Meta-analysis of positive controls and laboratory metainformation in microbiome data

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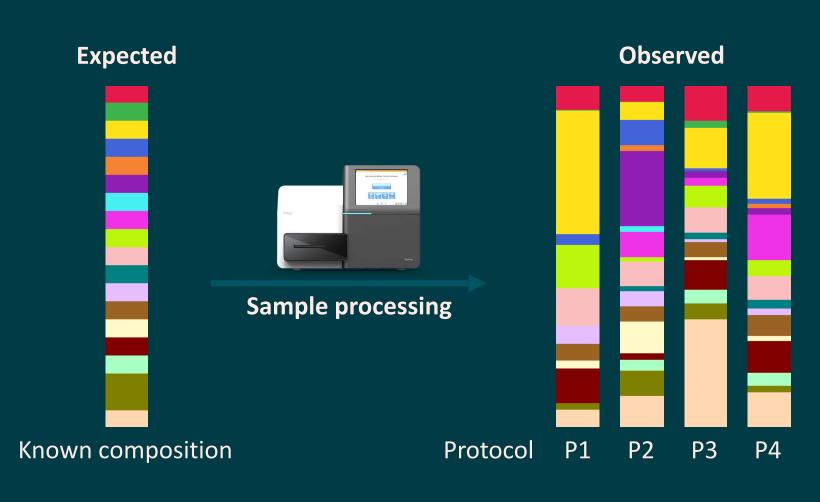
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Introduction

Human microbiome research has revolutionized our understanding of the microbiome's contribution to human health, and diseases such as obesity, Inflammatory Bowel Disease, or Atopic Dermatitis.

The huge variety of available methods for generating microbiome data leads to distinct errors and biases depending on the chosen laboratory method, which limits the comparability and clinical application of microbiome data.

These protocol-specific errors and biases can be quantified by mock samples, i.e. positive controls with known species composition that are processed along with biological samples.



Aim

We aim to build a database of published microbiome studies that used standardized, commercially available mock communities as positive controls. We then collect the studies' laboratory metadata to quantify the impact of different laboratory methods on microbiome data.

References

Study ID 1: Martí et al. Cell Rep Med. 2021. doi: 10.1016/j.xcrm.2021.100206. Study ID 2: Glendinning et al. Poult Sci. 2022. doi: 10.1016/j.psj.2021.101624. Study ID 3: Pollock et al. Anim Microbiome. 2021. doi: 10.1186/s42523-021-00144-x.

Study ID 4: Glendinning et al. Anim Microbiome. 2019. doi: 10.1186/s42523-019-0017-z

Study ID 5: Porcellato et al. Sci Rep. 2020. doi: 10.1038/s41598-020-77054-6. Study ID 6: Dumont-Leblond et al. Commun Biol. 2021. doi: 10.1038/s42003-021-01690-5.

Study ID 7: Martí et al. STAR Protoc. 2021. doi: 10.1016/j.xpro.2021.100652.

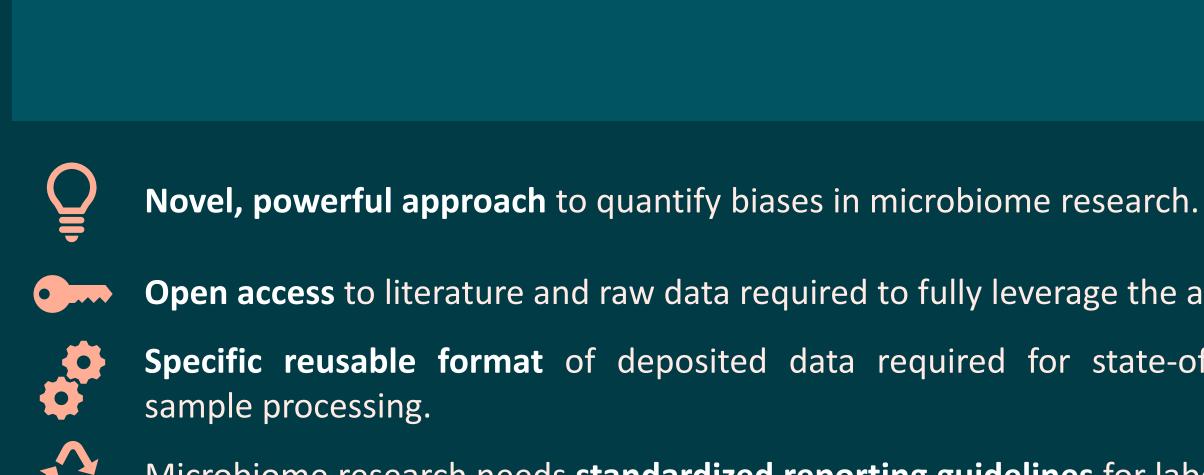
Comp

ATCC

BEI Reso

Zymo BIO

- leading to only n=7 studies included for collection of laboratory metadata.



Search strategy & exclusion criteria

pany	Mock name	[mock name]	"[mock name]"	[mock name] [company]	"[mock name]" [company]
	MSA-1000	116	116	17	17
	MSA-1001	21	21	14	14
•	MSA-1002	53	53	49	49
•	MSA-1003	37	37	32	32
	MSA-2002*	482	482	32	32
	MSA-2003	384	384	33	33
	HM-280	176	176	9	9
	HM-281	162	162	10	10
urces	HM-782D	140	140	137	137
	HM-783D	75	75	74	74
	D6300	10,400	741	134	134
	D6305	231	124	77	77
)-	D6310	192	36	7	7
AICS	D6311	200	74	9	9
	D6322	103	34	5	5
	D6331	130	38	13	13

After identifying companies that provide commercially available mock communities, we performed a systematic literature search in Google Scholar to find scientific papers that use these mock communities. More specific results were found when the company's name was added to the search, instead of the name of the mock alone.

The n=32 articles mentioning MSA-2002 by ATCC (highlighted by asterisk) were then screened. Studies were excluded for the following reasons: • Duplicate (n=1),

- Full text not available (n=3),
- Content not relevant (n=13),
- Sequencing technology out of scope (n=4),
- Raw sequencing data not deposited (n=4),

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The n=7 studies using MSA-2002 provided on average 13 (median) out of 23 required pieces of laboratory metadata, indicated by dots in the table. Study ID 7 specified the most pieces of laboratory metadata, but this study is a best-practice protocol for improved sample processing. The laboratory metainformation is particularly missing in beginning of sample processing (upper quarter of the table). Underlined metadata were further investigated in the bias evaluation.

Open access to literature and raw data required to fully leverage the approach's power.

Specific reusable format of deposited data required for state-of-the-art bioinformatic

Microbiome research needs standardized reporting guidelines for laboratory methods.



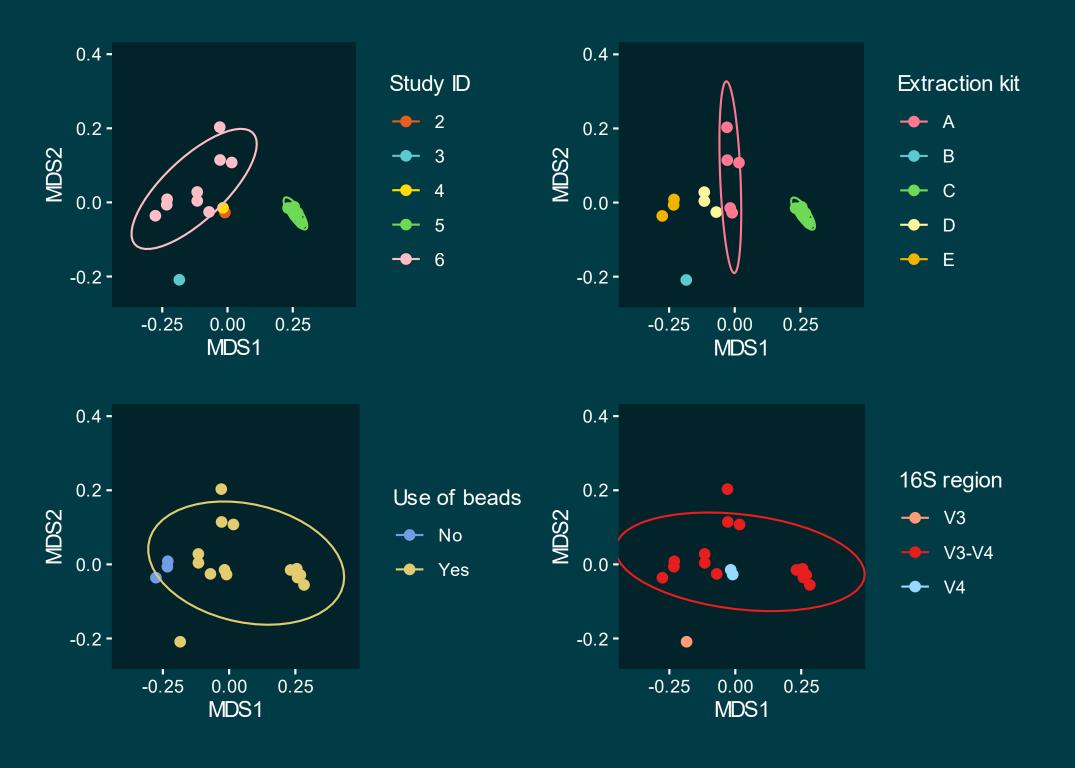
Pilot study

Metadata collection

Study ID	1	2	3	4	5	6	7
mber							
nput cells / dilution							
ant							
fer							•
perature							
<u>it</u>							•
<u>s</u>							•
nd material							•
ions							•
fication							•
							•
ience							•
PCR cycles							•
atures							•
							•
t quantification							•
ample pooling							•
platform							•
n							•
kit							•
I sequencing length							•
software							
23)	13	13	9	13	15	9	17

After further excluding studies with ID 1 and 7, which had their raw sequencing data deposited in a non-reusable format, n=5 studies with a total of 17 samples were bioinformatically processed. The microbiome sequencing results of the mock community MSA-2002, as generated by the different laboratory methods chosen per study, show substantial variation between individual studies (A), as indicated by nonmetric multidimensional scaling of Bray-Curtis dissimilarities between samples.

Among the chosen laboratory metadata for further bias evaluation, the choice of extraction kit seems to lead to the largest variation in results (B), compared to the use of beads (C) or 16S region (D).



Conclusions and next steps

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Can the search strategy (search terms or search engine) be improved? Which laboratory metadata are the most important, and need to be available for re-analysis in our database? Inconsistent description and scattered distribution of metadata across the papers' methods sections require a **paper scraping algorithm**?

Expansion of meta-analysis approach to other mock communities.

Bias evaluation