INTRODUCING THE HUMAN CONTAMINOME



International Conference on Clinical Metagenomics 16 November 2023

CHALLENGE: MICROBES ARE EVERYWHERE!

- Overview of DNA contaminants
- Example experimental design for evaluating contaminants
- Example of contaminant removal
- Key take aways



Part human, part microbes



 \rightarrow Without understanding the interactions between microbes and humans, it is impossible to obtain a complete picture of our biology

Sequencing microbial communities directly from patient samples



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What it detects	What it reveals	Advantage	Disadvantage
Bacteria, fungi, viruses, human	Taxonomy All genes	Breadth, function	\$ \$ \$ host depletion complex bioinformatics

Sequencing microbial communities directly from patient samples



What it detects	What it reveals	Advantage	Disadvantage
Bacteria only (16S rRNA gene) Fungi only (ITS or 18S rRNA gene)	Taxonomy	Low cost Increased depth No/Little host DNA	Limited view of the microbiome, Amplification bias Compositionality

Microbial load varies across body sites



Gut = $10^{11}/mL$ Vagina = $10^{8}/mL$ Urinary tract = $10^{3} - 10^{5}/mL$

<u>Need</u> to evaluate methods with microbial biomass load in mind

Adapted from Neugent et al. mBio. (2020) PMID: 32345639

duganella rhizodium paenibacillus estathrip anaerc flavobacterium kingella gr Mul sphinaobium entotrichia

- Contaminant DNA is any microbial DNA that did not originate from the sample
- Several studies have reported bacteria found in negative controls
 - Extremophiles
 - Deinococcus, Rubrobacter, ...
 - Lab-associated
 - Pseudomonas, Stenotrophomonas, ...
 - Human-associated
 - Lactobacillus, Escherichia, ...

Bacteria listed reported in: Salter et al 2014 PMID 25387460; Eisenhorf et al 2019, PMID: 30497919; Karstens et al 2019



Environmental Contaminants

- Bacteria from surfaces, materials, people when collecting and handling specimens
- Common taxa: Propionibacterium acnes, Staphylococcus...

Controls to detect:

 Empty tube / collection material at point of sample collection

Minimize by:

 Use best practices, wipe down surfaces, use gloves, laboratory coats

How to detect and remove:

• Evaluation of controls, associations of taxa with key metadata decontam other software

'Kitome' Contaminants



- Bacterial DNA from reagents and materials used in laboratory processing
- Common taxa: Deinococcus, Pseudomonas, Stenotrophomonas, Xanthomonas, ...

Controls to detect:

• Extraction blanks, Mock microbial dilution series

Minimize by:

 Use DNA-free reagents and kits designed for low microbial biomass

How to detect and remove:

• Evaluation of controls, batch effect correction, decontam, SCRuB, other software



Image modified from Austin et al. Nat Biotechnol 2023 Mar 16 PMID: 36928429

Well-to-well Contaminants

- Bacteria from other samples in adjacent samples due to sampling / sample handling / spillover
- Common taxa: Any in the dataset

Controls to detect:

Mock community / Positive control

Minimize by:

- Process similar biomass specimens together
- Use a block design on 96-well plates

How to detect and remove:

• Individual taxa visualization, SCRuB

Index hopping

- Sequencing artifact
- Common taxa: Any in the dataset



Controls to detect:

Mock community / Positive control

Minimize by:

- Use unique dual indexing pooling combos
- Remove free adapters from library preps

Image from: www.illumina.com/techniques/ sequencing/ngs-library-prep/multiplexing/ index-hopping.html

How to detect and remove:

Abundance filter, Barcode error correction

Tools for evaluating contaminants

Amplicon

Decontam

Davis N et. al. *Microbiome* 226 (2018) PMID: 30558668

microDecon

McKnight, DT, et al Environmental DNA. 2019; 1: 14–25

 SCRuB: Source-tracking for Contamination Removal in *micro* Biomes

Austin et al. Nat Biotechnol 2023 Mar 16 PMID: 3692842

ConQuR

Ling W et al. Nature Comm 2022 Sep 15;13(1):5418

PMID: 36109499

Shotgun Metagenomics

Squeegee
 Lui et al Nature Comm 2022 Nov 10;13(1)
 PMID: 36357382

Visualization

GRIMER –

Piro & Renard, GigaScience, Volume 12, 2023, giad017 https://pirovc.github.io/grimer/

Experimental design to detect contaminants



Undiluted mock

Microbial community



Benchmarking study:

- Mock microbial dilution series
- Extracted DNA and sequenced along with biological samples
- Assess accuracy
- Guide bioinformatics



Benchmarking study:

- Unexpected sequences increase as microbial biomass decreases
- Has a major impact on diversity measures and bacterial abundances





Karstens et al **mSystems** 4(4): e00290-19 (2019))



No method is perfect.

Best approaches require additional information

- Decontam (frequency) requires DNA measured for each sample
- Decontam (prevalence) requires at least 5 negative controls
- SourceTracker requires a well-defined communities

Likely need more than one approach is needed

No current available method accounts for all types of contaminants

Parameterization is experiment dependent

- Need controls and exploratory data analysis to evaluate
- Need to prioritize goals of the study

Take home points

Many methods exist and are being developed to handle contaminants



- Appropriately applying these methods requires:
 - Good experimental design (Positive AND Negative controls)
 - Thorough exploratory data analysis
 - A team effort (reviewing/discussing results)
 - Patience

Take home points



- Contaminants can be problematic for sequenced-based microbiome studies.
 Especially when:
 - Sample type has a wide range of possible biomass
 - Biomass is linked to the outcome of interest

Interpretation is key

- Association with contamination indicates a change in microbial biomass / change in microbiome structure
- Presence of DNA is NOT always evidence of a microbiome BUT can still be meaningful

Steps to improve rigor of sequence-based microbiome studies



Transparency when publishing findings is key

- Critical assessment is needed by both authors and reviewers
- Limitations should be clearly stated

Use community developed checklists:

- STORMS (Mirzayi et al 2021, PMID 34789871; www.stormsmicrobiome.org)
- RIDE (Eisenhorf et al 2019, PMID: 30497919)

• Share data from negative and positive controls

- Enable re-use by method developers
- Enable large data libraries for AI/ML



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