

Response to Invited Commentary

Waldron et al. Reply to “Commentary on the *HMP16S*Data Bioconductor Package”

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We thank Griffith and Morgan (1) for their excellent summary of the Human Microbiome Project phase I data set and of our efforts to remove technical hurdles to its use by epidemiologists. Their commentary provides a clear overview of our *HMP16S*Data (2) Bioconductor (3) package and the necessary precautions for users of these data. In this reply, we expand more generally on the need to lower barriers to reuse of public-access genomic datasets.

The importance of public availability of published data is already broadly accepted across disciplines from perspectives of reproducibility, transparency, and further scientific discovery. Open resistance to data sharing and reuse policies (e.g., to “research parasites” (4)) has been overwhelmed, and the prevalence of data sharing has expanded due to journal policies (such as that of the *Journal*, which adopts recommendations of the International Committee of Medical Journal Editors (5)), funding policies (such as the National Institutes of Health genomic data sharing policy (6) and the European Commission Open Research Data Pilot (7)), and recognition of its importance by authors and peer reviewers. The benefit of data sharing, however, comes “not from providing access to data or depositing them somewhere, but from making it possible for others to find and reanalyze the data in a meaningful way.” (8, p. 2409) Toward this objective, however, there is less consensus about how to move forward.

Our work and the commentary by Griffith and Morgan highlight technical barriers to utilizing the HMP 16S rRNA gene sequencing data set, but such barriers are by no means limited to this data set. Decentralized researcher-driven studies provide a majority of publicly available genomic data and present additional challenges of standardization and completeness. For example, the National Center for Biotechnology Information (NCBI) Sequence Read Archive (SRA) database enforces provision only of a minimal set of mandatory metadata that are relevant across all areas of genomic investigation (such as library and instrument information, and species) (9), whereas the participant metadata of critical interest in epidemiology are provided with no requirements for inclusion or vocabulary. Key attributes such as age, sex, and disease status may be missing, and when present they

must be cleaned and standardized. Our related *curatedMeta-genomicData* project (10) developed a system for manual standardization and automatic syntax-checking of participant metadata when made possible by the voluntary provision of key metadata by the researchers who upload data. The adoption of more specific standards for how metadata from health studies are shared would make such manual standardization unnecessary, but significant practical work and consensus-building remains. Groups like the Society for Epidemiological Research may be able to play a leadership role in establishing such community standards.

The growth of multiomic data sets, where multiple types of molecular data are collected on the same specimens, raises additional bioinformatic hurdles to reanalysis. Such data sets may require multiple data-processing pipelines and complex data linkage. The “Integrative Human Microbiome Project” (iHMP) (11) is providing longitudinal measurements of metagenomics, metatranscriptomics, metabolomics, metaproteomics, and other data, presenting an even greater data-integration challenge than the current project. Such complex data sets can leave error-prone and nongeneralizable sets of tasks to perform for every analysis, exposing limitations in traditional approaches to data management. We and others are working to use recent software for multiomic data integration in Bioconductor (12) to provide a similar level of usability for the iHMP data.

In summary, the sharing of research data is key to allowing reproducibility of existing studies and to maximizing research investments in public health. However, the details of that sharing and ongoing community efforts towards standardization will determine the extent to which hard-earned and expensive research data are used to their full potential for public good.

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