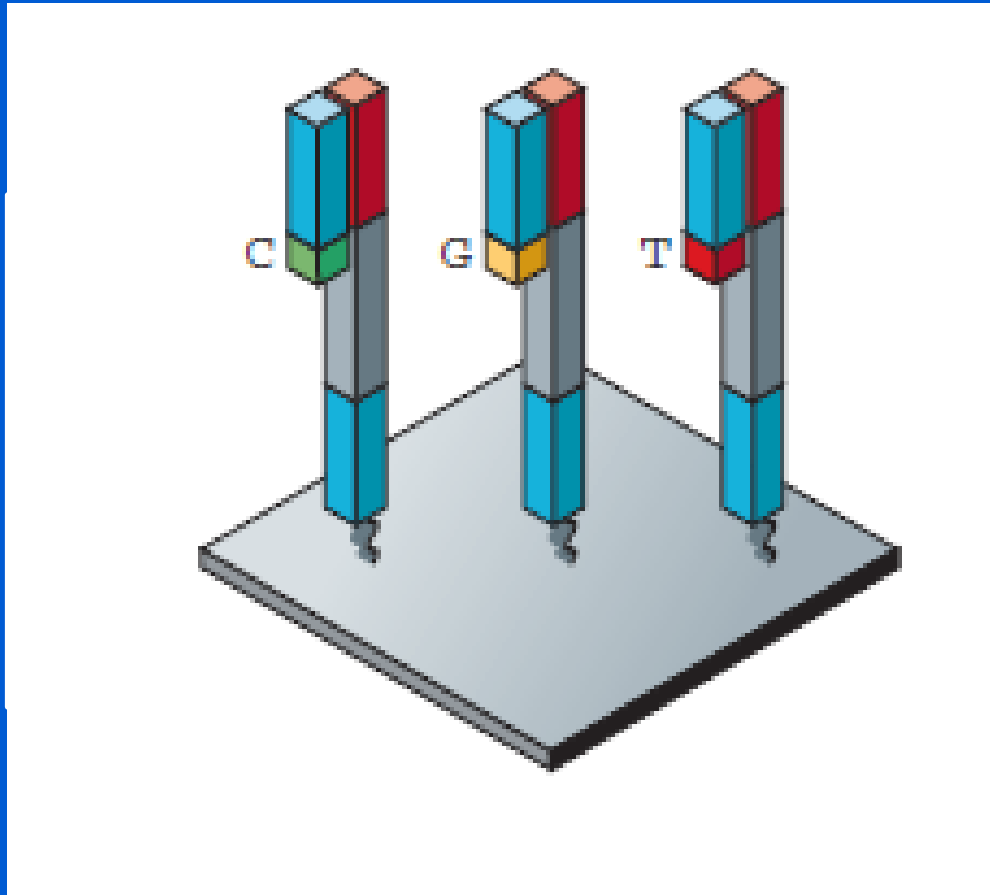
Metzker, *Nature Rev Genet* 2010;11:31



Wash out unused nucleotides; image

# Cyclic reversible termination

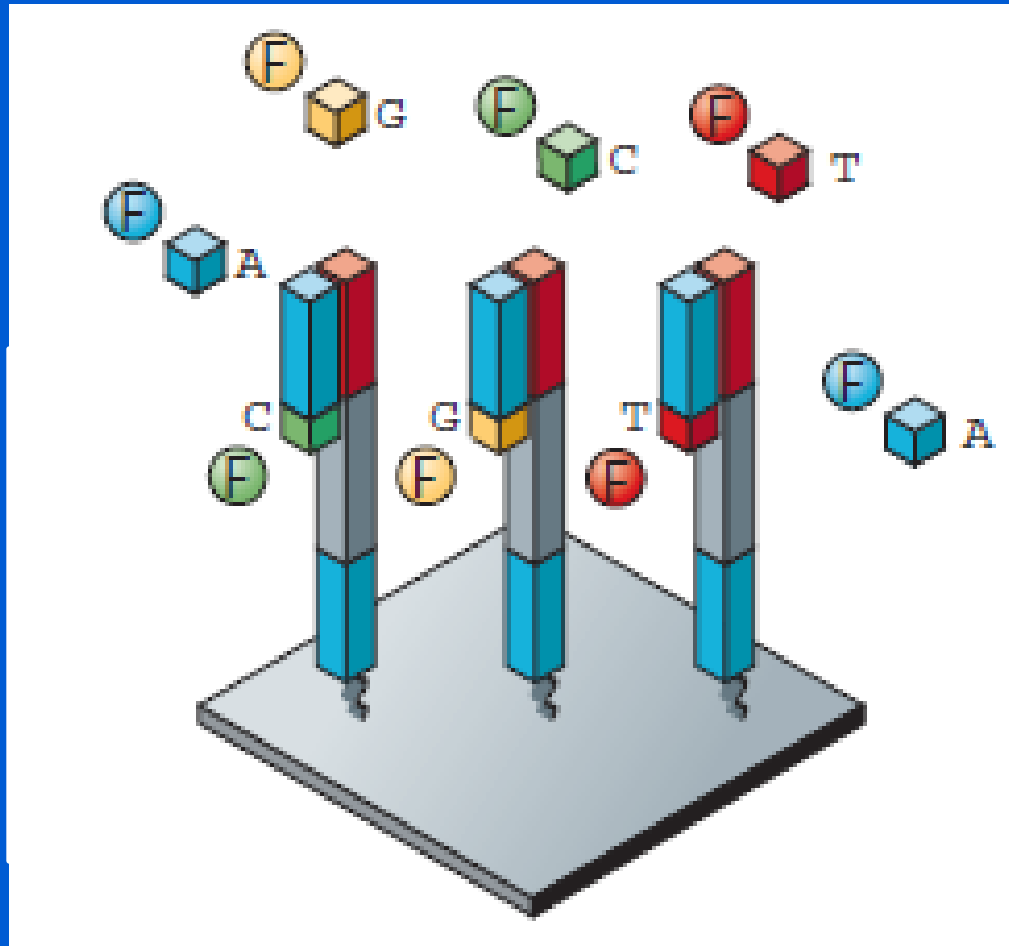
Metzker, *Nature Rev Genet* 2010;11:31



Cleave dye and terminating groups

# Cyclic reversible termination

Metzker, *Nature Rev Genet* 2010;11:31



Repeat

# Barcodes to sort out samples



**Sample 1: AGGTTCCA**

**Sample 2: GGCAATTT**

**Sample 3: TTGGAAAC**

# Outline

- Where to sample
- Sequencing
- **Data analysis**
- Next steps

# Output: fastq files

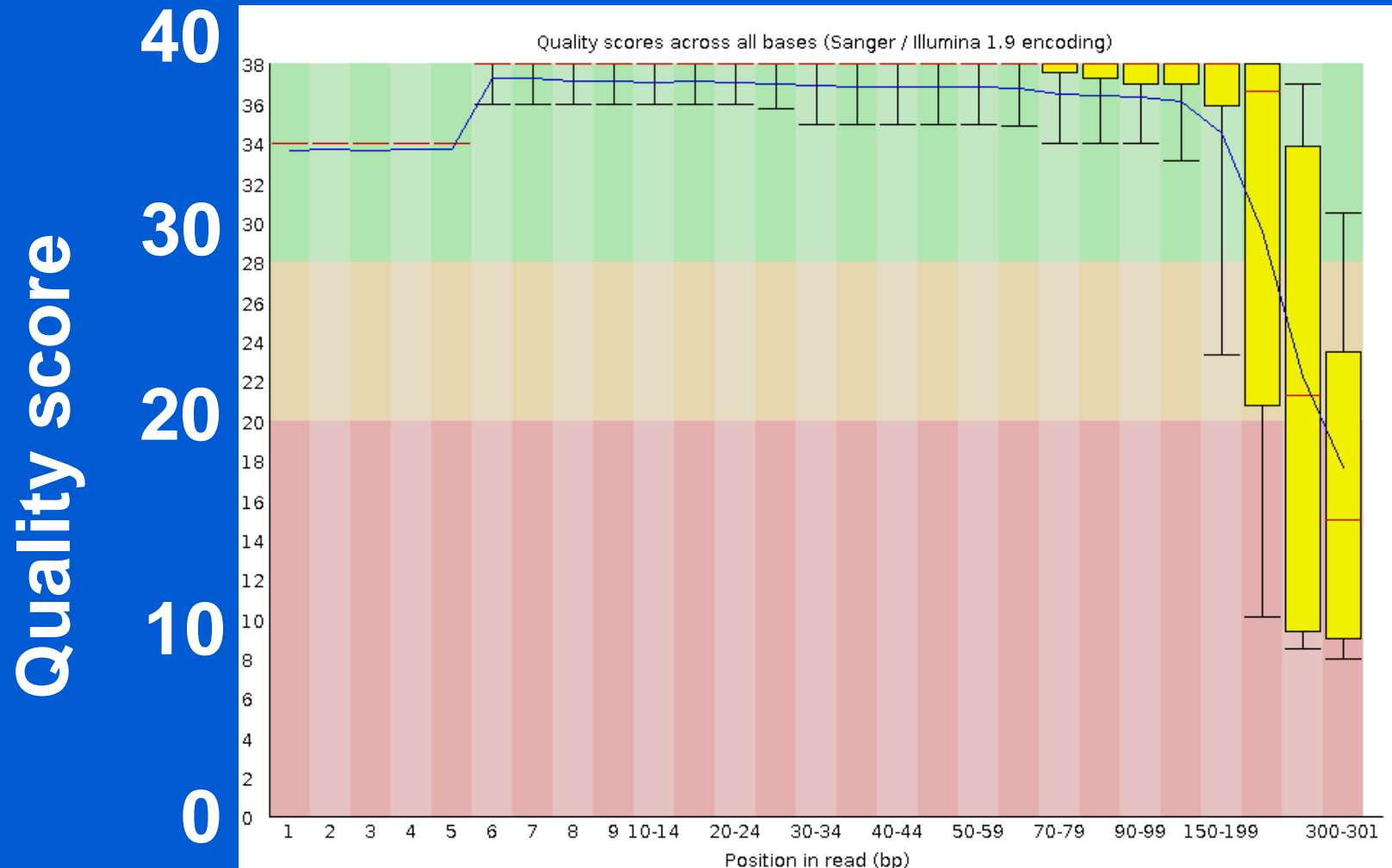
## Sample Fastq output (two DNA strands)

**Header:** @M02079:147:000000000-AK0J5:1:1101:15736:1676 1:N:0:49  
**Sequence:** TACAGAGGTCTCAAGCGTTGTTCCGGAATCACTGGGCGTAA  
**Additional line:** +  
**Quality:** >||>>EEGGFFE/////<||>-<0>DBF1<F<1.<<-<GD

**Header:** @M02079:147:000000000-AK0J5:1:1101:15989:1722 1:N:0:49  
**Sequence:** TACGGAGGATGCGAGCGTTATCCGGATTTATTGGGTTTAAAG  
**Additional line:** +  
**Quality:** B3EHFGGHF3FB43/E?EFGGFFGH3/B4?//B/FG?122FB



# Assess quality of reads



Position in read (BP)

# Quality filtering options

- Trim the low-quality tails
  - Option: remove sequence if more than a set percentage of bases are trimmed
- Remove sequences with:
  - Ambiguous bases
  - Quality scores below a threshold
  - More than a certain number of predicted errors

# **The bane of amplicon analysis is PCR errors**

- PCR errors, especially if they occur early in the cycle, can look like unique species
- Additionally, sequencing errors may have occurred, especially if the quality is low
- A common approach to addressing this issue is clustering of similar sequences into operational taxonomic units (OTUs)

# QIIME

- Open-source bioinformatics pipeline
- Designed for 16S sequence analysis
- Every step from fastq processing through data analysis



Quantitative Insights Into Microbial Ecology

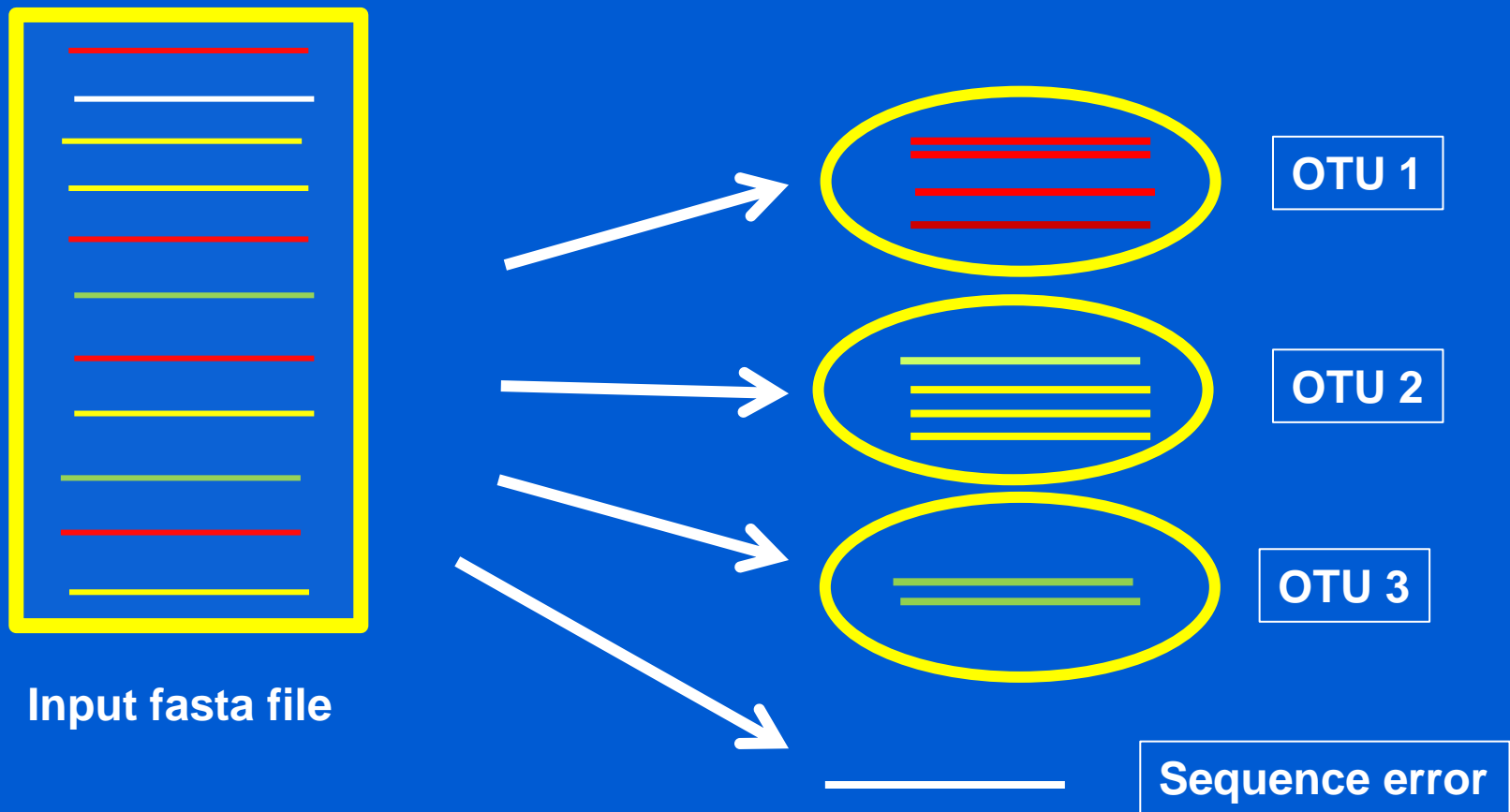
[www.QIIME.org](http://www.QIIME.org)

# Operational taxonomic unit picking

- Similar (typically 97%) sequences are clustered together
- Creates an “OTU table”
  - List of all OTUs
  - Frequency of each OTU in each sample
  - Taxonomic assignment of each OTU

OTUId	Sample1	Sample2	Sample3	Taxonomy
OTU_1	30	70	35	k__Bacteria;p__Firmicutes...
OTU_2	40	10	40	k__Bacteria;p__Bacteroidetes...
OTU_3	30	20	25	K__Bacteria;p__Firmicutes...

# Basic idea of OTU picking



# Limitations to OTU picking

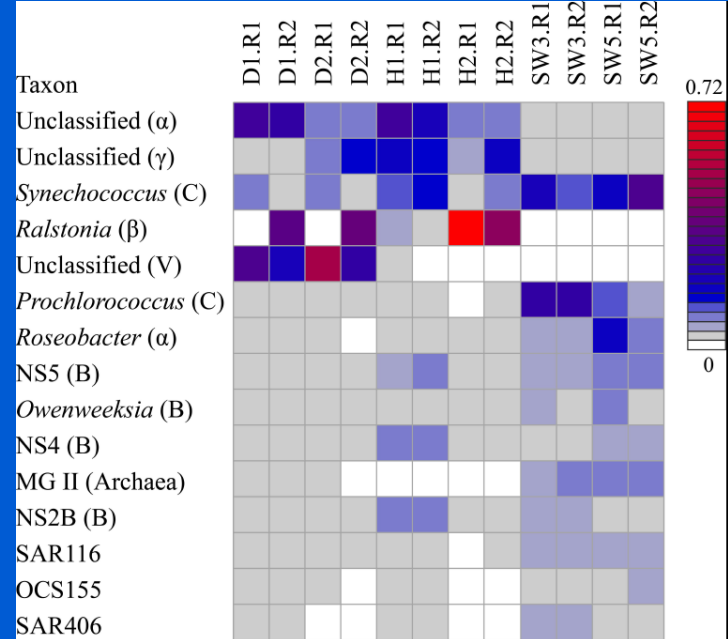
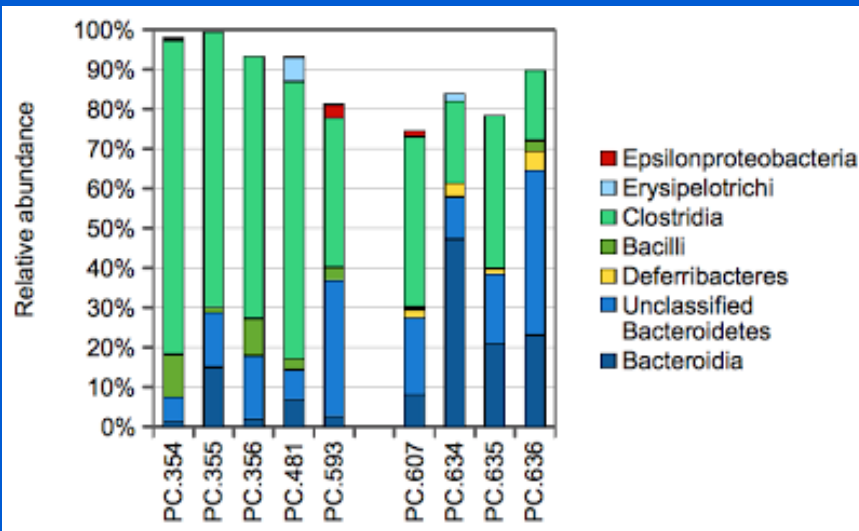
- OTUs are clustered based upon similarity
- Strain- and even species-level distinctions are often lost
- PCR amplification errors and low quality reads with high error rates can be mistaken as unique OTUs

# Downstream analyses with OTU tables

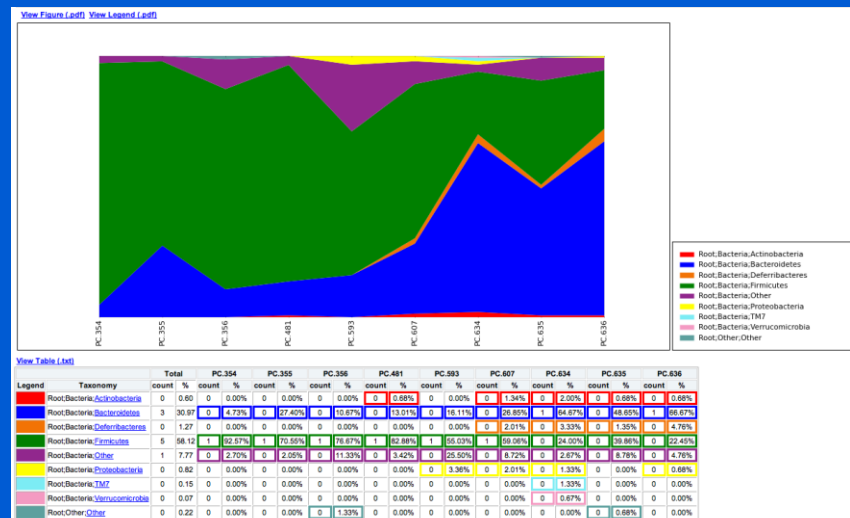
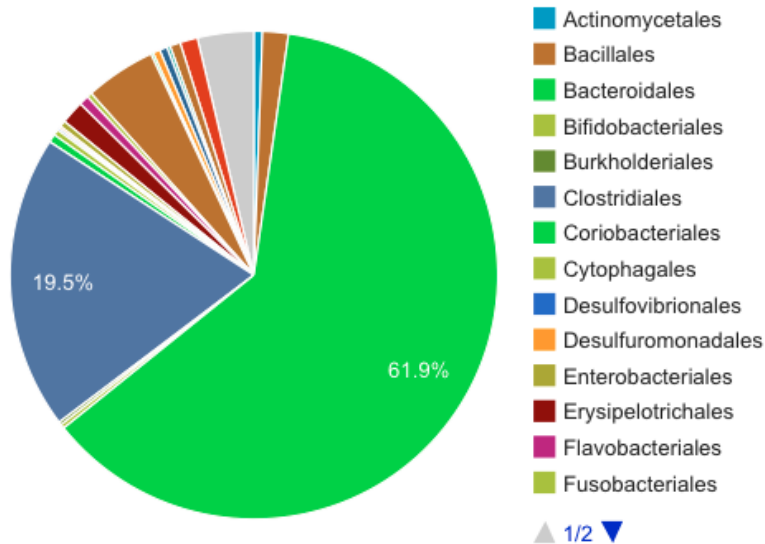
- Display taxonomy
- Alpha diversity (within group)
- Beta diversity (between groups)



# Displaying taxonomy

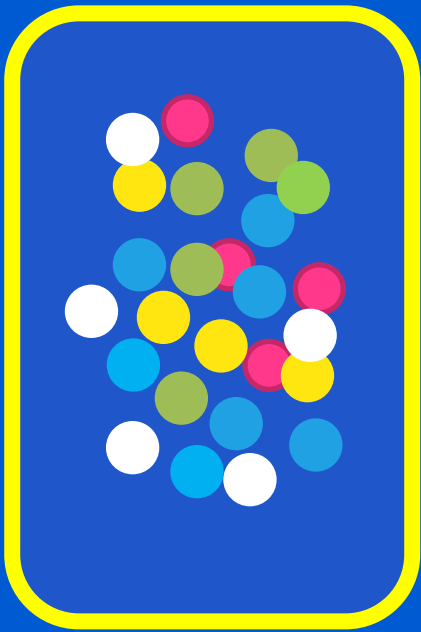


order [Download chart data](#)  
[View krona graph of order chart](#)

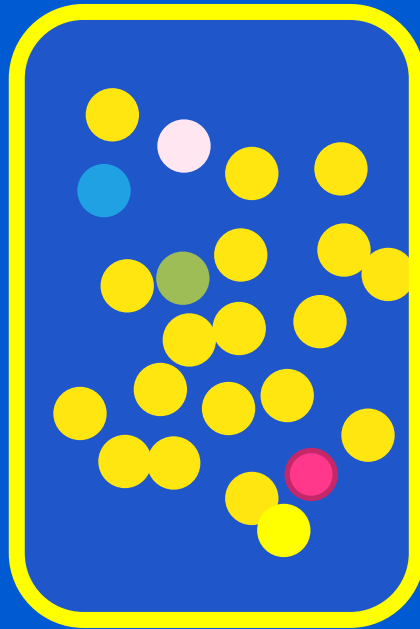


NOTE: the counts displayed pertain to either relative or absolute values depending on your selection from `summarize_taxa.py`. For relative values, the numbers are converted to integer, so counts

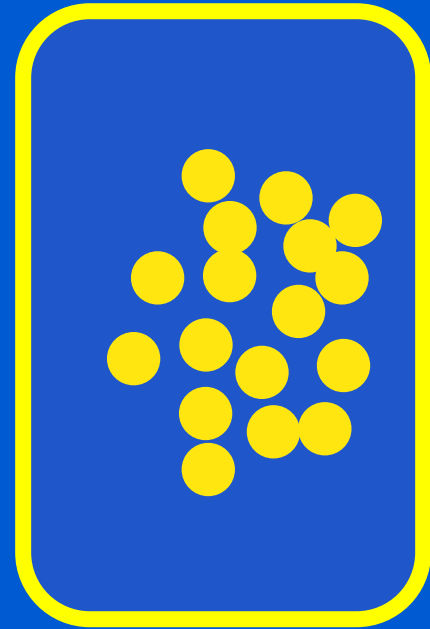
# Alpha diversity



Rich and even



Rich, not even

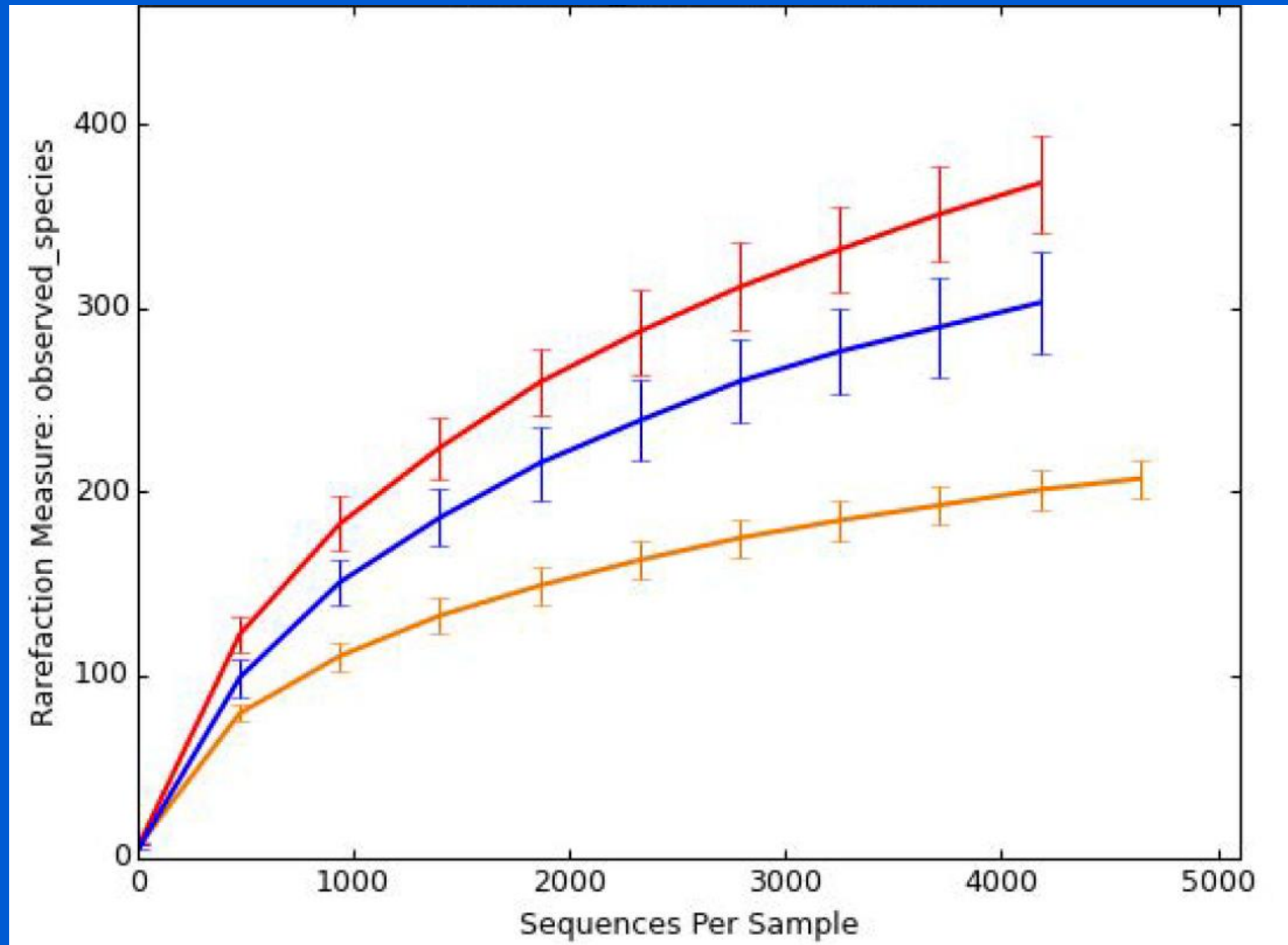


Not rich or even

# Cautionary note about measuring alpha diversity

- **Must take into account sequencing depth**
- **Typically, at UAB, depth is 50K – 150K sequences per sample for 16S**
- **To a point, diversity increases with higher depth, as you pick up more rare species**

# Illustration of rarefaction curves



# Beta diversity

- This speaks to the diversity between two different groups
- Typically depicted as PCOA plots
- Consider sequencing depth

# Sample distance matrix

	C1	C2	C3	C4	T1	T2	T3
C1	0	0.35	0.31	0.39	0.88	0.79	0.91
C2	0.35	0	0.42	0.22	0.92	0.90	0.74
C3	0.31	0.42	0	0.35	0.74	0.79	0.91
C4	0.39	0.22	0.35	0	0.82	0.84	0.92
T1	0.88	0.92	0.74	0.82	0	0.29	0.21
T2	0.79	0.90	0.79	0.84	0.29	0	0.32
T3	0.91	0.74	0.91	0.92	0.21	0.32	0

# Sample distance matrix

	C1	C2	C3	C4	T1	T2	T3
C1	0	0.35	0.31	0.39	0.88	0.79	0.91
C2		0	0.42	0.22	0.92	0.90	0.74
C3			0	0.35	0.74	0.79	0.91
C4				0	0.82	0.84	0.92
T1					0	0.29	0.21
T2						0	0.32
T3							0

Statistical tests compare distance between vs within groups  
e.g. Permanova

# Microbiota in juvenile spondyloarthritis

## Spondyloarthritis

- Type of pediatric and adult arthritis
- Distinctive demographic & clinical features
- High prevalence of gut inflammation
  - Frank IBD: 5 – 10% (IBD-associated arthritis)
  - Subclinical: 67%
- In children, goes by the name enthesitis-related arthritis (ERA)



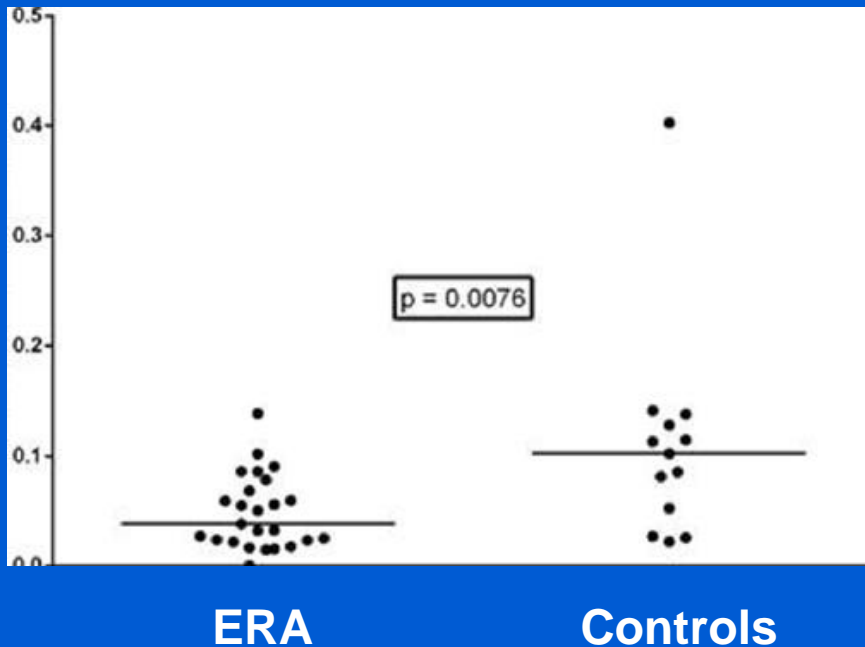
# Initial foray into microbiota analysis

- Obtained fecal specimens from children with ERA, along with healthy controls
- Patients included those who were newly diagnosed as well as those with long-standing disease
- Performed sequencing of 16S rDNA
- Analyzed with QIIME platform

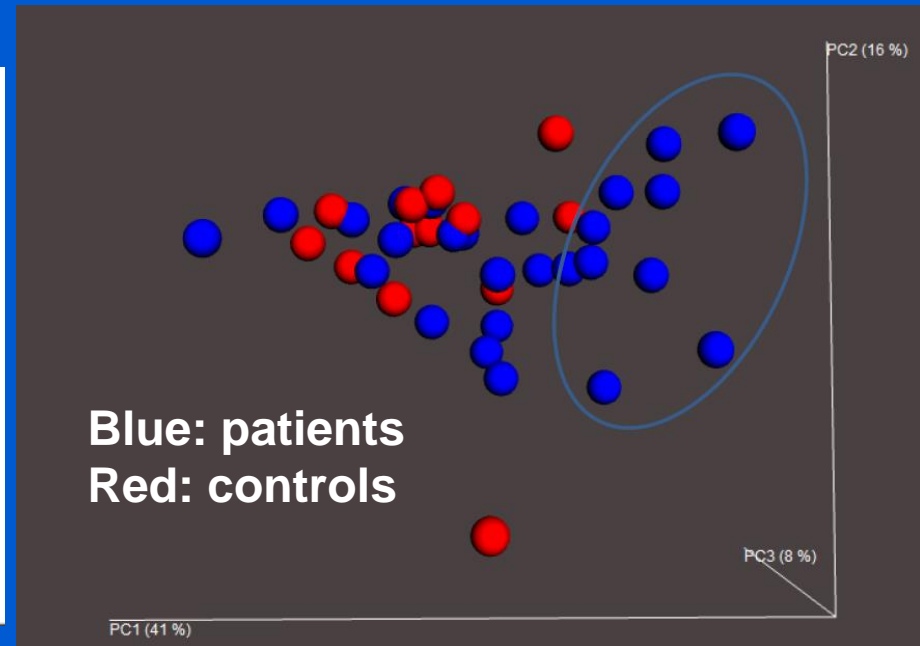
# Microbiota in ERA

- Decreased *Faecalibacterium prausnitzii* in children with ERA
- Small patient cluster identified by PCoA

Fecal abundance of *F. prausnitzii*



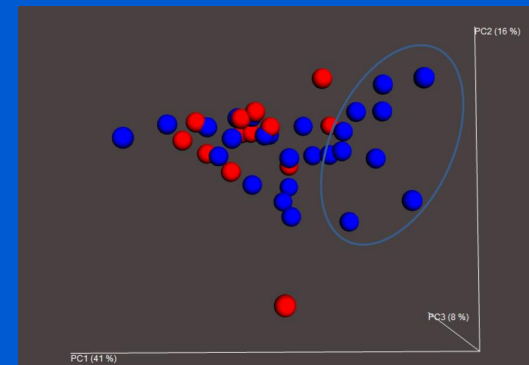
Principal coordinates analysis



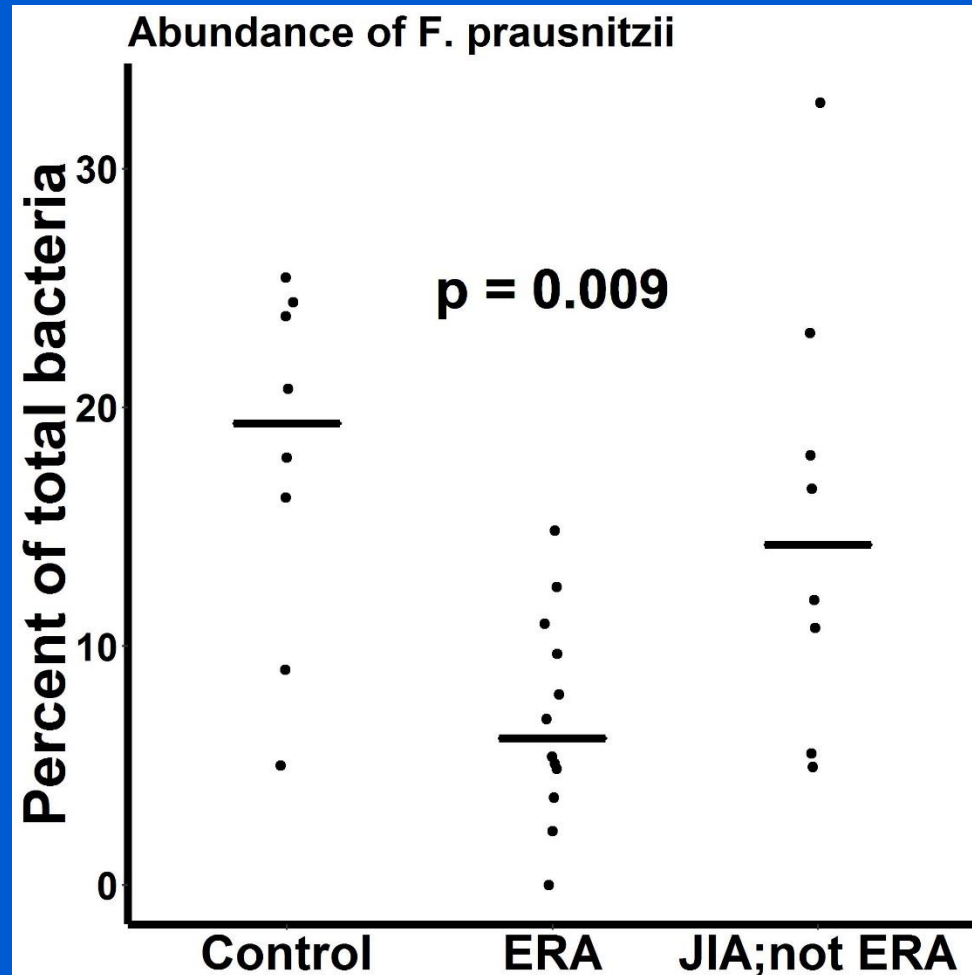
# Clusters are defined by *Bacteroides* and *Akkermansia*

- The two clusters had similar abundance of *F. Prausnitzii*
- The subjects who clustered with the controls tended to have high abundance of *Akkermansia* (> 1% in 7/17 vs 0/8)
- The subjects forming their own cluster had high *Bacteroides* abundance (41% vs 13%)

Stoll, *Arth Res Ther* 2014;16:486



# Altered *F. prausnitzii* is largely limited to SpA subtype of pediatric arthritis



# ***F. prausnitzii is an anti-inflammatory organism***

- Decreased fecal abundance in IBD<sup>1</sup>
- Increased IL-10 production by PBMCs<sup>2,3</sup>
- Reduces inflammation in colitis model<sup>2</sup>
- Major butyrate producer<sup>4</sup>
  - Beneficial effects on enterocytes<sup>5</sup>
  - Increases colonic regulatory T cells<sup>6</sup>

<sup>1</sup>Cao, *Gastroenterol Res Practice* 2014:872725 (review)

<sup>2</sup>Sokol, *PNAS* 2008;105:16731

<sup>5</sup>Hamer, *Aliment Pharmacol Ther* 2008;27:104

<sup>3</sup>Rossi, *Sci Rep* 2016;6:18507

<sup>6</sup>Smith, *Science* 2013341:569

<sup>4</sup>Hold, *Appl Environ Microbiol* 2003;69:4320

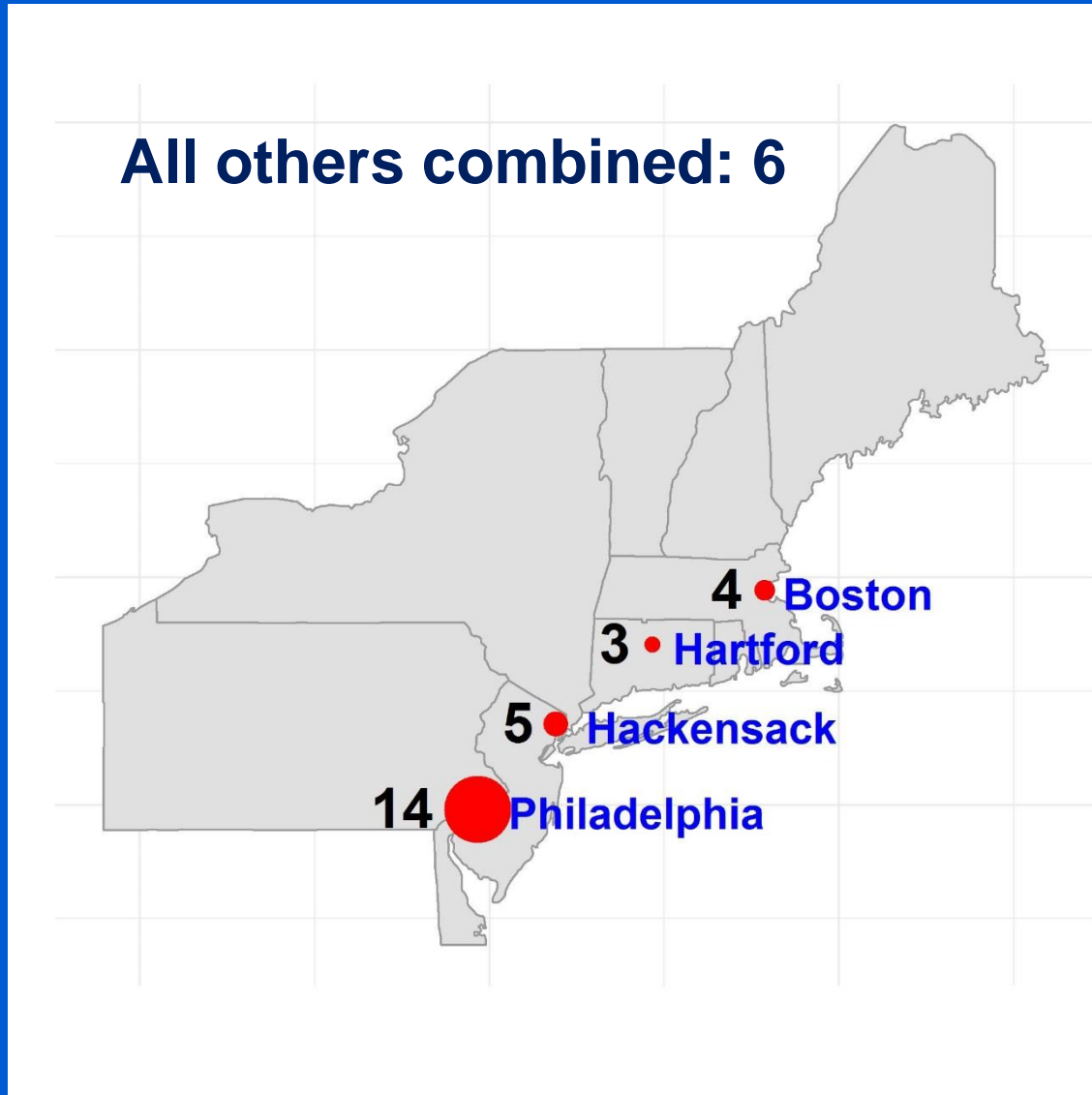
# Focus on treatment-naïve subjects

- Recruited pediatric spondyloarthritis patients from centers around the country
- Attempted to recruit healthy controls from each site
- Analyzed with QIIME (initially)

# Collaborating centers



# Collaborators are not all created equal





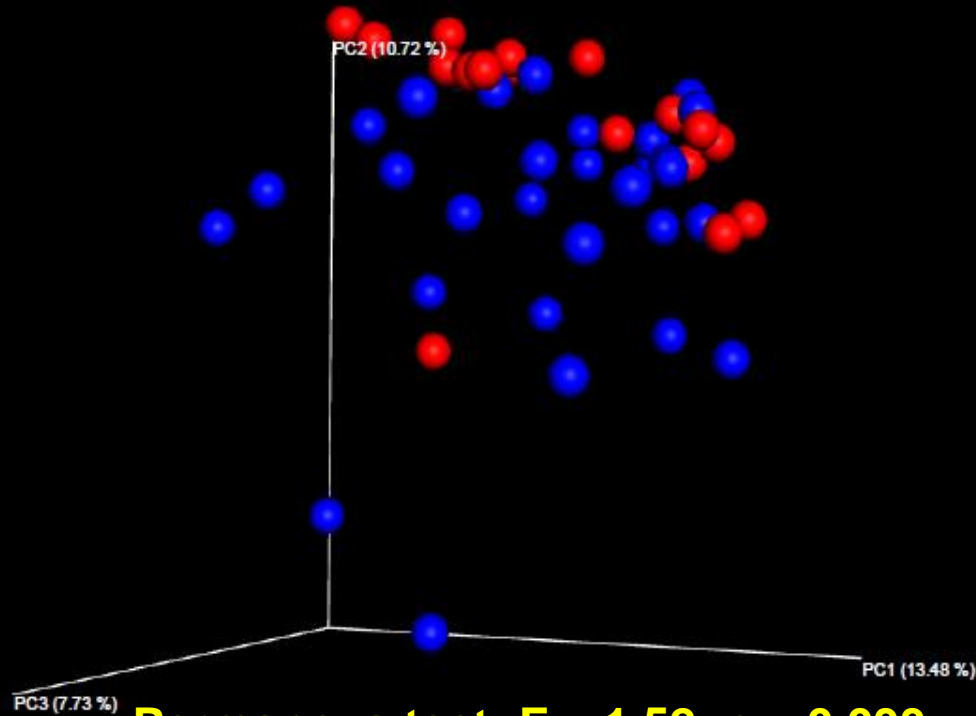
# Subjects

Characteristic	ERA	Controls
n	30	19
Age (years; mean $\pm$ SD)	13.5 $\pm$ 3.0	13.6 $\pm$ 2.7
BMI (kg / m <sup>2</sup> ; mean $\pm$ SD)	20.7 $\pm$ 4.1	21.5 $\pm$ 6.1
Male	19, 63%	13, 68%
Caucasian	23, 77%	17, 89%

# Clustering by diagnosis

Blue: patients  
Red: controls

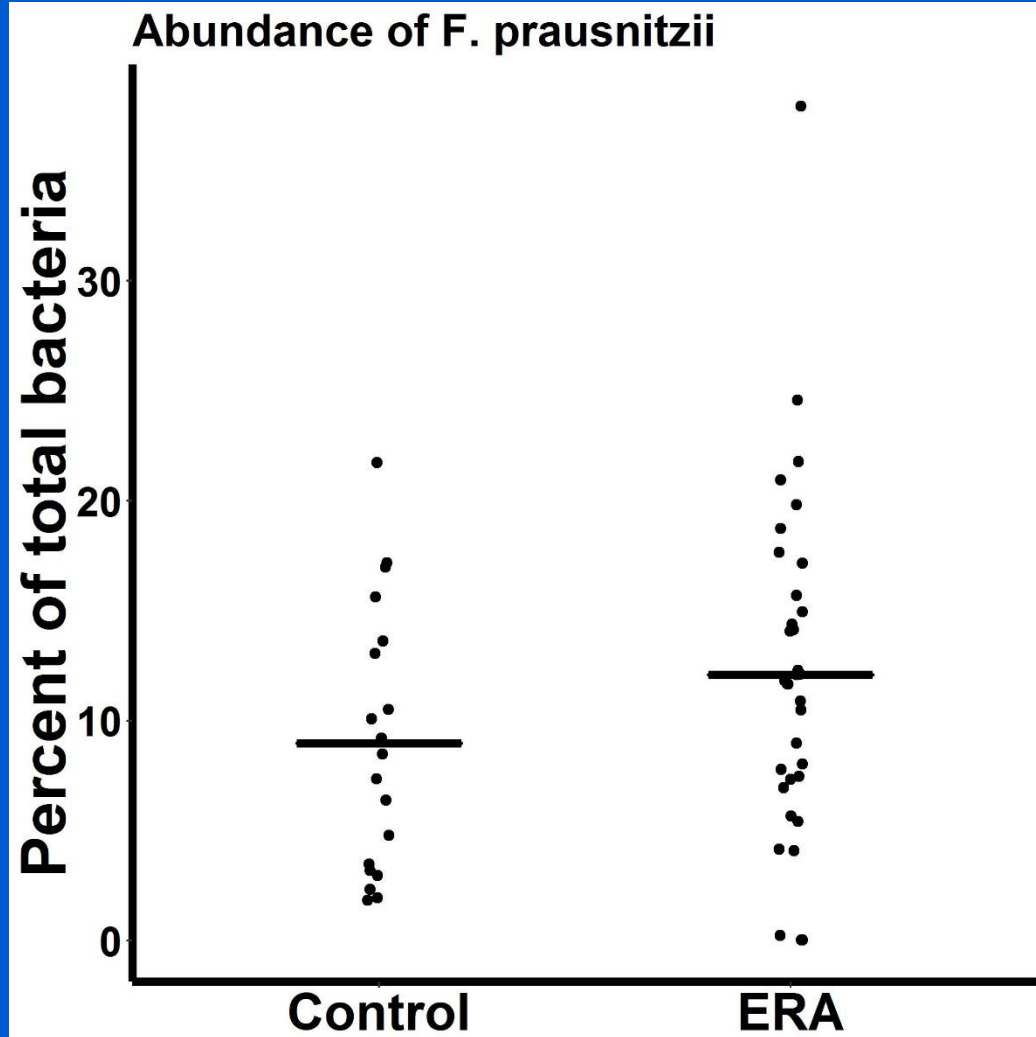
Bray Curtis



# Taxonomic assessment

- Pairwise comparisons revealed minimal differences between the groups
  - Especially after adjustment for multiple comparisons
  - At the phylum level, decreased Actinobacteria in ERA (3.3 vs 9.0%,  $p < 0.002$ )
- Focused on organism of interest based upon previous work
  - *Faecalibacterium prausnitzii*

# Not what I expected



# Limitations to OTU picking

- OTUs are clustered based upon similarity
- Strain- and even species-level distinctions are often lost
- PCR amplification errors and low quality reads with high error rates can be mistaken as unique OTUs

# Alternative to OTU picking

Published in final edited form as:

*Nat Methods*. 2016 July ; 13(7): 581–583. doi:10.1038/nmeth.3869.

## **DADA2: High resolution sample inference from Illumina amplicon data**

**Benjamin J Callahan<sup>1,\*</sup>, Paul J McMurdie<sup>2</sup>, Michael J Rosen<sup>3</sup>, Andrew W Han<sup>2</sup>, Amy Jo A Johnson<sup>2</sup>, and Susan P Holmes<sup>1</sup>**

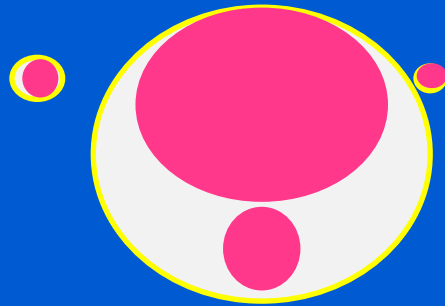
<sup>1</sup>Department of Statistics, Stanford University, Stanford, CA, USA

<sup>2</sup>Second Genome, South San Francisco, CA, USA

<sup>3</sup>Department of Applied Physics, Stanford University, Stanford, CA, USA

- **Use error modeling to distinguish sequence errors from sequence variants**
  - **Real OTUs versus fake OTUs**

# DADA2 offers a replacement to traditional OTU picking methods

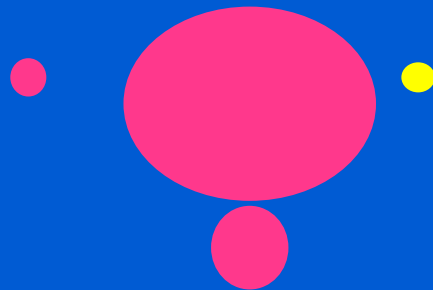


# DADA2 – basic concept

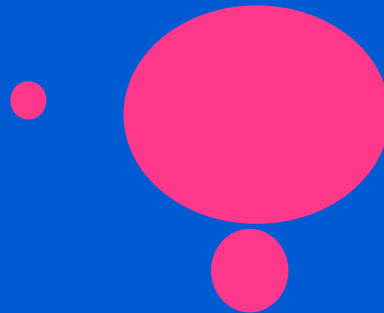
- Start with the most abundant sequence (centroid)
- For each other sequence, test the null hypothesis that it is the same as the centroid
- Two factors are used to compare sequences with the centroids
  - Distance
    - Number of nucleotide differences
    - Quality of the sequences
    - Specific nucleotide transitions (e.g.  $T \rightarrow A$ ,  $T \rightarrow C$ , etc)
  - Abundance
- If the lowest (smallest) p-value is below a cutoff, then the null hypothesis is rejected, and the sequence becomes a centroid
- Otherwise, the sequence is combined with the nearest centroid



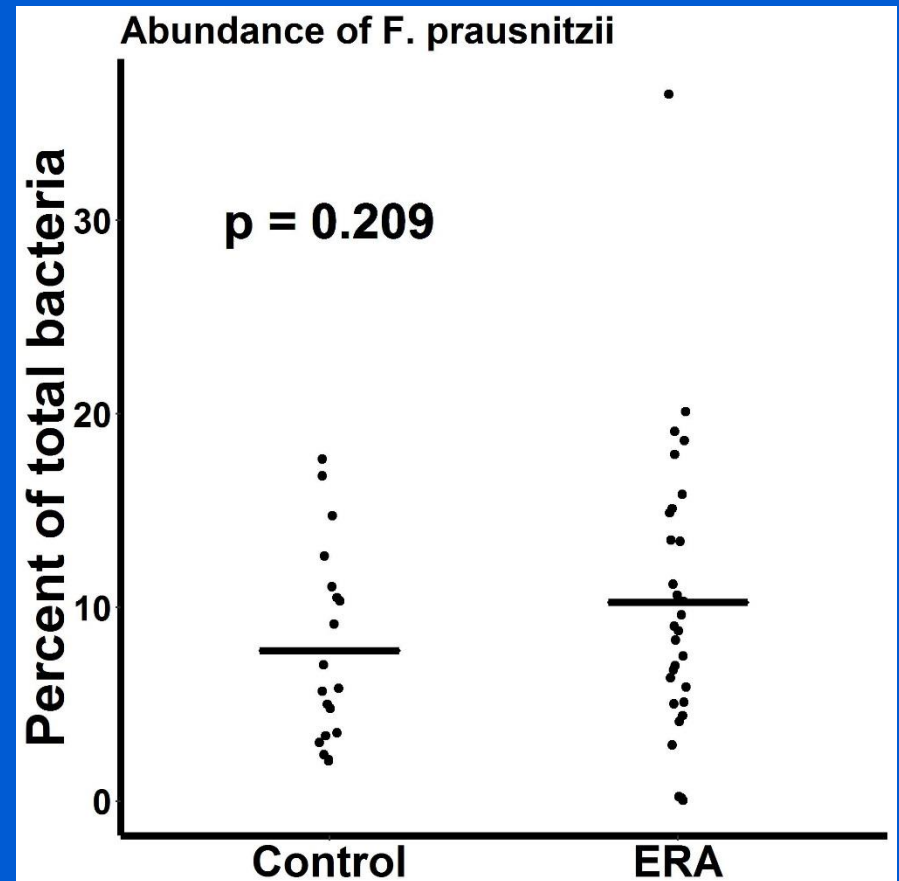
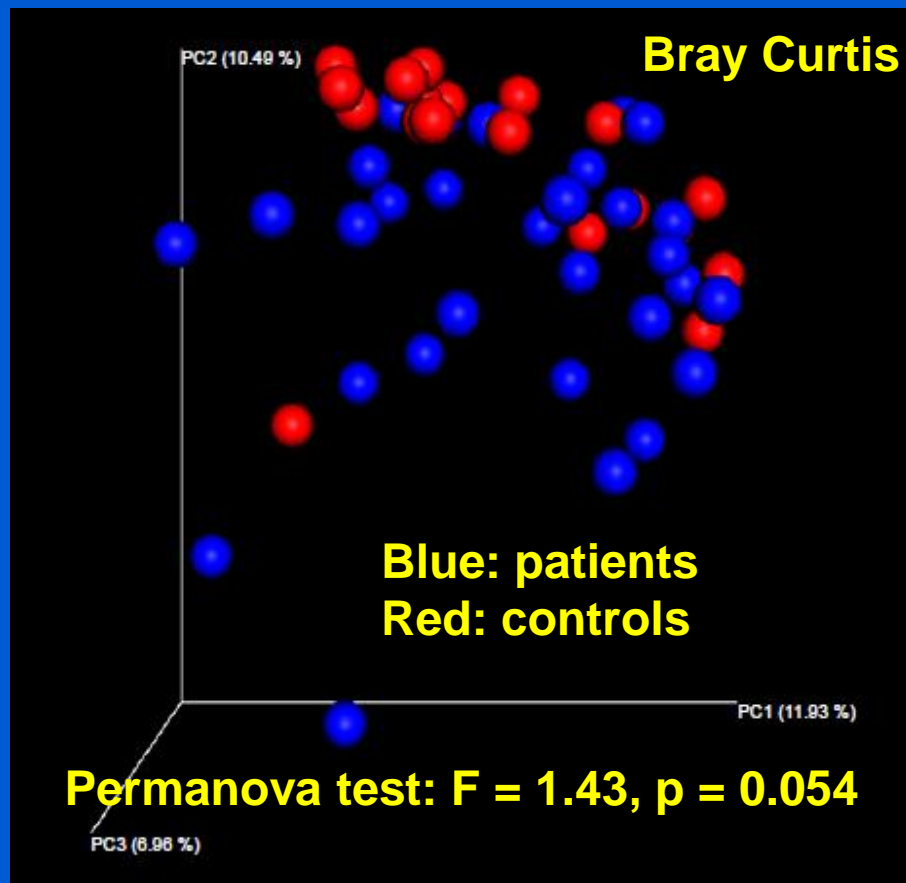
# DADA2 offers a replacement to traditional OTU picking methods



# DADA2 offers a replacement to traditional OTU picking methods

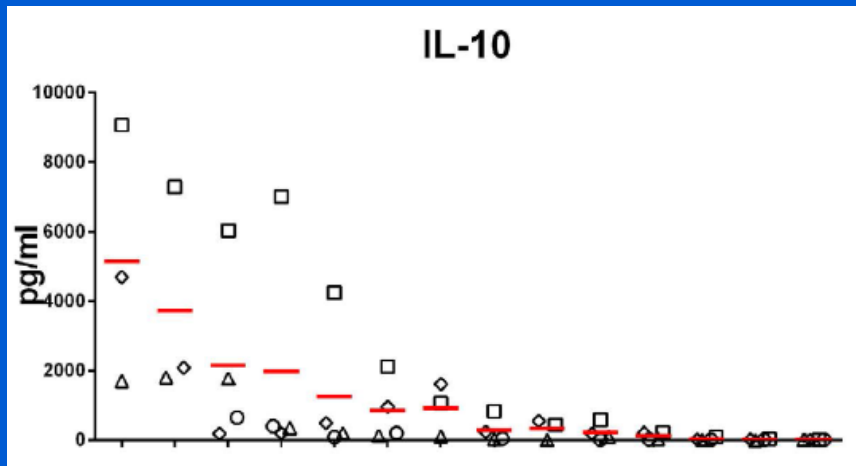


# At first blush, results from the DADA2 analysis were similar



# There are differences within strains of *F. praustnizii*

IL-10 production from hDC



A2-165

L2/6

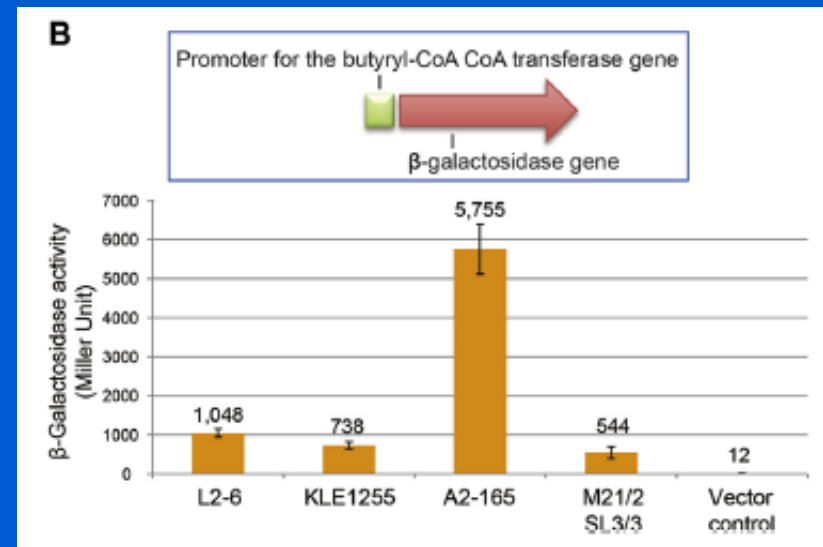
M21/2

S3I/3

HTF-F

Rossi, *Sci Rep* 2016;6:18507

Butyryl-CoA transferase gene promoter activity



L2/6

KLE1255

A2-165

M21/2

Vector control

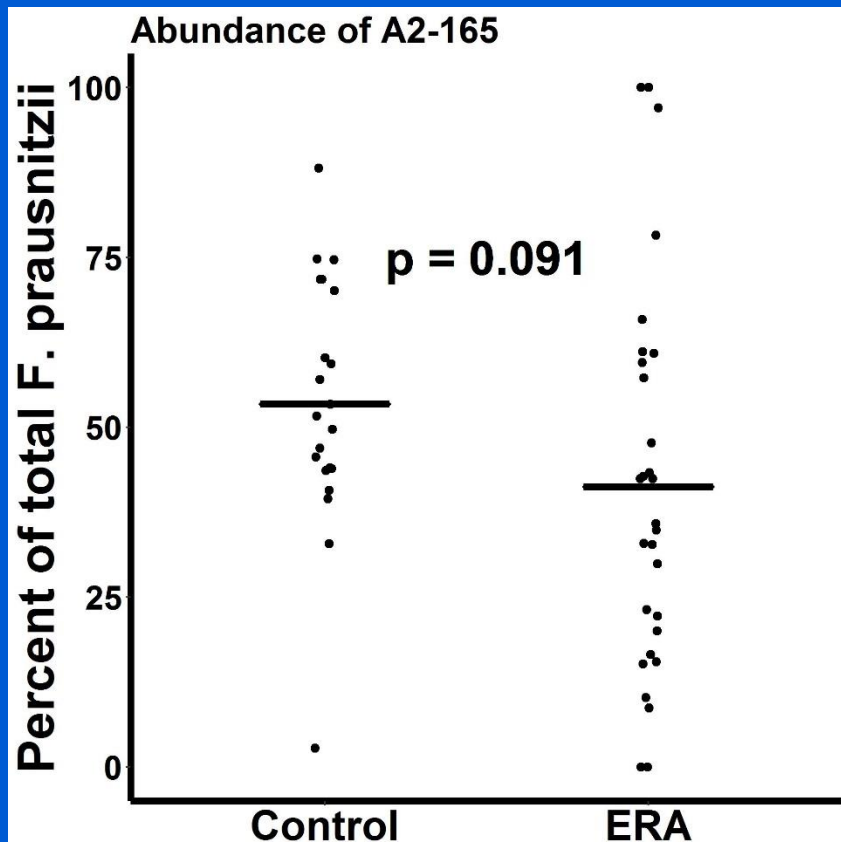
Song, *JACI* 2016;137:852

## 23 sequence variants matching *F. prausnitzii* were detected

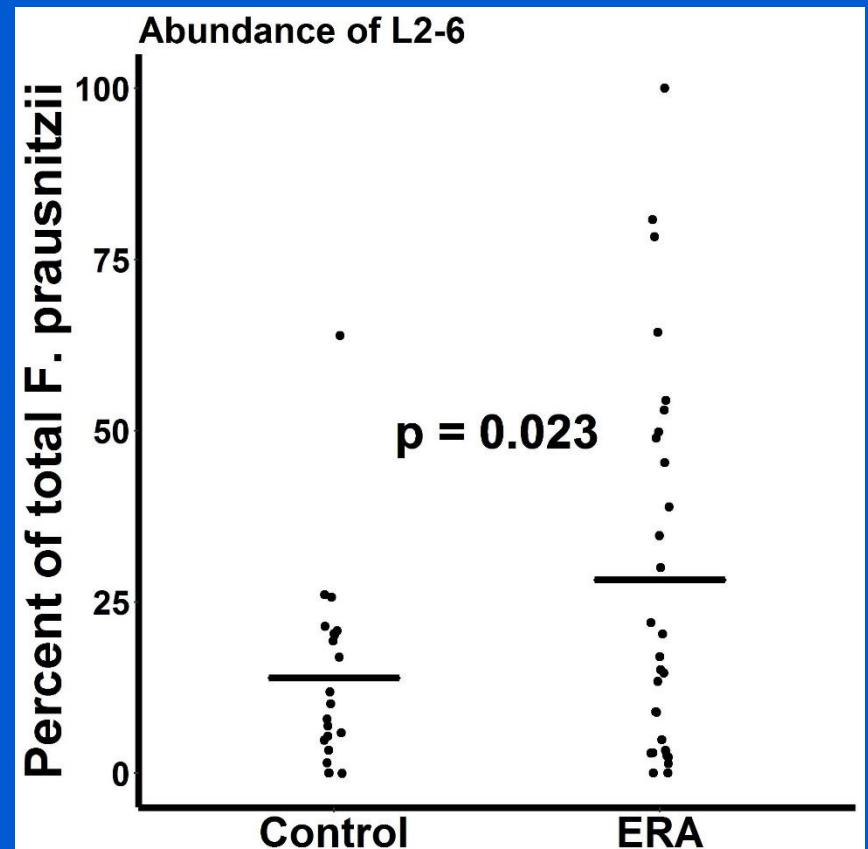
- I downloaded the 16S sequences for A2-165 and L2/6
- Ran BLAST of the 23 SVs against these two sequences
- One SV was a 100% match (253 / 253) to A2-165
- Another SV was a 99% match (252 / 253) to L2/6

# Distribution of FP strains differs between patients and controls

## A2-165 (“good”)



## L2/6 (less “good”)



# Whole genome sequencing

- Shotgun sequencing of all the DNA present in a sample
- May not include viral particles
- Will include human contaminant DNA

# Removal of host DNA sequences

- **Not required with 16S analysis**
  - Host DNA should not be amplified
- **Contamination can occur with WGS**
  - Variable with fecal microbes
  - High likelihood with other habitats



# Options for host DNA removal

- Reference database of host DNA
  - Filter out alignments
  - BLAST or Bowtie2 / BWA
- Reference database of microbes
  - Include sequences that align with dominant bacteria
  - Output will be limited to these bacteria

# Assembly

- **Most packages not designed for microbiota**
  - Hundreds of species
- **Unclear if even required**

# Options for taxonomic assessment

- **Metaphlan<sup>1</sup>**
  - Assigns taxonomy based upon marker genes
  - Second version includes 17K reference genomes
- **Alignment-based**
- **Composition-based: bacteria have unique sequence features**
  - GC content
  - Nucleotide repeats
  - Codon usage

<sup>1</sup>Truong *Nature Methods* 2015;10:902

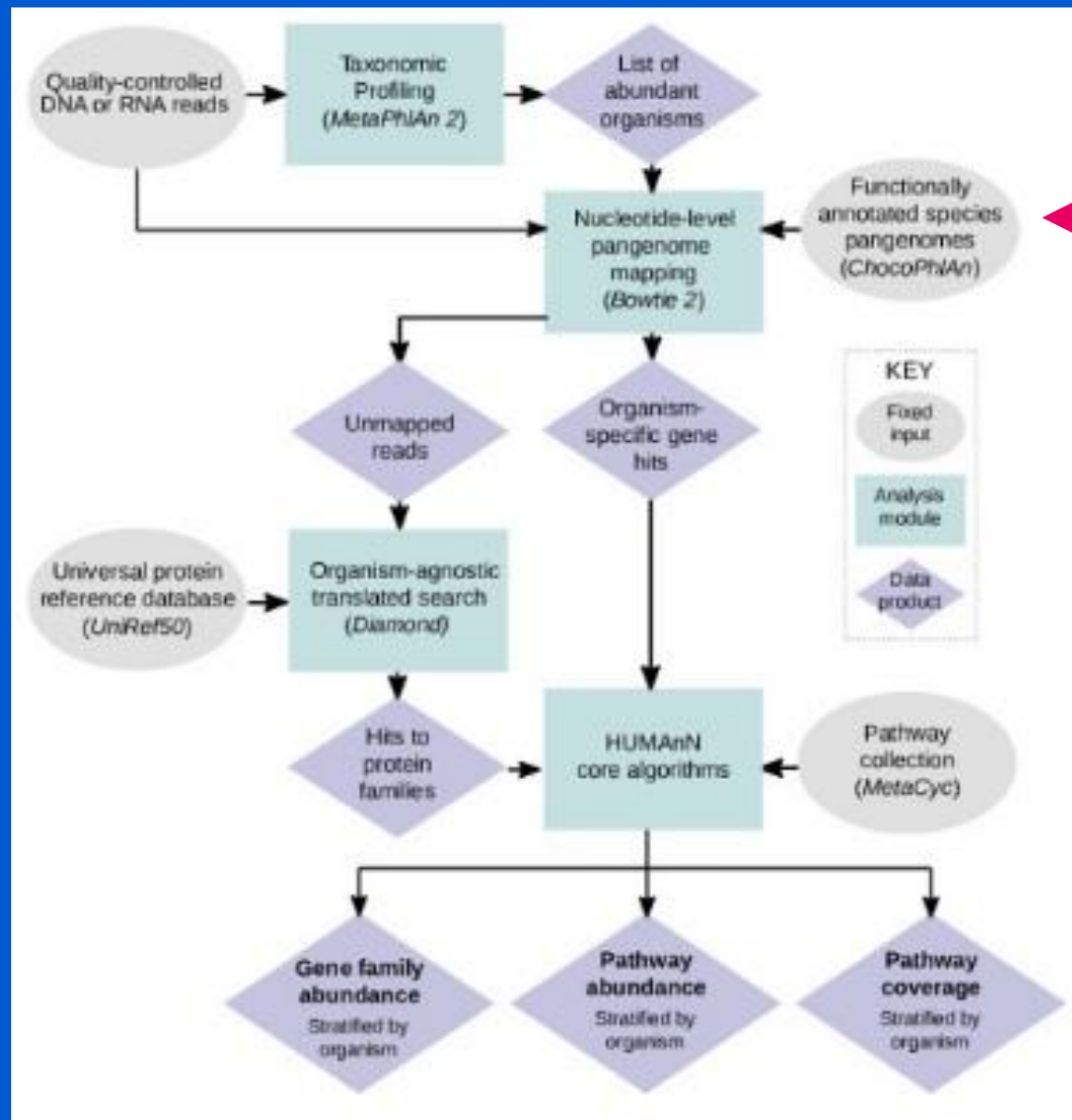
# Functional annotation

- Probably the most straightforward approach is with HUMAnN2
- “What are the microbes in my community of interest capable of doing?”

<sup>1</sup>Abubucker *Plos Comput Biol* 2012;8:e1002358

<http://huttenhower.sph.harvard.edu/humann2>

# HUMAnN2 workflow

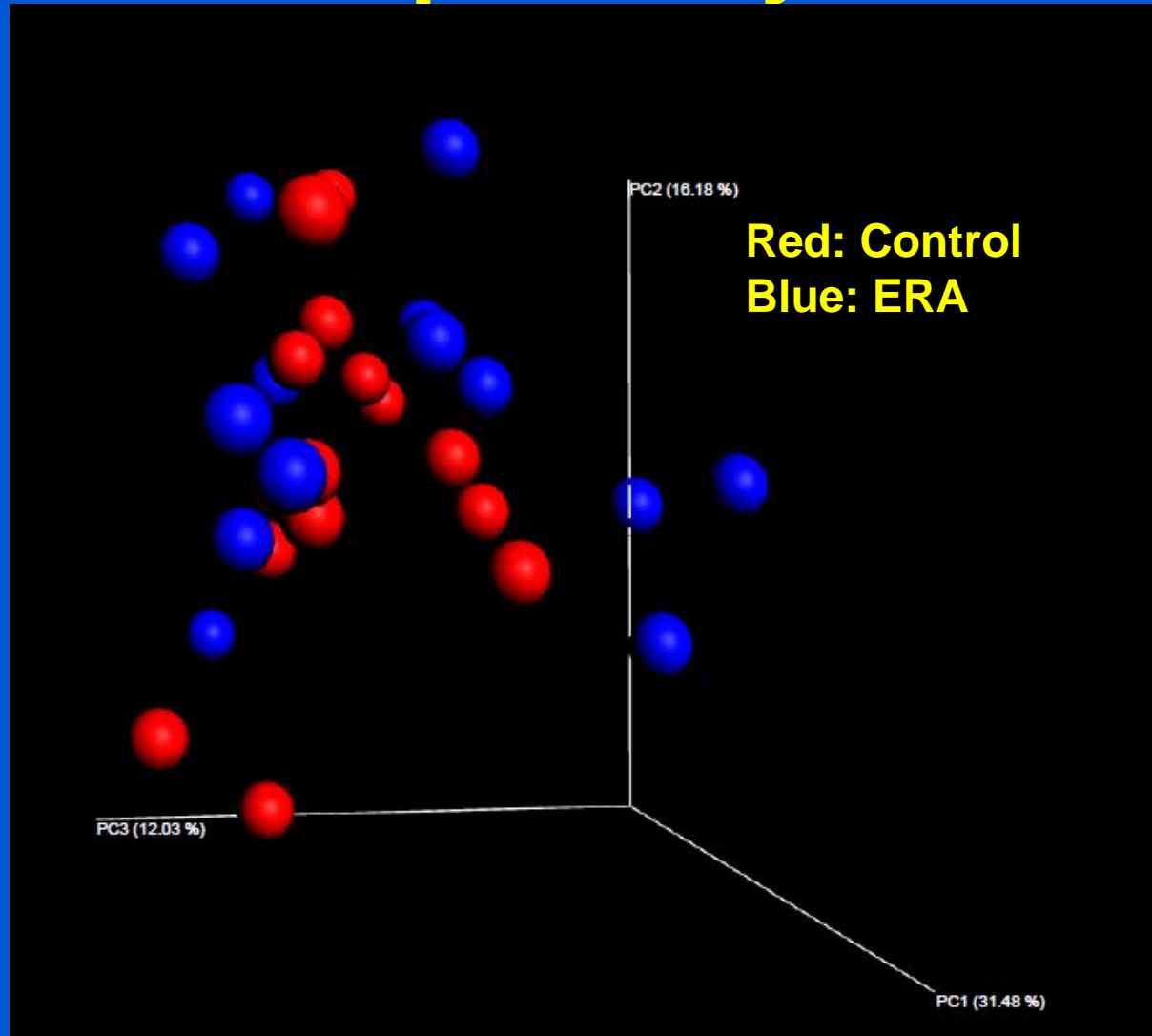


NCBI genomes

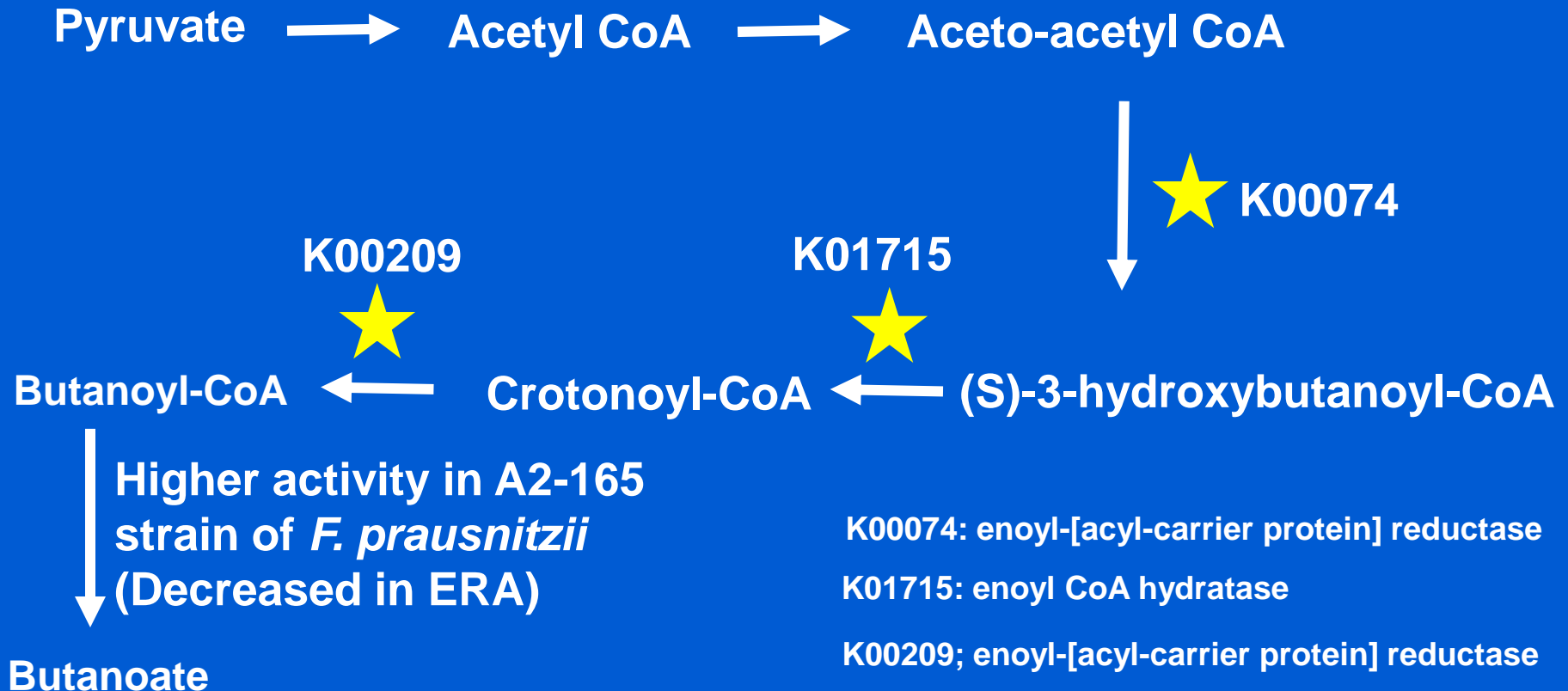
# **Performed whole genome sequencing of fecal microbiota from ERA patients and controls**

- **14 ERA patients and 14 healthy controls**
- **Subset of newly diagnosed patients used for 16S analysis**
- **Custom quality control pipeline**
- **Analysis with HUMAnN2**

# No obvious global differences at the pathway level



# Butanoate metabolism





# Summary

- Study confirms association of *F. prausnitzii* in children with ERA
  - Findings are specific to the A2-165 strain
- Patients with ERA may have decreased ability to make butyrate

# Outline

- Where to sample
- Sequencing
- Data analysis
- **Next steps**

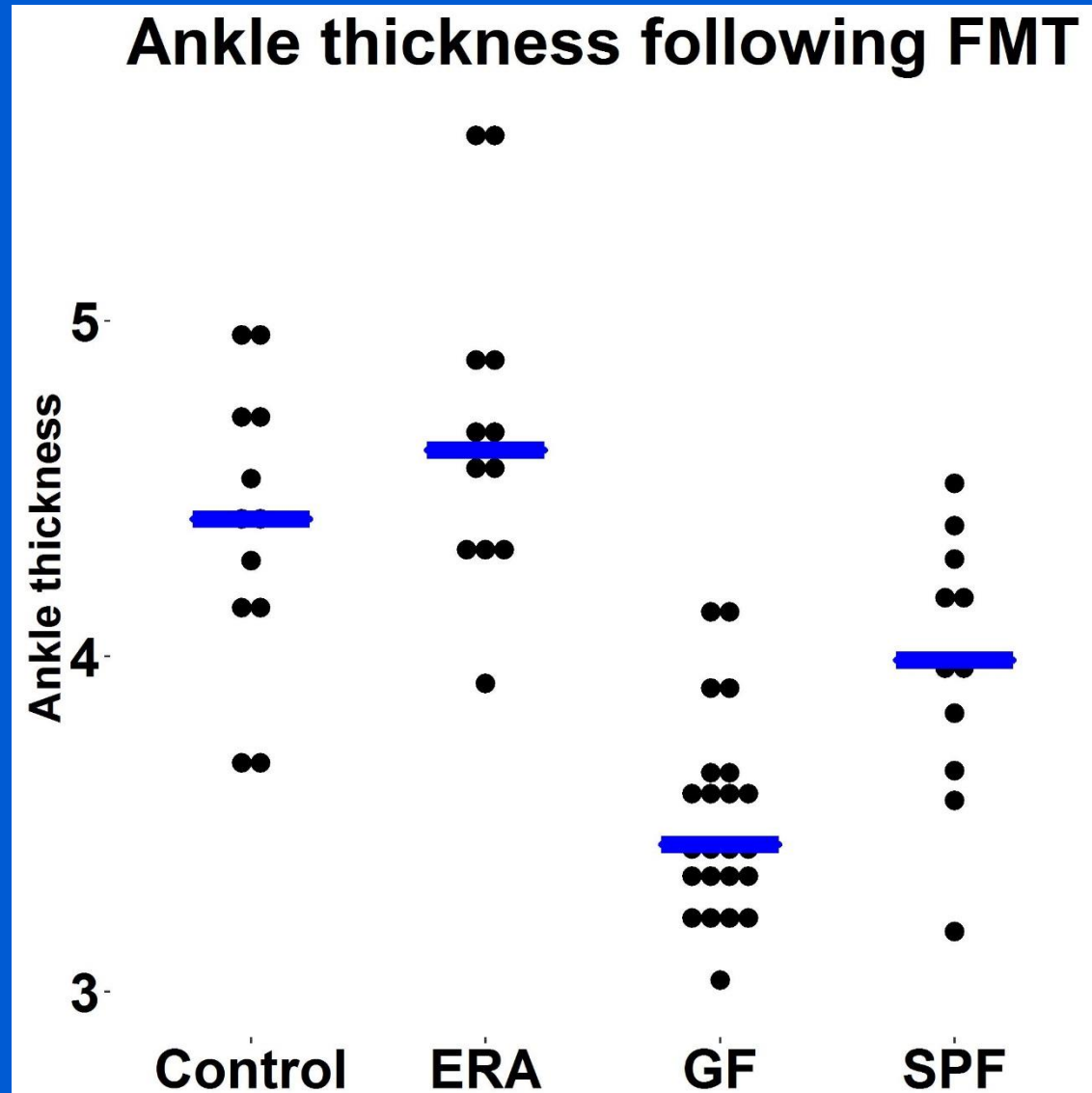
# **Assessing the role of the microbiota: FMT into mice**

- Patient and control fecal specimens were collected in media that permits live recovery of organisms and frozen in glycerol
- Transferred to germ-free K/BxN mice
  - Arthritis model
  - Dependent on microbiota
  - Rapid onset
- Studied 3 – 4 weeks after transfer

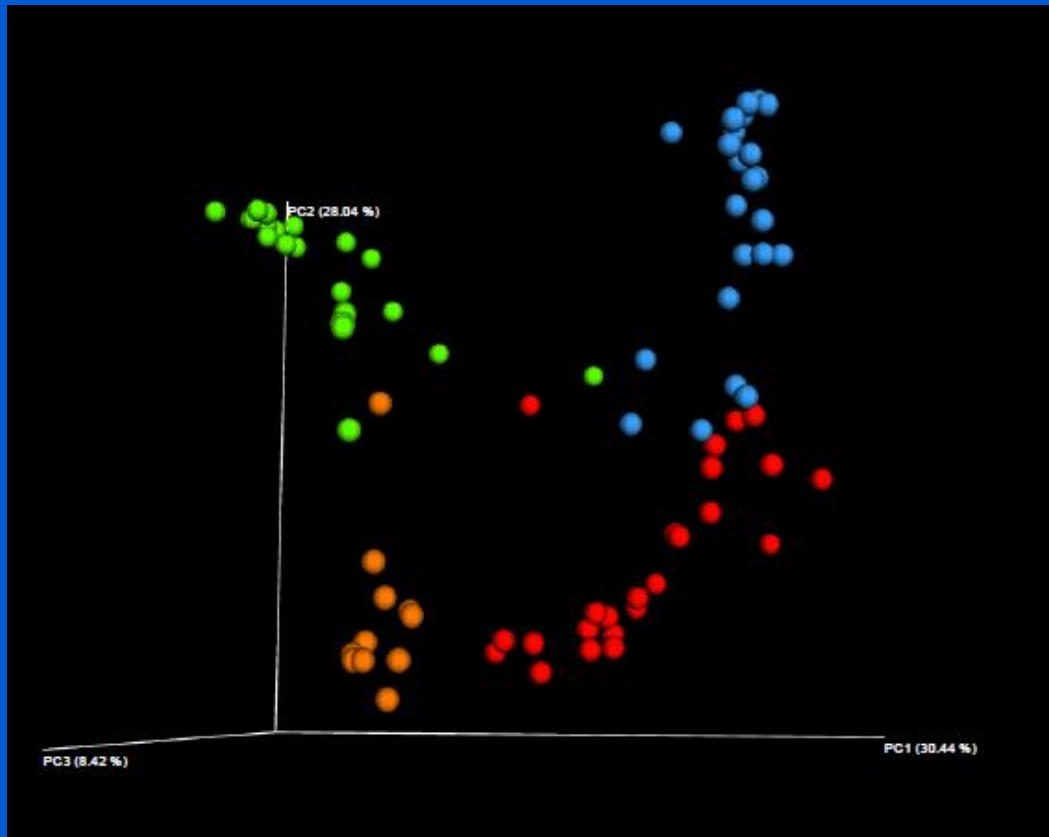
# Groups

- **Human donors**
  - ERA, n = 12
  - Controls, n = 12
- **Mouse recipients**
  - ERA, n = 12
  - Controls, n = 12
- **Germ-free (GF)**
- **Specific pathogen free (SPF)**
  - Transferred from GF facility to SPF facility

# Ankle swelling after FMT

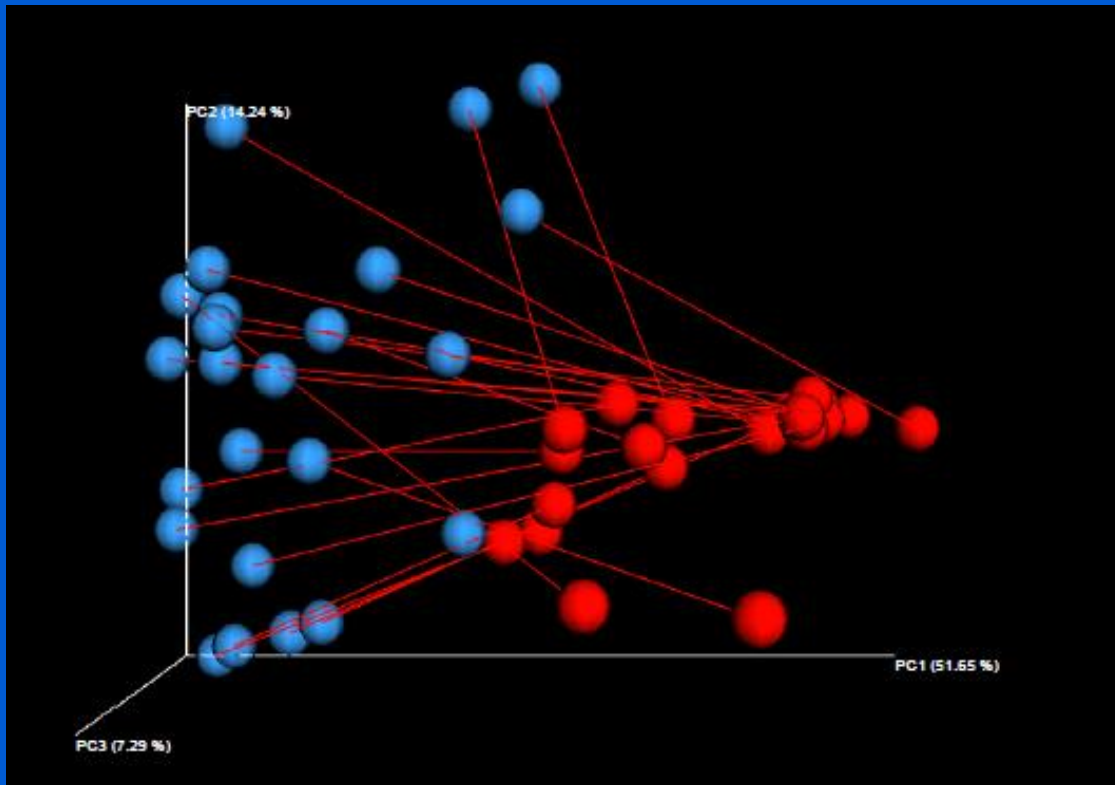


# Incomplete humanization



**Green: GF mice**  
**Blue: Transplanted mice**  
**Red: Humans**  
**Orange: SPF mice**

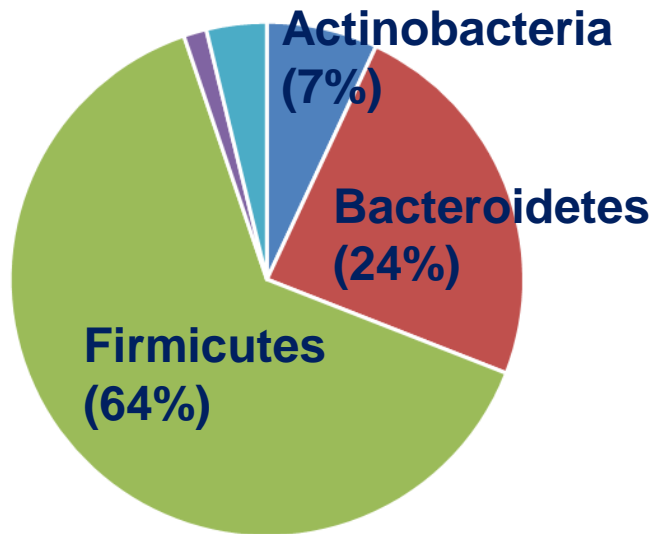
# No clustering between individual patient/donor dyads



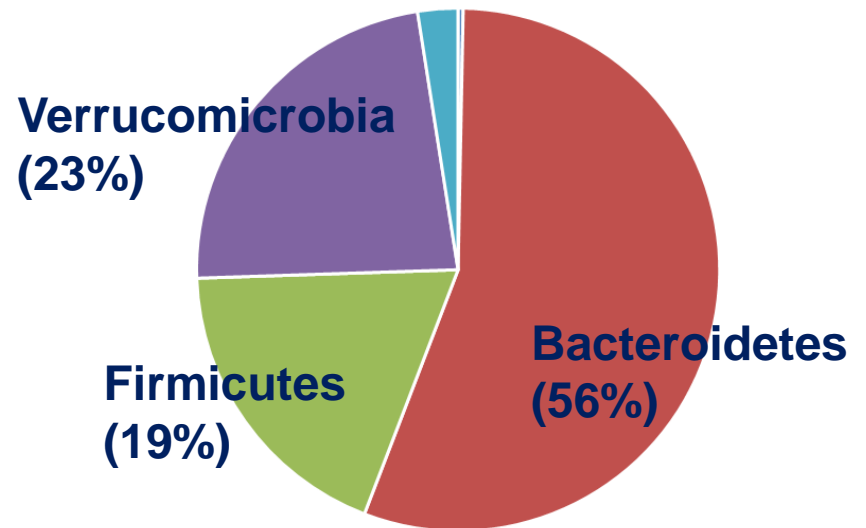
Red: Humans  
Blue: Transplanted mice

# Taxonomy: Phylum level

## Humans



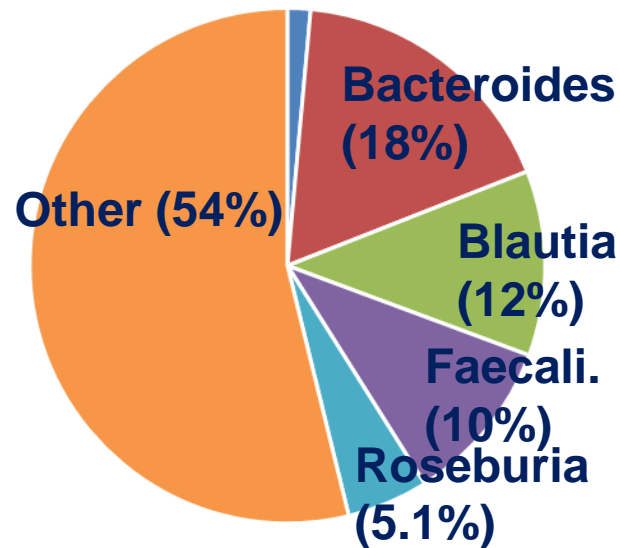
## Mice



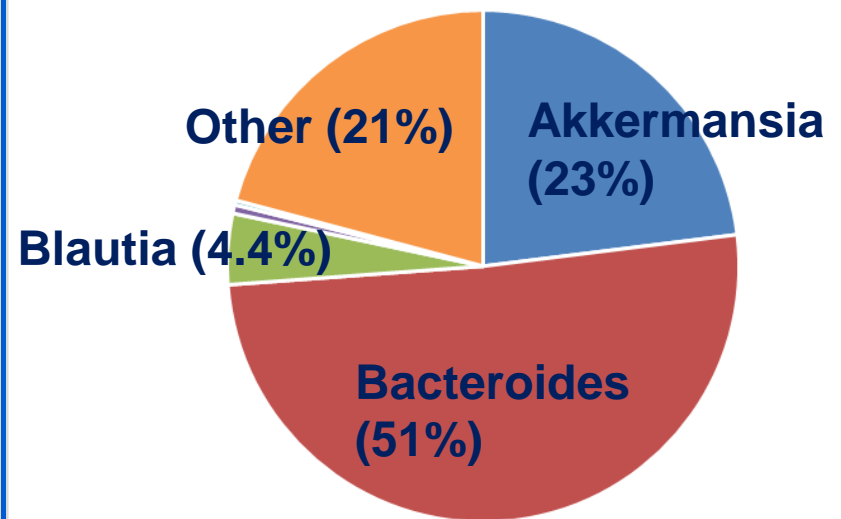


# Taxonomy: Genus level

## Humans



## Mice



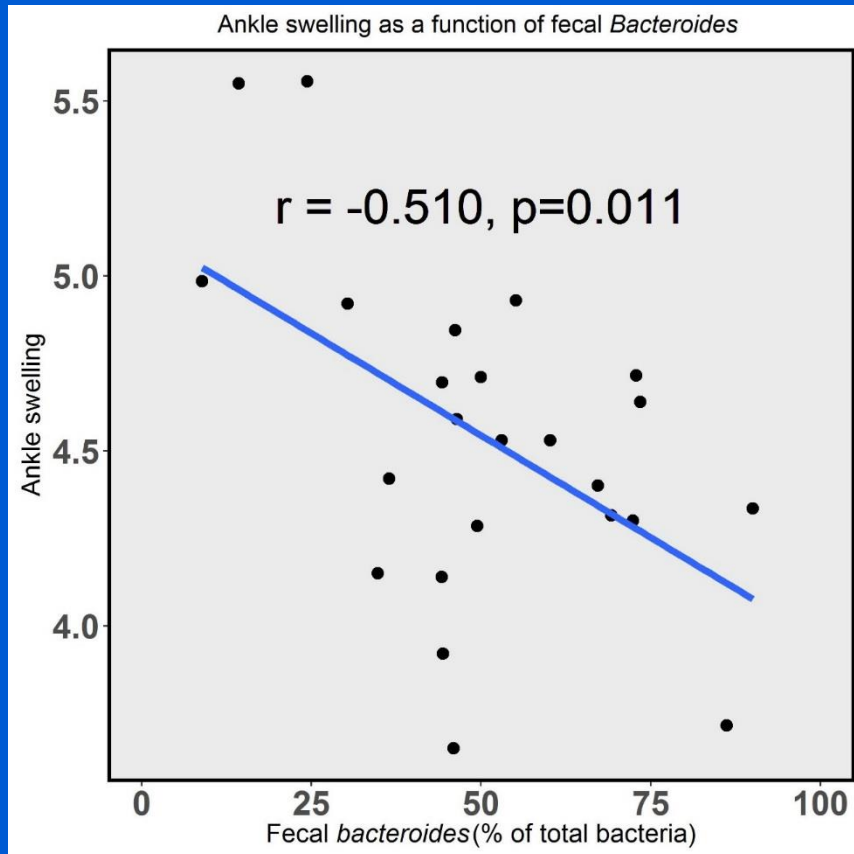
Genera present in  $\geq 5\%$  in either group

# Relationship between microbiota and arthritis, among the transplanted mice

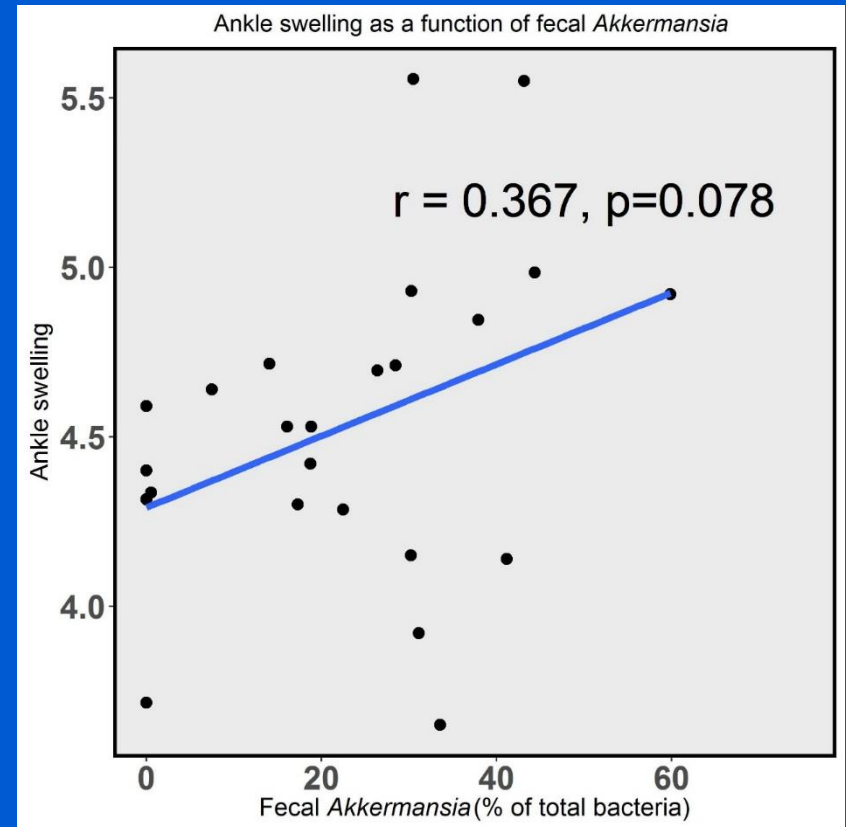
- Ran the Adonis test, which looks at whether a continuous predictor variable (in this case, ankle swelling) predicts the structure of the microbiota
- It did to some extent;  $R^2 = 0.185$ ,  $p = 0.018$
- Which bacteria were driving this?

# Association of common taxa with ankle swelling

## *Bacteroides*



## *Akkermansia*



# ***A. muciniphila* may be pro- or anti-inflammatory in arthritis**

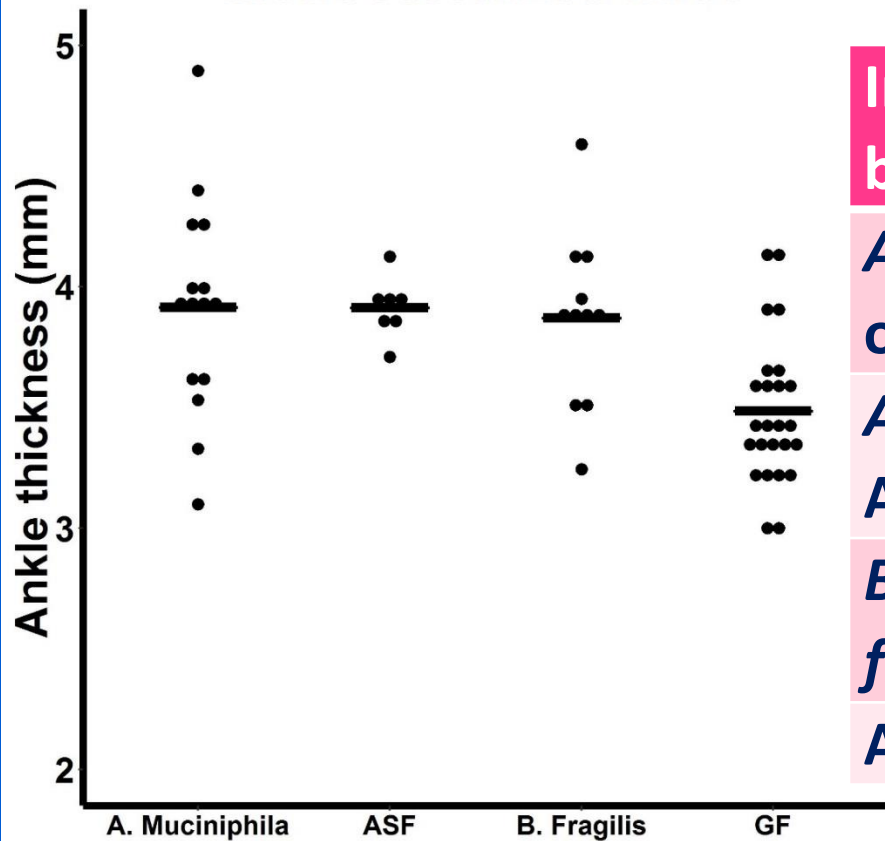
- **Given its name due to ability to degrade intestinal mucin**
  - **Potential for increased permeability**
- **Associated with disease in an animal model of spondyloarthritis<sup>1</sup>**
- **Decreased abundance in adults with psoriatic arthritis<sup>2</sup>**

<sup>1</sup>Asquith, *Arth Rheum* 2016;68:2151

<sup>2</sup>Scher, *Arth Rheum* 2015;67:128

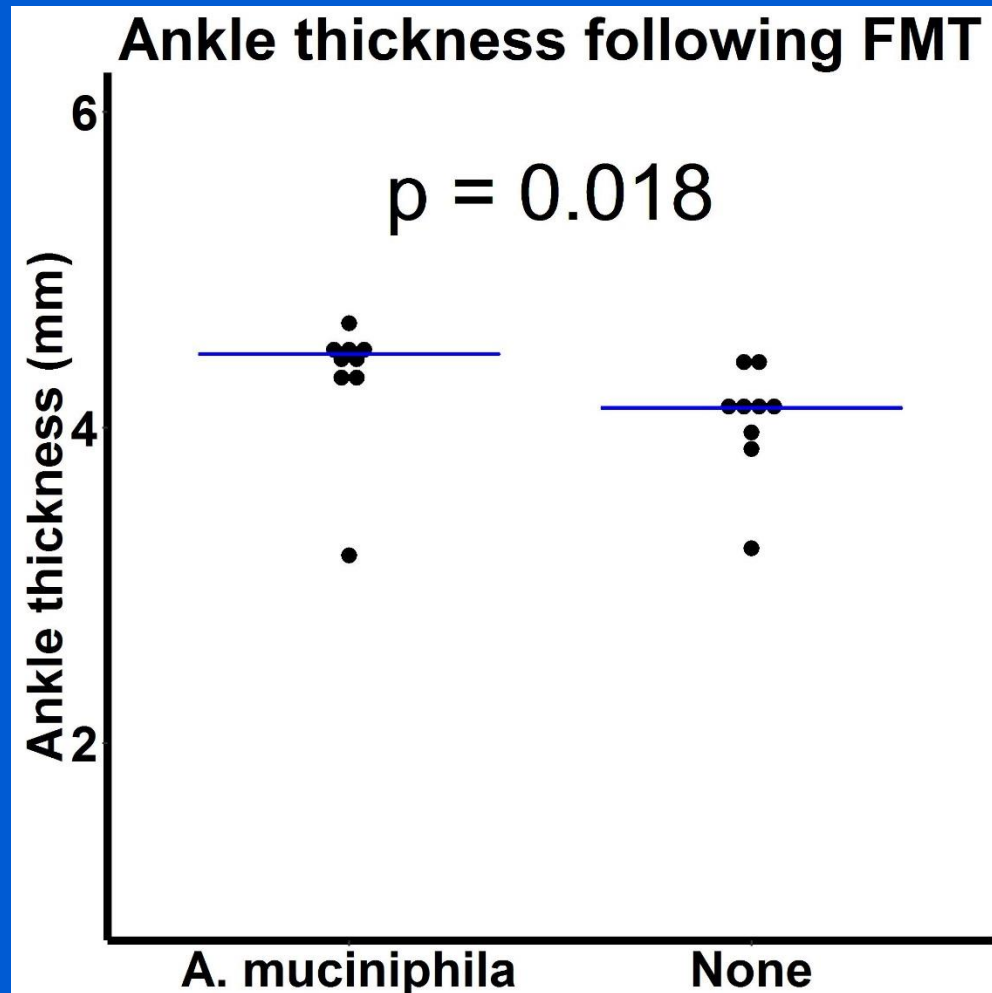
# Monocolonization with *A. muciniphila* did not affect ankle swelling

Monocolonized mice



Inoculated bacteria	Abundance of <i>Akkermansia</i> (%)
<i>A. muciniphila</i> only	88 ± 7.3
<i>A. muciniphila</i> + ASF	39 ± 14
<i>Bacteroides fragilis</i> + ASF	0.3 ± 0.5
ASF alone	0.3 ± 0.6

# Adding *A. muciniphila* to mice receiving normal microbiota did



# Summary

- *Akkermansia* may facilitate arthritis, but does not appear to be directly pathogenic
- May relate to ability of this organism to thrive on intestinal mucin
- In progress: studies to evaluate for increased invasiveness of other organisms in mice receiving *Akkermansia*

# Acknowledgements

## UAB

Randy Cron  
Wayne Duck  
Peter Eipers  
Charles Elson  
Ranjit Kumar  
Elliot Lefkowitz  
Casey Morrow  
Kathy Pierce  
Trenton Schoeb

## Other sites

Pamela Weiss  
Jennifer Weiss  
Barbara Edelheit  
Peter Nigrovic  
Charles Spencer  
Lynn Punaro  
Kenneth Schikler  
Andreas Reiff

## Funding

NIAMS  
NIEHS  
ACR  
UAB  
CoA  
CARRA



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