

“So you want to do
microbiome research...”

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Outline of Discussion

- Bacteria, archaea, fungi, or viruses???
- Sample collection, storage and processing
- 16S or shotgun
- Library construction and sequencing
- How the experiments are done?

Microbiota defined

- We are born consisting not only of our own eukaryotic human cells, but over the first few days of our life, our skin surface, oral cavity and gut are colonized by a tremendous diversity of **bacteria**, **archaea**, **fungi**, and **viruses** - a new microbial ecosystem defined as the **human microbiota**.
- The human microbiota contains almost **ten times** as many cells as are in the rest of our bodies and accounts for several pounds of body weight.

Microbiota research

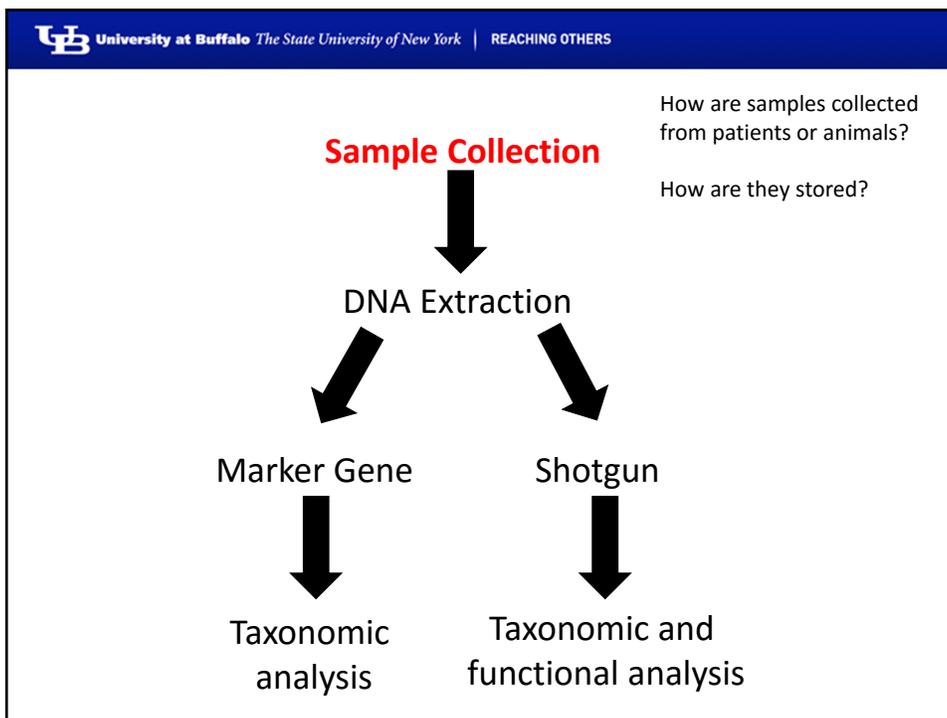
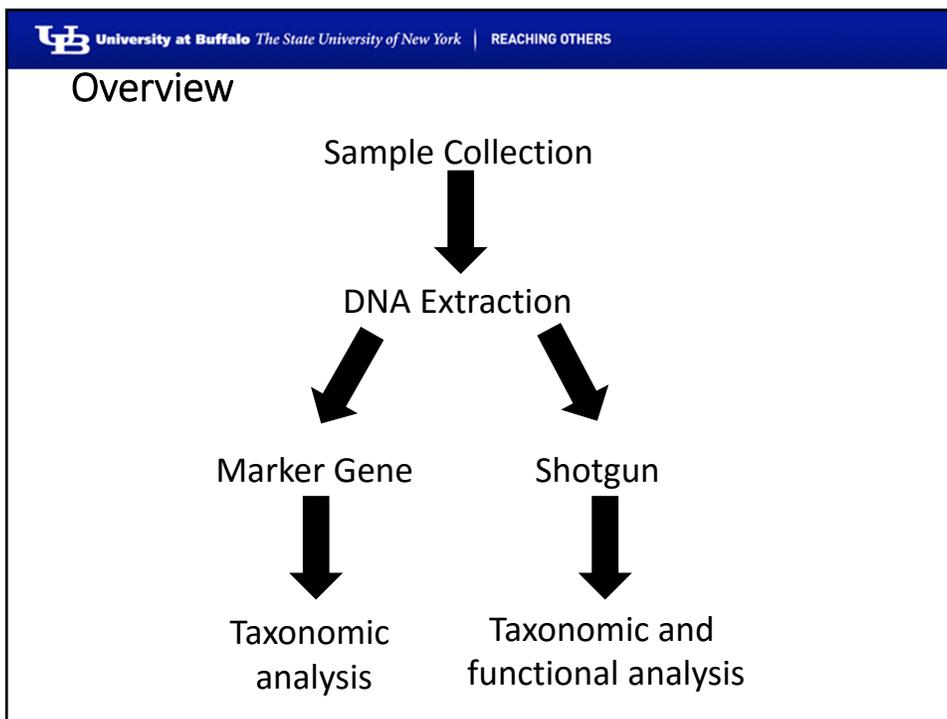
- It has long been recognized that many microbes visualized by microscopy cannot be cultivated.
- Despite advances in cultivation technology, >99% of the microbial species residing in various habitats cannot be recovered by available techniques, a phenomenon termed the '**great plate count anomaly**'.
- Currently most scientists use a PCR- and sequence-based approach that exploits 16S ribosomal DNA (rDNA) to profile bacterial diversity.

Bacteria, archaea, fungi, viruses or all of the above.

- Depending on the research question, different parts of the microbiome should be studied
- Bacteria and archaea – 16S
- Fungi and other eukaryotic – 18S
- Viruses – targeted or shotgun
- Everything - shotgun

How do we study the microbiome?

- Marker gene
 - 16S / 18S
 - Amplify region and compare
 - Cheap (\$50 per sample), biased but effective
- Shotgun
 - Extract all genomic DNA
 - Fragment, sequence and analyze
 - Expensive (\$500 per sample), information rich, should be less biased



Sample collection, storage and processing

- Bacteria like to grow, *E. coli* doubles every 20-30 minutes!
- Anaerobic versus aerobic bacteria will grow at different rates in sample collection tubes.
 - So if a person collects a sample at home and stores it in the fridge or even the freezer the population of bacteria will change over time.
- Samples need to be chemically preserved or flash frozen at -80

Sample storage



Journal of Microbiological Methods

Volume 95, Issue 3, December 2013, Pages 381–383



Note

Differential recovery of bacterial and archaeal 16S rRNA genes from ruminal digesta in response to glycerol as cryoprotectant

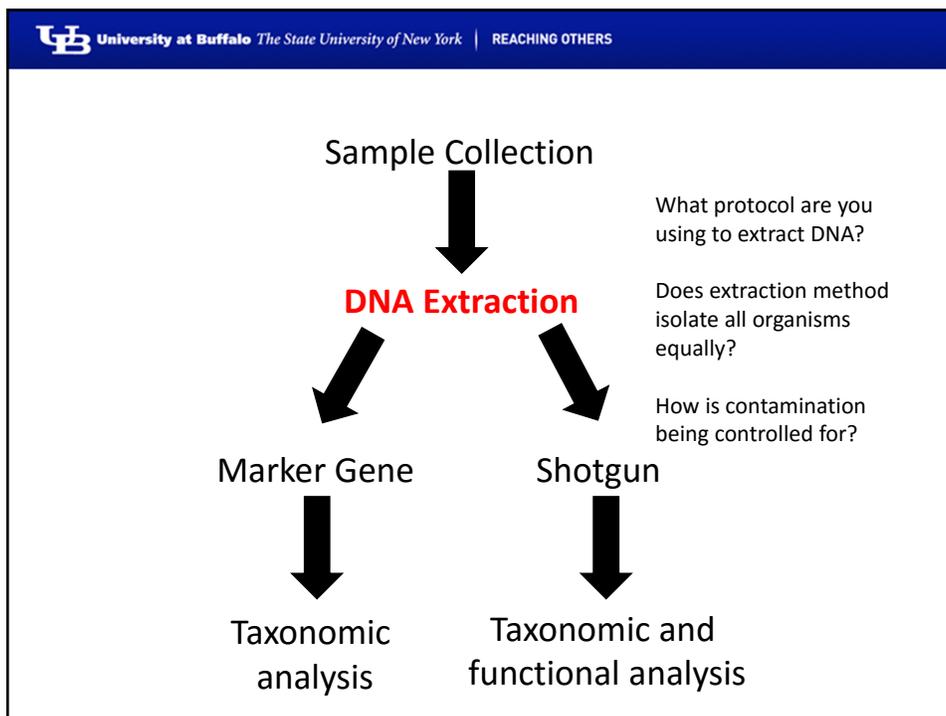
Nest McKain^a, Buğra Genç^b, Timothy J. Snelling^a, R. John Wallace^a  

[Show more](#)

doi:10.1016/j.mimet.2013.10.009

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“Samples frozen with and without glycerol as cryoprotectant indicated a major loss of *Bacteroidetes* in unprotected samples”



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DNA extraction techniques can introduce bias

Wesolowska-Andersen et al. *Microbiome* 2014, 2:19
<http://www.microbiomejournal.com/content/2/1/19>

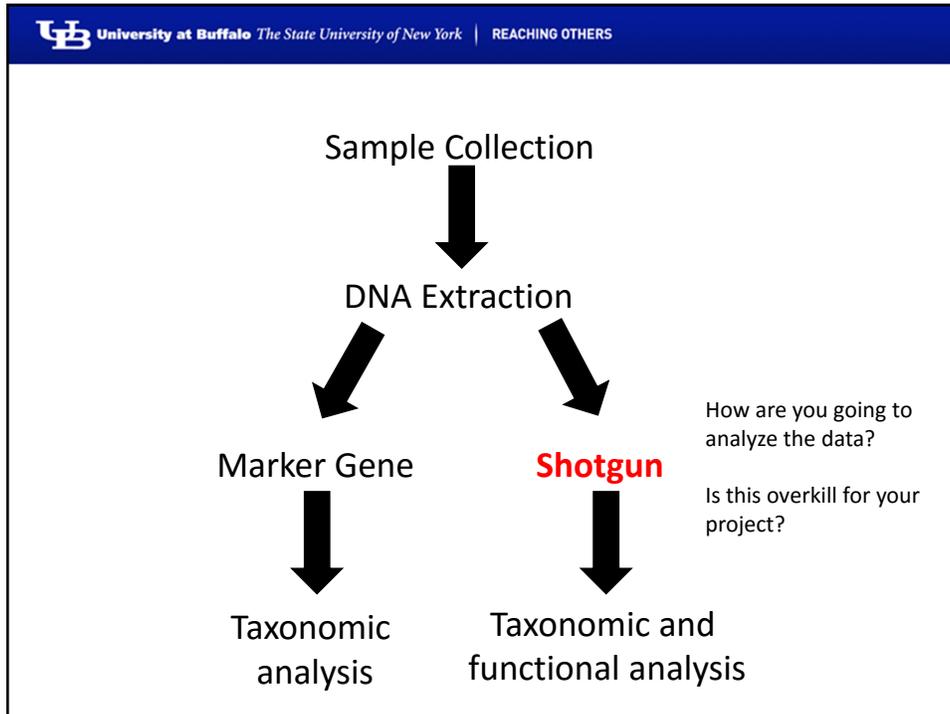


RESEARCH **Open Access**

Choice of bacterial DNA extraction method from fecal material influences community structure as evaluated by metagenomic analysis

Agata Wesolowska-Andersen¹, Martin Iain Bahl², Vera Carvalho², Karsten Kristiansen³, Thomas Sicheritz-Pontén¹, Ramneek Gupta^{1*} and Tine Rask Licht^{2*}

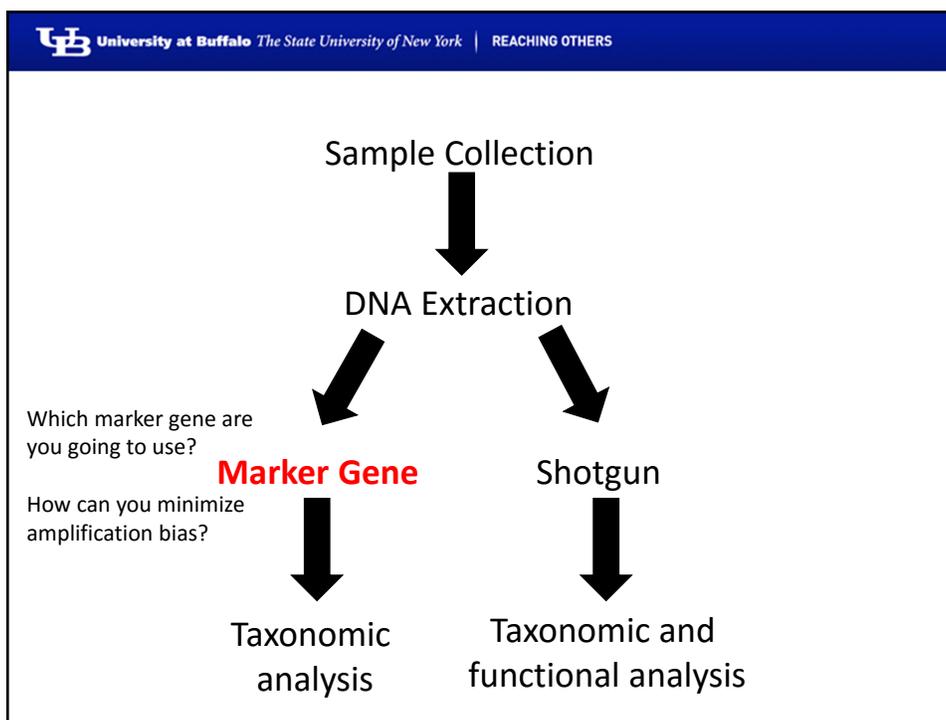
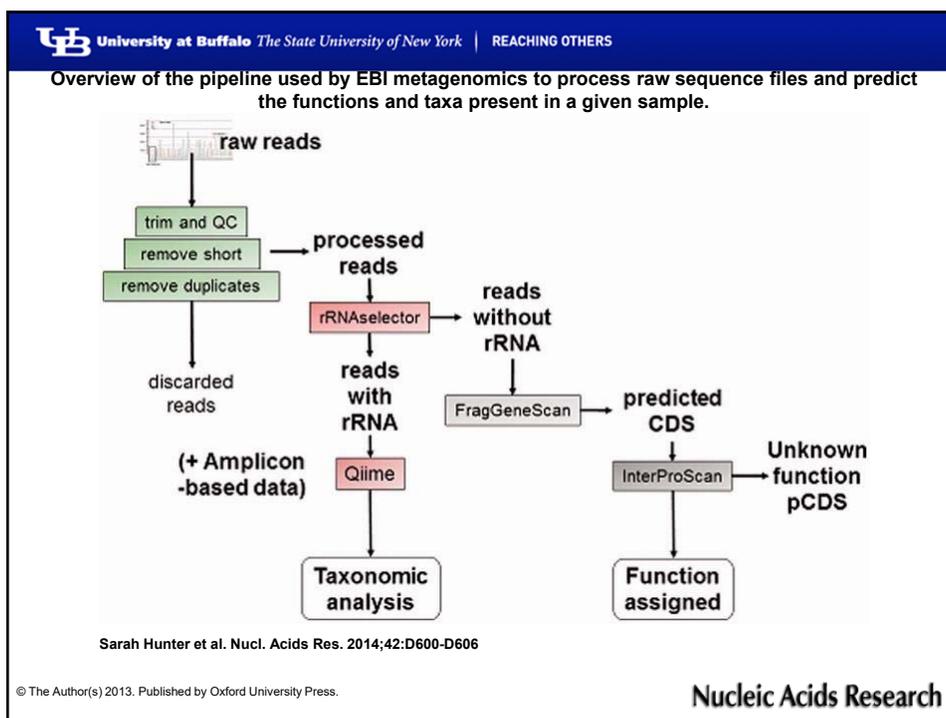
“We observed significant differences in distribution of bacterial taxa depending on the method”



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Shotgun sequencing

- Isolate all DNA from a sample and sequence it
- Should be less biased compared to “marker studies”




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16S rDNA

- 16S rDNA: is a component of the 30S small subunit of prokaryotic ribosomes.
- 16S rDNA: it satisfies the criteria of a marker by containing both highly conserved, ubiquitous sequences and regions that vary with greater or lesser frequency over evolutionary time.
- The products of the rRNA genes can fold into a complex, stable secondary structure, consisting of stems and loops. The sequences of some of the loops are conserved across nearly all bacterial species because of the essential functions involved, whereas the features of the structural parts are largely variant and specific to one or more classes.

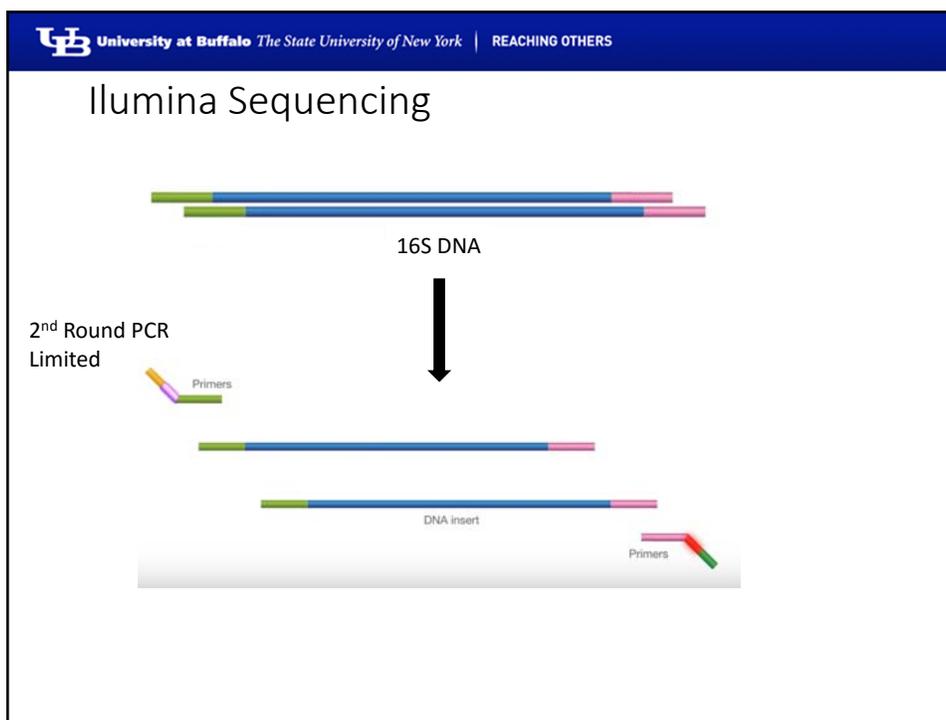
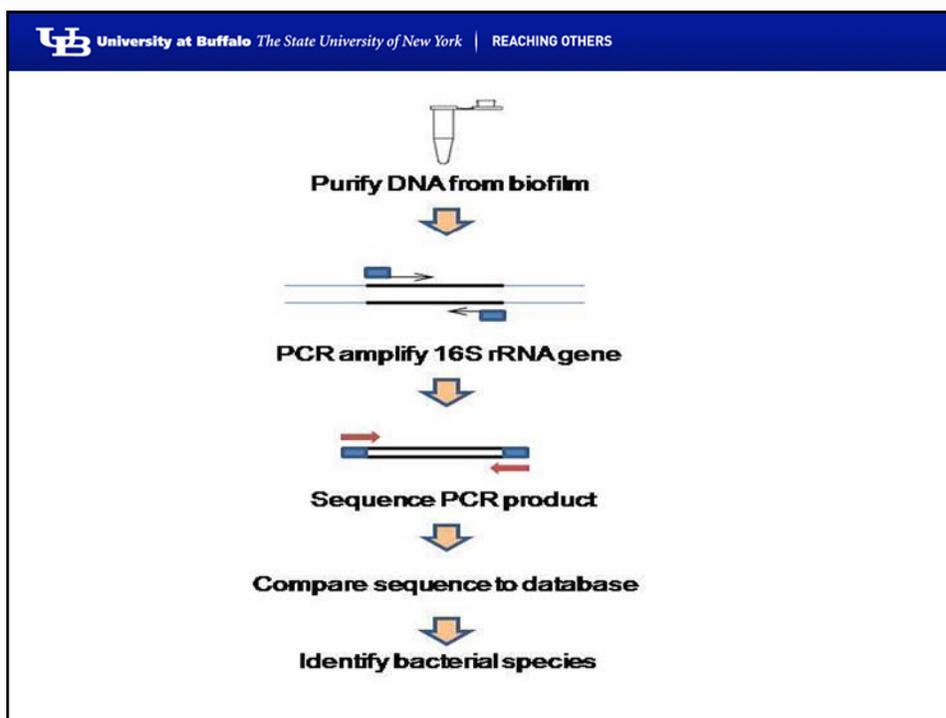

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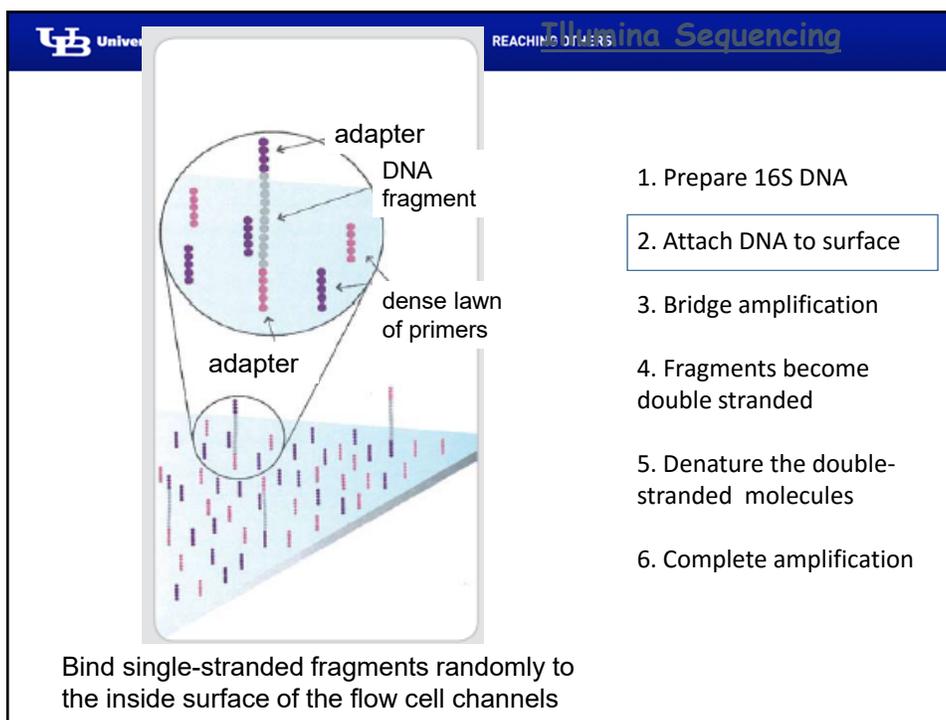
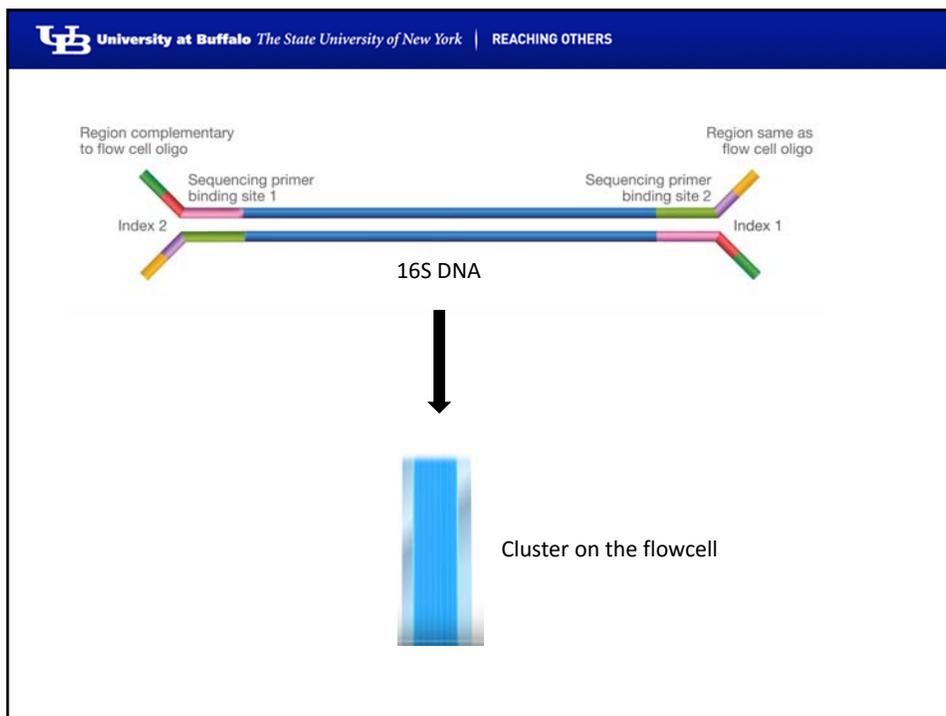
16S rDNA

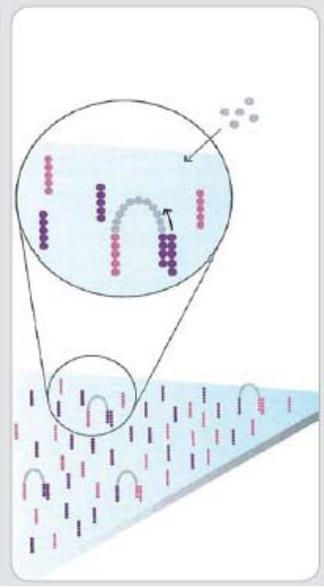
0 100 200 300 400 500 600 700 800 900 1000 1100 1200 1300 1400 1500 bp

V1	V2	V3	V4	V5	V6	V7	V8	V9
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CONSERVED REGIONS: unspecific applications
VARIABLE REGIONS: group or species-specific applications

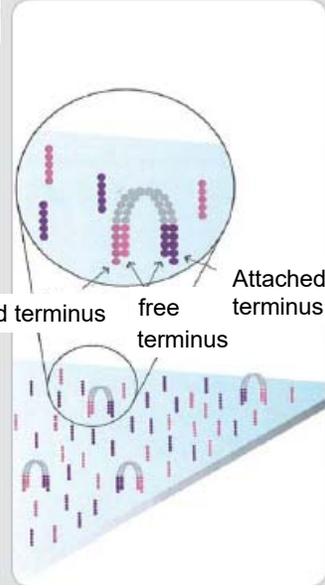






1. Prepare genomic DNA
2. Attach DNA to surface
3. Bridge amplification
4. Fragments become double stranded
5. Denature the double-stranded molecules
6. Complete amplification

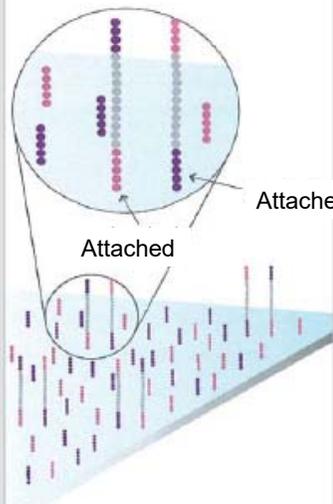
Add unlabeled nucleotides and enzyme to initiate solid-phase bridge amplification



1. Prepare genomic DNA
2. Attach DNA to surface
3. Bridge amplification
4. Fragments become double stranded
5. Denature the double-stranded molecules
6. Complete amplification

The enzyme incorporates nucleotides to build double-stranded bridges on the solid-phase substrate

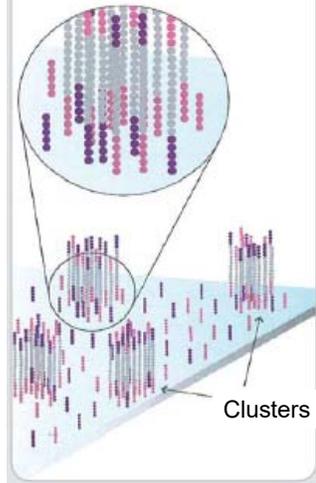
UvB Universiteit
REACHING OTHERS **Illumina Sequencing**



1. Prepare genomic DNA
2. Attach DNA to surface
3. Bridge amplification
4. Fragments become double stranded
5. Denature the double-stranded molecules
6. Complete amplification

Denaturation leaves single-stranded templates anchored to the substrate

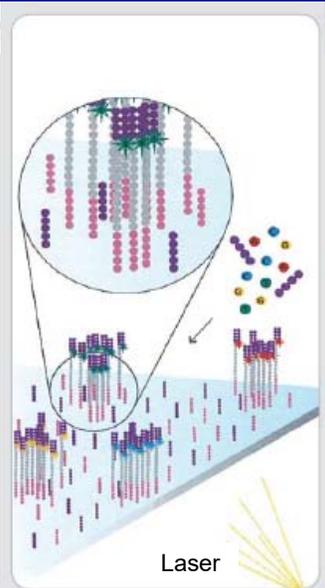
UvB Universiteit
REACHING OTHERS **Illumina Sequencing**



1. Prepare genomic DNA
2. Attach DNA to surface
3. Bridge amplification
4. Fragments become double stranded
5. Denature the double-stranded molecules
6. Complete amplification

Several million dense clusters of double-stranded DNA are generated in each channel of the flow cell

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Laser

The first sequencing cycle begins by adding four labeled reversible terminators, primers, and DNA polymerase

7. Determine first base
8. Image first base
9. Determine second base
10. Image second chemistry cycle
11. Sequencing over multiple chemistry cycles
12. Align data

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After laser excitation, the emitted fluorescence from each cluster is captured and the first base is identified

7. Determine first base
8. Image first base
9. Determine second base
10. Image second chemistry cycle
11. Sequencing over multiple chemistry cycles
12. Align data

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```

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+
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NGAGAGCTTGGCTGTACCCAGGAGTGGCATGGTCTGAGCTACTATA
    
```

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UB NSG Core

<p>Illumina HiSeq 2500</p>  <p>8 flow cell lanes with 2 flow cells 500-1000 Gb 2 billion reads per flow cell 2 x 250 bp max read length</p>	<p>MiSeq</p>  <p>1 flow cell lane 0.3 – 15 Gb 25 million reads 2 x 300 max read length - 1 billion 35-100 bp</p>
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Sample analysis

- Quality filter -- Was the sequencing good
- Paired-end sequence joining
- OTU calling
 - Reference-based
 - Non-reference based
- Reference-based analysis will change over time as databases are updated.

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OTU Table

The screenshot shows a spreadsheet with the following structure:

- Count**: A red arrow points to the column header for the numerical values.
- Bacteria Name**: A red arrow points to the column containing the taxonomic names of the bacteria.
- Sample Name**: A red arrow points to the column headers at the top of the data grid, which represent individual samples.

The data grid contains rows of OTU names and their corresponding counts across multiple samples.

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Counts need to be converted into relative frequency

$$\text{Frequency} = (\text{OTU count} / \text{reads per sample}) * 100$$

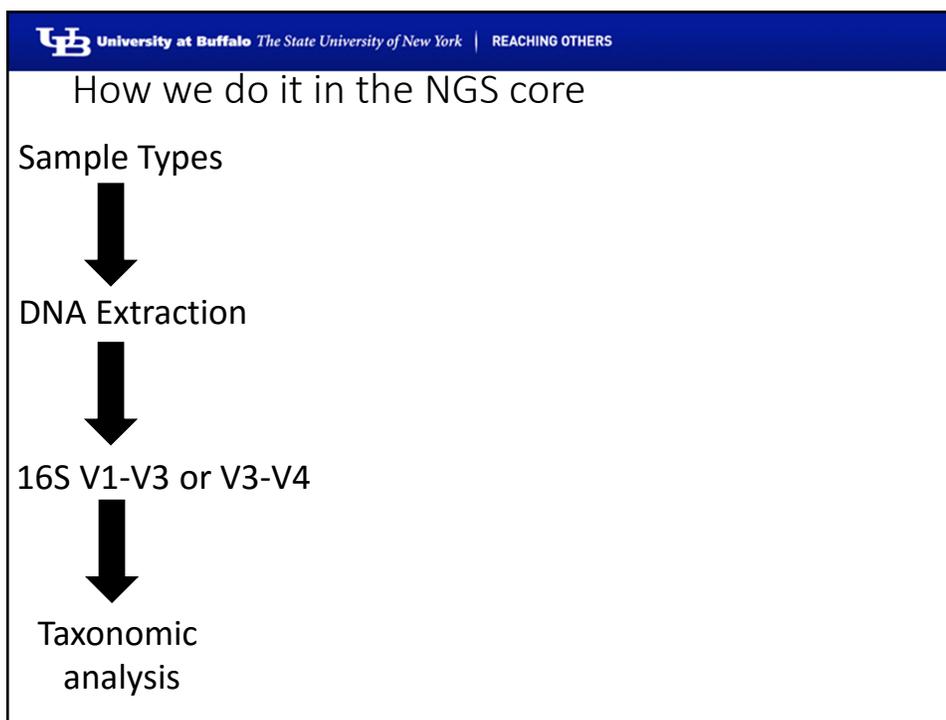
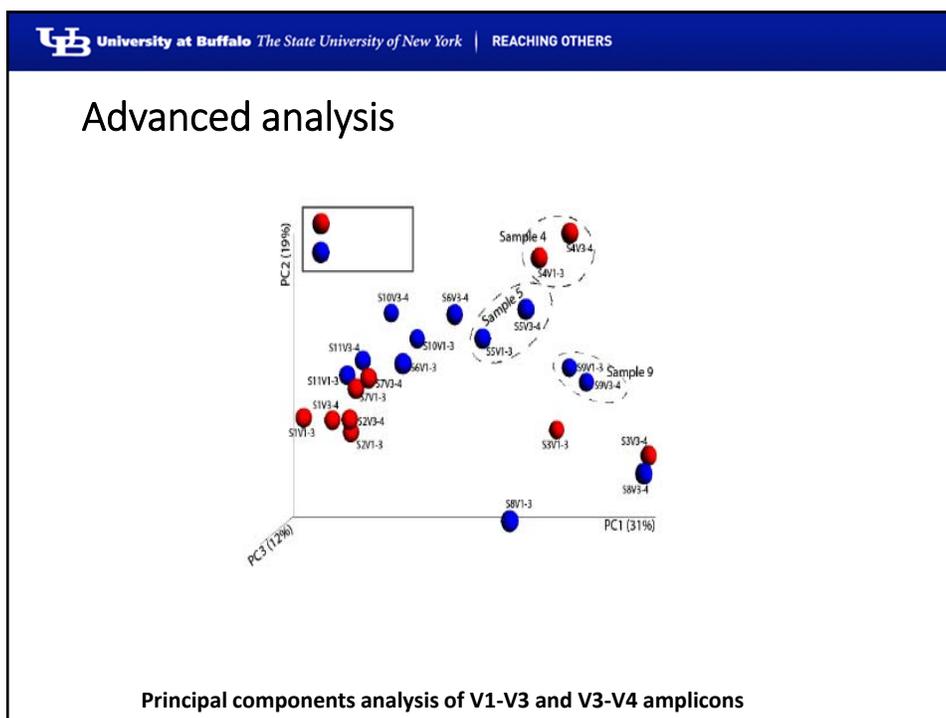
Total # of Reads per sample
➔

Number of reads per sample will vary due to sequencing
 >>>> Each sample needs to be normalized to each other <<<<<

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Statistical tests can then be performed

*Fusobacterium
 *Ireponeme
 *Ralstonia
 *Porphyromonas
 *Fallobacter
 *Streptococcus
 *Famelleia
 *Campylobacter
 *Favonimonas
 *Leptotrichia
 *Capnocytophaga
 *Filiifactor
 *Neisseria
 *Haemophilus
 *Aggregatibacter
 *Actinomyces
 *Corynebacterium
 *Eikenella
 *Burkholderia
 *Catonella
 *Pseudomonas
 *Kingella
 *Staphylococcus
 *Bulleidia
 *Lautropia
 *Mesorhizobium
 *Oribacterium
 *Sphingobacterium
 *Acinetobacter
 *Agrobacterium
 *Stenotrophomonas
 *Elizabethkingia
 *Pedobacter
 *Defflesia
 *Achromobacter
 *Salinispora
 *Curtibacter
 *Moraxella
 *Rothelia
 *Granulicatella
 *Atopobium
 *Megaspheara
 *Dialister
 *Schwartzia
 *Prevotella
 *Selenomonas
 *Veillonella
 *Prevotella
 *TTC5



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Sample handling

Sample Types

- Saliva
- Plaque
- Fecal (Human, rat, mice)
- Lavage

DNA Extraction

16S V1-V3 or V3-V4

Taxonomic analysis

Samples are processed in our BSL-2 lab



```
graph TD; A[Sample Types] --> B[DNA Extraction]; B --> C[16S V1-V3 or V3-V4]; C --> D[Taxonomic analysis];
```

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Robotic DNA Extraction

Sample Types

DNA Extraction

16S V1-V3 or V3-V4

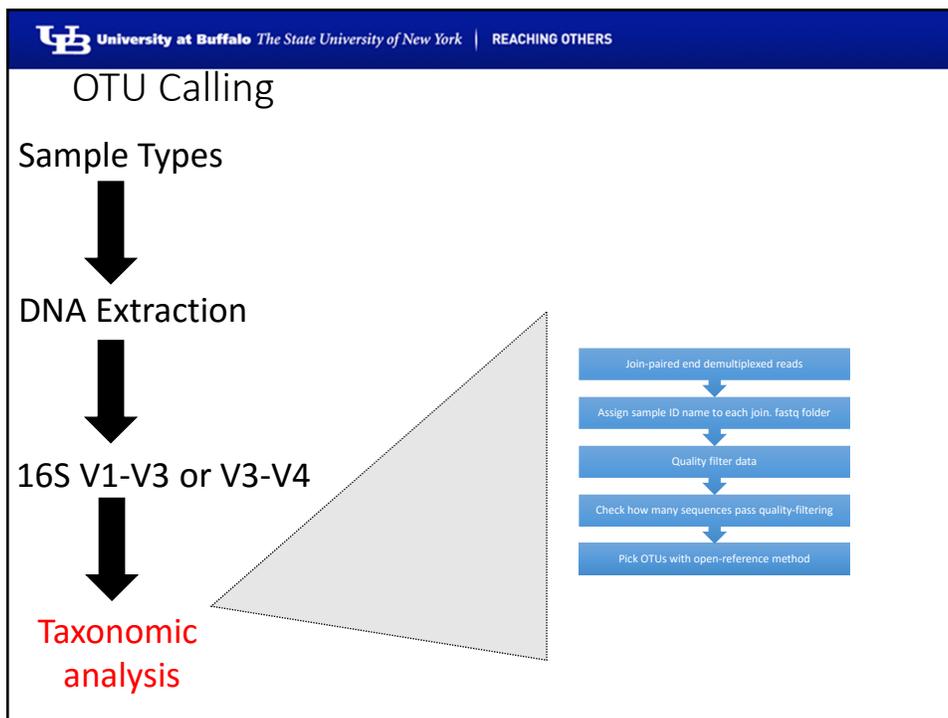
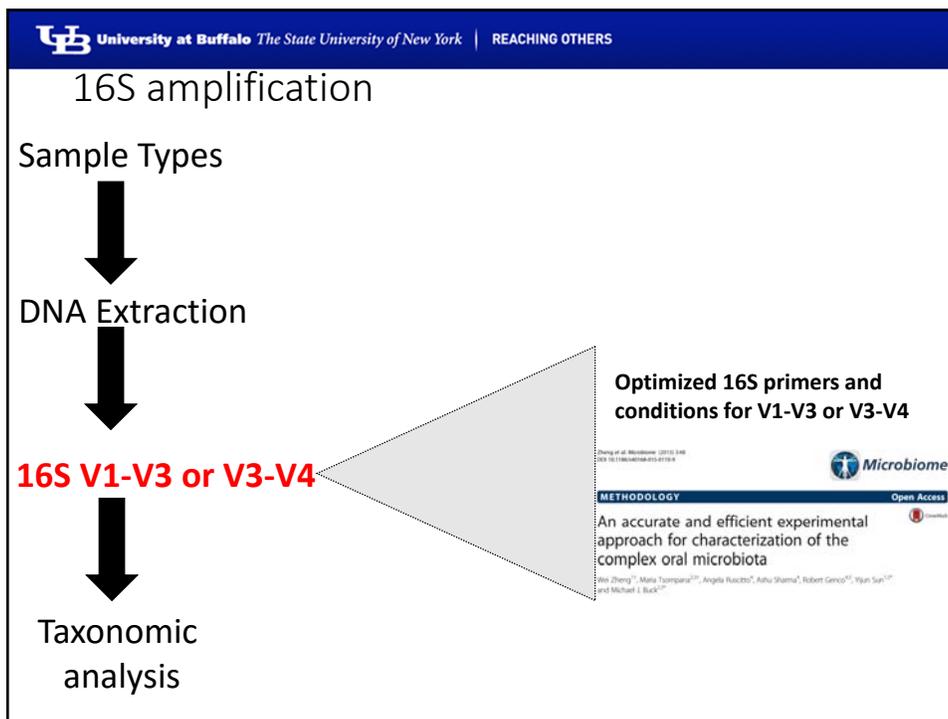
Taxonomic analysis

DNA is extracted 96 samples at time using a QIAGEN QIASYMPHONY

Samples are tracked by barcodes



```
graph TD; A[Sample Types] --> B[DNA Extraction]; B --> C[16S V1-V3 or V3-V4]; C --> D[Taxonomic analysis];
```



Limitations of 16S

- However, our ability to taxonomically characterize the microbiota using sequencing data is still restricted by the lack of universally accepted similarity thresholds, and the differential discriminatory power of the nine 16S rRNA hypervariable regions (V1-V9).
- Not all primer pairs work well for all genus/species. Amplification of non-representative genomic targets can heavily bias microbiome phylogenetic and diversity studies leading to inconclusive or inaccurate results.
- Requires PCR amplification, which can compress differences
- Does not capture viruses and eukaryotes (fungi)

Bacterial DNA is everywhere!

- Sample collection tubes, collection liquids, processing liquids will all likely have low amount of bacterial DNA.
- Even **sterile** solutions have bacterial DNA!

- Good experimental design is essential


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Salter et al. *BMC Biology* 2014, **12**:87
<http://www.biomedcentral.com/1741-7007/12/87>



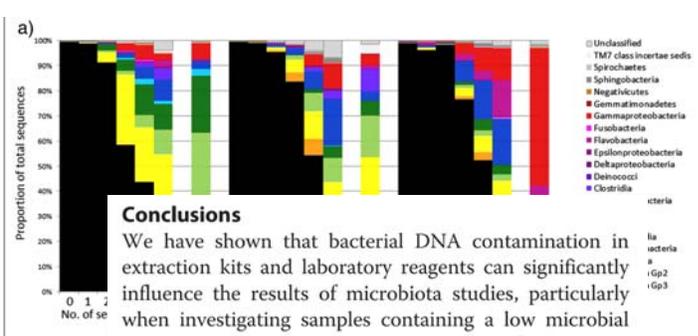
RESEARCH ARTICLE **Open Access**

Reagent and laboratory contamination can critically impact sequence-based microbiome analyses

Susannah J Salter^{1*}, Michael J Cox², Elena M Turek², Szymon T Calus³, William O Cookson², Miriam F Moffatt², Paul Turner^{4,5}, Julian Parkhill¹, Nicholas J Loman³ and Alan W Walker^{1,6*}

- Sequence a pure culture of *Salmonella bongori*
- Extracted DNA using different kits
- Did serial dilutions of the pure culture to assess impact of contaminating species


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Conclusions
 We have shown that bacterial DNA contamination in extraction kits and laboratory reagents can significantly influence the results of microbiota studies, particularly when investigating samples containing a low microbial biomass. Such contamination is a concern for both 16S rRNA gene sequencing projects, which require targeted PCR amplification and enrichment, and also for shotgun metagenomic projects which do not. Awareness of this issue by the microbiota research community is important to ensure that studies are adequately controlled and erroneous conclusions are not drawn from culture-independent investigations.

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Acknowledgments

<p>UB Genomics & Bioinformatics Facility</p> <p>Norma Nowak Maria Tsompana Sujith Valiyaparambil Natalie Waldron Jonathan Bard Brandon Marzullo</p>	<p>WHI Microbiome Team</p> <p>Jean Wactawski-Wende Robert Genco Mike Lamonte Amy Millen Chris Andrews Jo Freudenheim Yijun Sun Karen Falkner Kathy Hovey Wei Zheng Xiaodan Mai</p>
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Walker et al. *Microbiome*
DOI 10.1186/s40168-015-0087-4

 **Microbiome**

RESEARCH **Open Access**



16S rRNA gene-based profiling of the human infant gut microbiota is strongly influenced by sample processing and PCR primer choice

Alan W. Walker^{1,2}, Jennifer C. Martin¹, Paul Scott², Julian Parkhill², Harry J. Flint¹ and Karen P. Scott^{1*}

Abstract

Background: Characterisation of the bacterial composition of the gut microbiota is increasingly carried out with a view to establish the role of different bacterial species in causation or prevention of disease. It is thus essential that the methods used to determine the microbial composition are robust. Here, several widely used molecular techniques were compared to establish the optimal methods to assess the bacterial composition in faecal samples from babies, before weaning.

Results: The bacterial community profile detected in the faeces of infants is highly dependent on the methodology used. Bifidobacteria were the most abundant bacteria detected at 6 weeks in faeces from two initially breast-fed babies using fluorescent in situ hybridisation (FISH), in agreement with data from previous culture-based studies. Using the 16S rRNA gene sequencing approach, however, we found that the detection of bifidobacteria in

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He et al. *Microbiome* (2015) 3:20
DOI 10.1186/s40168-015-0081-x

 **Microbiome**

RESEARCH **Open Access**

Stability of operational taxonomic units: an important but neglected property for analyzing microbial diversity

Yan He¹, J Gregory Caporaso^{2,3}, Xiao-Tao Jiang¹, Hua-Fang Sheng¹, Susan M Huse⁴, Jai Ram Rideout³, Robert C Edgar⁵, Evgenia Kopylova⁶, William A Walters⁷, Rob Knight^{6,8} and Hong-Wei Zhou^{1*}

Abstract

Background: The operational taxonomic unit (OTU) is widely used in microbial ecology. Reproducibility in microbial ecology research depends on the reliability of OTU-based 16S ribosomal subunit RNA (rRNA) analyses.

Results: Here, we report that many hierarchical and greedy clustering methods produce unstable OTUs, with membership that depends on the number of sequences clustered. If OTUs are regenerated with additional sequences or samples, sequences originally assigned to a given OTU can be split into different OTUs. Alternatively, sequences assigned to different OTUs can be merged into a single OTU. This OTU instability affects alpha-diversity

analysis such as read-pairing, error, heterogeneity, and other such as distributed, and prediction for systems. Distinct

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