ASaiM-MT: a validated and optimized ASaiM workflow for metatranscriptomics analysis within Galaxy framework

[version 2; peer review: 2 approved]

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Abstract
The Earth Microbiome Project (EMP) aided in understanding the role of microbial communities and the influence of collective genetic material (the ‘microbiome’) and microbial diversity patterns across the habitats of our planet. With the evolution of new sequencing technologies, researchers can now investigate the microbiome and map its influence on the environment and human health. Advances in bioinformatics methods for next-generation sequencing (NGS) data analysis have helped researchers to gain an in-depth knowledge about the taxonomic and genetic composition of microbial communities. Metagenomic-based methods have been the most commonly used approaches for microbiome analysis; however, it primarily extracts information about taxonomic composition and genetic potential of the microbiome under study, lacking quantification of the gene products (RNA and proteins). On the other hand, metatranscriptomics, the study of a microbial community’s RNA expression, can reveal the dynamic gene expression of individual microbial populations and the community as a whole, ultimately providing information about the active pathways in the microbiome. In order to address the analysis of NGS data, the ASaiM analysis framework was previously developed and made available via the Galaxy platform. Although developed for both metagenomics and metatranscriptomics, the original publication demonstrated the use of ASaiM only for metagenomics, while thorough testing for metatranscriptomics data was lacking. In the current study, we have focused on validating and optimizing the tools within ASaiM for
metatranscriptomics data. As a result, we deliver a robust workflow that will enable researchers to understand dynamic functional response of the microbiome in a wide variety of metatranscriptomics studies. This improved and optimized ASaiM-metatranscriptomics (ASaiM-MT) workflow is publicly available via the ASaiM framework, documented and supported with training material so that users can interrogate and characterize metatranscriptomic data, as part of larger meta-omic studies of microbiomes.

**Keywords**
Galaxy, metatranscriptomics, microbiome, functional analysis

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This article is included in the Galaxy gateway.

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REVISED Amendments from Version 1

1. In the updated version, we have added Dr. Phil B. Pope as a co-author. He is the principal investigator of the project that generated the metatranscriptomics datasets that we used for benchmarking.

2. The Human Microbiome Project and the role of microbiome was replaced with the Earth Microbiome Project according to the reviewer suggestion.

3. We have added references to the software tools mentioned in the Introduction along with italicising the software tools to distinguish them.

4. We have updated Figure 2 by fixing typos.

5. We reformat ted the Method section into sub-categories as suggested by the reviewer.

6. An explanation was provided for the usage of FASTQ interlacer in the ASaiM-MT workflow compared to the original ASaiM, batch processing of data by using dataset collection and also specified the advantages of the tools incorporated in ASaiM-MT workflow.

Any further responses from the reviewers can be found at the end of the article.

Introduction

Understanding the role of microbiome diverse ecosystems such as fresh water lakes, permafrost soils from Alaskan forests, and deep-sea oil plumes due to oil spills has opened up various avenues of research. In clinical research, the role of microbiomes has been studied in patho-physiological conditions such as inflammatory diseases, obesity, and cancer. Experimental design and biological interpretation of microbiome data has become an area of intense focus as the contributions to human health and disease are becoming clearer. The ‘meta-omics’ approaches, such as metagenomics, metatranscriptomics and metaproteomics have been developed to study microbiomes without culturing and target the major macromolecules that constitute the community, namely DNA, RNA and proteins. While metagenomics (16S rRNA or whole genome sequencing) focuses on the taxonomy profile and functional potential, metatranscriptomics, metaproteomics and meta-metabolomics uncover the functional response of the microbiome to stimuli on the short and long time-scale, respectively.

Metatranscriptomics has been used to analyze microbial gene expression profiles from a variety of complex sample types, e.g. human microbiome, aquatic or terrestrial environments, plant-microbe interactions. Despite these applications, challenges still exist in the analysis of the complex metatranscriptomics data. Metatranscriptomics data is usually generated using Illumina sequencing technology and high-throughput sequencing of short RNA-Seq reads using experimental design and biological interpretation of microbiome data has become an area of intense focus as the contributions to human health and disease are becoming clearer. The ‘meta-omics’ approaches, such as metagenomics, metatranscriptomics and metaproteomics have been developed to study microbiomes without culturing and target the major macromolecules that constitute the community, namely DNA, RNA and proteins. While metagenomics (16S rRNA or whole genome sequencing) focuses on the taxonomy profile and functional potential, metatranscriptomics, metaproteomics and meta-metabolomics uncover the functional response of the microbiome to stimuli on the short and long time-scale, respectively.

Metatranscriptomics has been used to analyze microbial gene expression profiles from a variety of complex sample types, e.g. human microbiome, aquatic or terrestrial environments, plant-microbe interactions. Despite these applications, challenges still exist in the analysis of the complex metatranscriptomics data. Metatranscriptomics data is usually generated using high-throughput sequencing of short RNA-Seq reads using Illumina sequencing technology. Many software tools and workflows are available for metatranscriptomics analysis. These include tools for RNA-Seq Data Preprocessing: Quality Control (FastQC), Ribosomal RNA removal (SortMeRNA, barrnap), host RNA removal (BMTagger), De Novo Assembly (Trinity, MetaVelvet, Oases, IDBA-MT), Transcript Taxonomy (Kraken, GOTCHA, MetaPhlAn2), FunctionalAnnotation (HUMAnN2), Annotation of assembled contigs are subjected to gene finding programs such as FragGeneScan followed by functional assignment using DIAMOND searches against KEGG, NCBI RefSeq, UniProt. Differential Expression analysis is performed by tools such as EdgeR, DESeq2 and limma. "Reads-Based" analysis is performed by tools such as MetaTrans, COMAN, FMAP, SAMSA, ASaiM and Assembly Based analysis: SqueezeMeta, IMP, MOSCA. Some of these open source tools have been incorporated within the Galaxy bioinformatics workbench to make it more accessible to users on a single platform.

ASaiM framework was previously developed by Batut et al. to perform metagenomics and metatranscriptomics data analysis. The major goal of ASaiM was to develop an accessible, resharable, and user-friendly framework for microbiome researchers, implemented within the Galaxy platform. The framework integrates a comprehensive set of microbiota related tools, predefined and tested workflows as well as supporting training material and documentation. It is available for users as a Docker image but also as a web server (https://metagenomics.usegalaxy.eu). This implementation also enables flexibility, so that the workflow can be customized for datasets of diverse origins and methods emerge. To address the need for optimizing ASaiM for metatranscriptomics data, we added the ASaiM-metatranscriptomics (ASaiM-MT) (Figure 1), a metatranscriptomics workflow, and rigorously tested it to ensure reliable analysis of metatranscriptomics data. Our testing and validation focused on using contemporary tools in their most current version (Table 1), capable of handling large datasets, and ensuring that the outputs from each of the tools were compatible in order to build an integrated and automated workflow. The workflow also has potential for integration with other meta-omic tools and workflows in Galaxy, such as those designed for metaproteomics, to enable multi-omic data analysis.

The ASaiM-MT workflow is available via the ASaiM framework, specially at https://metagenomics.usegalaxy.eu/, for users to test their metatranscriptomics data. It is supported by a step-by-step tutorial, available on the Galaxy Training Network (GTN), which provides explanation for the different steps and the opportunity for online, hands-on training in using the workflow, with a trimmed dataset. The Galaxy training network also provides online support through the Gitter channel (https://gitter.im/Galaxy-Training-Network/Lobby), where the users can interact with the developers.

Methods

Workflow implementation

The ASaiM-MT workflow contains all the processing steps and the parameters required for the metatranscriptomics analysis from RNA-Seq data collected under a single biological condition. This workflow is also compatible with the single-end sequencing reads although parameters have to be changed to accommodate this input. This workflow is a multi-step analysis with preprocessing/ data cleaning, taxonomy analysis and functional analysis. The starting data input for the workflow are the FASTQ files - forward and reverse reads (obtained from the Illumina sequencer). We describe below the tools and their functions in the workflow. For comparative analysis of multiple
Figure 1. ASaiM-MT workflow: The workflow is divided into 4 parts. (i) Preprocessing: Process raw metatranscriptomics data to perform further analysis. (ii) Taxonomy Quantitation: Assignment of taxonomy along with abundance values and visualization. (iii) Functional Quantitation: metabolic assignment of identified functions and gene and pathway abundance annotation. (iv) Taxonomy-Function Quantitation: combine taxonomy and functional quantitation values into relative abundance values at different levels such as e.g., the abundance of a pathway between phyla.

1) Preprocessing
   i) Input files:
      Our optimized ASaiM-MT workflow (available via https://metagenomics.usegalaxy.eu/) accepts Illumina paired end FASTQ sequence files (Forward read and Reverse read). As an alternative, a single-end FASTQ sequence can also be used as an input, with minor modifications in the downstream processing tool (such as changing the sequence type in CutAdapt and Filter with SortmeRNA as single end reads and also removing the FASTQ interlacer tool).

   ii) Quality control:
      Occasionally, sequencing can introduce incorrect identification of nucleotides and these errors can lead to misinterpretation of the data, thus bringing in the need to preprocess (Figure 2(a)) the data before analysis. The first step in our analysis is to perform quality control to remove such sequencing errors. For this, we use FastQC to assess the quality of each sample and MultiQC to combine each result into a single report. The quality control profiles and the rRNA sequence proportion changes in case of metagenomics data.

   iii) Adapter Trimming:
      To improve the quality of the data, CutAdapt was used to trim low-certainty bases from reads, filter out reads of poor quality or short length, unwanted sequences, including adapters, primers, and poly-A tails. The ASaiM-metagenomics shotgun workflow uses Trim Galore! for trimming of adapters. Trim Galore! works as a wrapper that includes CutAdapt and FastQC. However, for the ASaiM-MT workflow we chose CutAdapt for adapter trimming because it is more error tolerant, processes fast and modifies and filters reads according to user’s preference compared to TrimGalore!.

   iv) RNA Filtering:
      Next, SortMeRNA was used to remove any rRNA sequences, which are often used for easy taxonomic characterization of microbiomes but do not provide functional information.
| Tool Function          | ASaiM Shotgun Metagenomics | ASaiM-MT | Updates                        |
|------------------------|-----------------------------|----------|-------------------------------|
| Quality control        | FASTQC                      | FASTQC   | Version change (0.69 → 0.72)  |
| Adapter Trimming       | TrimGalore!                 | CutAdapt | Tool replaced                 |
| Dereplication          | VSearch                     | -        |                               |
| rRNA selection         | FilterwithSortmeRNA         | FilterwithSortmeRNA | Version change (2.1b.4 → 2.1b.6) |
| Interlacing            | FASTQ-Join                  | FASTQ interlacer | Tool replaced              |
| Taxonomic assignment   | MetaPhlAn2                  | MetaPhlAn2 | No change                   |
| Formatting for the     | Format MetaPhlAn2           | Format MetaPhlAn2 | No change               |
| different taxonomic    |                             |          |                               |
| levels                 |                             |          |                               |
| Functional assignation | HUMAnN2                     | HUMAnN2  | No change                   |
| Visualization          | Export to GraPhlAn          | Export to GraPhlAn | Parameters changed |
|                       | GraPhlAn                    | GraPhlAn | No change               |
| Regroup to GO terms    | Group abundances            | Group abundances | Tool updated |
| Unpack Pathway         | -                           | Unpack Pathway abundance to show gene families | Tool added |
| abundance             |                             |          |                               |
| Extracting Gene level  | -                           | Create gene level families file | Tool added |
| information            |                             |          |                               |
| Text manipulation      | -                           | Select, Sort, Group | Tools added |
| tools                  |                             |          |                               |

We eliminated the step of de-replication of reads (*V-Search*) in the ASaiM-MT workflow, in order to retain the multiple copies of sequences for metatranscriptomics quantitation.

The final step in cleaning and processing the data is to interlace the forward and reverse reads since the following steps require a single file per sample. For performing this action, the original ASaiM Shotgun workflow used the *FASTQ-joiner* to join the reads. However, in the ASaiM-MT version, we use the *FASTQ interlacer*. *FASTQ interlacer* joins the forward (/1) and the reverse reads (/2) using the sequence identifiers; sequences without designation will be named as single reads. The reason ASaiM-MT uses *FASTQ-interlacer* rather than *FASTQ-joiner* is because the joiner tool combines the forward and reverse read sequence together while the interlacer puts the forward and reverse read sequences in the same file while retaining the entity of each read along with an additional file with unpaired sequences. The interlacer tool mainly replaced the joiner tool in the ASaiM-MT workflow, as we wanted the output file to maintain the integrity of the reads and to distinguish between forward and reverse reads. We perform the interlacing on the data both before and after the *SortMeRNA* step since the following steps require both data with and data without rRNA.

2) **Extraction of taxonomic profile**

To understand a microbial community, we must first understand which organisms are present along with their abundance. There are several approaches to microbial taxonomic profiling, but this workflow (Figure 2(b)) uses the marker gene approach.

i) **Taxonomic profile:**

*MetaPhlAn2* checks every read against a database of approximately one million clade-specific marker genes from nearly 17,000 reference genomes (bacterial, archaeal, viral, and eukaryotic)\(^2\). For this particular step, we use all reads, including rRNA since they are useful for taxonomic
profiling. The outputs of MetaPhlAn2 are a SAM (sequence alignment map) file and a BIOM (Biological Observation Matrix) file, which both show the mapping of reads onto the reference database, and Community Profile tabular output, which contains information regarding the taxa present, and their relative abundance. This table includes information at all taxonomic levels, so to parse it out into each separate level we use the Format MetaPhlAn2 tool.

3) Extraction of functional information

i) Functional information:
After characterizing the taxonomic profile of each sample, we must determine which genes are expressed and the biological processes involved. To perform functional profiling (Figure 2(c)), we use HUMAnN2, which is a pipeline to quickly and accurately determine the presence and abundance of functional gene families and pathways from metagenomic or metatranscriptomics data.

ii) Normalization and Formatting HUMAnN2 output:
For identifying the functions expressed by the community, we filter out the rRNA sequences, due to their high noise levels and the compute time needed for their analysis. The output of SortMeRNA and the identified community profile from MetaPhlAn2 helps HUMAnN2 to focus on the known sequences for the identified organisms. Software tools such as ‘Renormalize’, ‘Unpack Pathway abundance to show gene families’, ‘Create gene level families file’, ‘Group abundances’ and text manipulation tools such as ‘Select
Figure 3. Graphical representation of Biogas reactor Dataset. A 100 µl inoculum was collected from a lab-scale biogas reactor incubated at 55 °C and transferred to an anaerobic flask containing 10 g/L of cellulose. Triplicate mRNA samples were taken in a time series from 0 to 43 hours after inoculation. Metagenomic and metaproteomic sequencing was performed for all time points. For this tutorial we used one of the triplicates (T1A) in the 8 hours’ time point.
| Sample | MetaPhlAn2_Analysis |
|--------|---------------------|
| k_Bacteria | 0.0689 |
| k_Archaea | 0.0428 |
| k_Bacteriota | 0.0428 |
| k_Archaea | 0.0428 |
| k_Archaea | 0.0428 |
| k_Bacteriota | 0.0428 |
| k_Archaea | 0.0428 |
| k_Archaea | 0.0428 |
| k_Archaea | 0.0428 |
| k_Archaea | 0.0428 |

**Figure 4. MetaPhlAn2 community profile.** A tabular representation of MetaPhlAn2 community profile displaying the different levels of taxonomic classification and its relative abundance at that level.

The **HUMANN2** suite of tools were used to extract functional information along with their relative abundance (RPK). **HUMANN2** provides 3 different outputs- Gene family and their abundance (Extended data: Supplementary Figure S2a\(^49\)), pathways and their abundance (Extended data: Supplementary Figure S2b\(^49\)) and pathway and their coverage (Extended data: Supplementary Figure S2c\(^49\)).

In this workflow, the UniRef50 database was used to classify the gene family, but the users also have a choice to use UniRef90. Gene family abundance at the community level is stratified to show the contributions from known and unknown species. The gene family output shows total abundance value which is the sum total of individual species abundance values (reported as RPK values). Additionally, the tabular output also enlists the contribution of individual species to the gene family abundance (Extended data: Supplementary Figure S3\(^49\)).

The functional and taxonomic annotations from MetaPhlAn2 and HUMANN2 are further normalized and combined to create a single tabular output.

**Use cases**

Here we provide a trimmed version of the biogas reactor dataset to demonstrate the use of the ASaiM-MT workflow in the tutorial available in the GTN.

**Link:** [https://training.galaxyproject.org/training-material/topics/metagenomics/tutorials/metatranscriptomics/tutorial.html](https://training.galaxyproject.org/training-material/topics/metagenomics/tutorials/metatranscriptomics/tutorial.html)

**Trimmed input:** [https://zenodo.org/record/3362849](https://zenodo.org/record/3362849)

**Workflow:** [https://training.galaxyproject.org/training-material/topics/metagenomics/tutorials/metatranscriptomics/workflows/](https://training.galaxyproject.org/training-material/topics/metagenomics/tutorials/metatranscriptomics/workflows/)

**Discussion**

The ASaiM-MT workflow is made available in the Galaxy platform, enabling accessibility, shareability, and flexibility for customization. There a few tools for preprocessing are made available in Galaxy such as Trimmomatic, PRINSEQ, TrimGalore, etc. that can be used alternatively or in addition to the existing tools, however, they haven’t been tested rigorously for metatranscriptomics data. The ASaiM-MT workflow (as mentioned...
Figure 5. Visualization of Taxonomy output. a) The Krona output provides interactive representation of the community in the sample. In this case *Coprothermobacter* and *Clostridium* were the most abundant genera. b) The GraPhlAn outputs circular phylogenetic trees showing that Archaea and bacteria are present in the sample.
and interpreting results. Furthermore, we are in the process of developing tools that can help perform multi-omics studies by integrating results from the ASaiM-MT workflow to our existing metaproteomics workflows.

Conclusion
ASaiM-MT workflow is a robust and extensible Galaxy workflow which is now optimized and tested for metatranscriptomics data. The workflow consists of tested open-source tools in the area of RNA sequence analysis, such as SortMeRNA, MetaPhlAn2 and HUMAnN2. ASaiM-MT in Galaxy offers users a high-level control over their data and provides different analysis options. The GTN offers documentation and resources necessary for new users to gain mastery in using this workflow and associated tools for data analysis for their research projects. ASaiM-MT allows users to not only understand the taxonomy but also the functional composition and pathways expressed by the microbiome present in diverse microbial communities of interest.

Data availability
Underlying data
Zenodo: Training Data for “Metatranscriptomics analysis using microbiome RNASeq data”, http://doi.org/10.5281/zenodo.336284927.

Extended data
Zenodo: Supplementary for ASaiM-MT: A validated and optimized ASaiM workflow for metatranscriptomics analysis within Galaxy framework, http://doi.org/10.5281/zenodo.434139149.

This project contains the following extended data:
- Supplementary Figure S1: MetaPhlAn2 Genus-Level Abundance
- Supplementary Figure S2a: HUMAnN2 Gene Family Abundance

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Software availability
Software available from: https://metagenomics.usegalaxy.eu/

Source code available from: https://github.com/ASaiM/framework

Archived source code at time of publication: http://doi.org/10.5281/zenodo.445562728.

License: Apache 2 License

Docker: https://quay.io/repository/bebatut/asaim-framework (command: docker pull quay.io/bebatut/asaim-framework).

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Version 2

Reviewer Report 23 August 2021

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Won Kyong Cho

Research Institute of Agriculture and Life Sciences, College of Agriculture and Life Sciences, Seoul National University, Seoul, South Korea

In this study, the authors provided a new workflow referred to as ASaiM for metatranscriptomics. The introduction was easy to understand the background of the study. Results and methods were written in detail. Figures and tables are very good to understand the workflow of the ASaiM-MT. In addition, the authors provided a data set for the demonstration. Overall, the quality of the manuscript is high enough for publication as it is. I hope the newly developed platform can be usefully used by many biologists working with microbiomes.

Is the rationale for developing the new method (or application) clearly explained?
Yes

Is the description of the method technically sound?
Yes

Are sufficient details provided to allow replication of the method development and its use by others?
Yes

If any results are presented, are all the source data underlying the results available to ensure full reproducibility?
Yes

Are the conclusions about the method and its performance adequately supported by the findings presented in the article?
Yes

Competing Interests: No competing interests were disclosed.
Reviewer Expertise: Metagenomics, metatranscriptomics, viromes

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Reviewer Report 27 April 2021

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Caitlin Simopoulos
Department of Biochemistry, Microbiology and Immunology, University of Ottawa, Ottawa, ON, Canada

The authors have revised the manuscript to address my concerns, and I am satisfied with the revisions. I believe the re-formatting of the Methods section helps with comprehension.

I enjoyed the figures describing the method and feel like they add to the manuscript. Inclusion of the Galaxy workflow will also help users analyze their own data using this tool. I also appreciated the links to a separate tutorial and the Gitter community. This approach will help guide and welcome new users to the tool.

Some minor suggestions are to correct two typos:
1. In the intro: “Functional Annotation (HUMAnN2⁷), Annotation of assembled contigs are subjected to gene finding programs such as FragGeneScan²” where I believe the “,” should be a “.".

2. In Figure 3 legend, I believe "metaproteomics" or "metagenomics" should be "metatranscriptomics". This may be more important to correct.

Is the rationale for developing the new method (or application) clearly explained?
Yes

Is the description of the method technically sound?
Yes

Are sufficient details provided to allow replication of the method development and its use by others?
Yes

If any results are presented, are all the source data underlying the results available to ensure full reproducibility?

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Are the conclusions about the method and its performance adequately supported by the findings presented in the article?
Yes

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: Computational biology, bioinformatics, RNA-sequencing data pipeline development and analysis, microbiomes

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Version 1

Reviewer Report 18 February 2021
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Caitlin Simopoulos
Department of Biochemistry, Microbiology and Immunology, University of Ottawa, Ottawa, ON, Canada

The authors describe an optimized workflow for metatranscriptomic microbial community data named ASaiM-MT. ASaiM-MT is an extension of ASaiM, a workflow mostly focused on metagenomic data available via Galaxy. Using helpful diagrams, the authors describe how ASaiM-MT differs from the original ASaiM workflow, and detail which tools from the Galaxy platform are used. Finally, the authors use data from a biogas reactor study the pipeline and demonstrate results that can be expected using ASaiM-MT. I particularly enjoyed the fact that the article is accompanied by written training material for the workflow. I believe that tutorials and user-friendly tools expand the audience of traditionally “bioinformatician-only” tools. However, I have a few suggestions for the authors to strengthen their presentation of the metatranscriptomic workflow before publication.

Major comments:
1. In your discussion, you highlighted a major limitation of ASaiM-MT: the fact that it can only handle a single sample and does not complete comparative analysis. However, you also mention that you have developed post-processing tools for this reason. It might be good to highlight that ASaiM-MT is essential (in particular for data pre-processing and identification of functional and taxonomic information) to be completed before statistical analysis. In
addition, is it possible to process data in ASaiM-MT in batches? This might also lessen the perceived limitation of the tool.

2. Sometimes it is not completely clear why ASaiM-MT is needed if ASaiM was developed for both meta-genomics and -transcriptomics. It is important to be specific and highlight the rationale for developing the ASaiM-MT workflow. An example: “For performing this action, the original ASaiM Shotgun workflow used the FASTQ-joiner to join the reads. However, in the ASaiM-MT version, we use the FASTQ interlacer. FASTQ interlacer joins the forward (/1) and the reverse reads (/2) using the sequence identifiers; sequences without designation will be named as single reads. The reason ASaiM-MT uses FASTQ-interlacer rather than FASTQ-joiner is because the joiner tool combines the forward and reverse read sequence together while the interlacer puts the forward and reverse read sequences in the same file while retaining the entity of each read along with an additional file with unpaired sequences.” Was this change made because FASTQ interlacer is more appropriate for RNA sequencing, or was it a change to improve on the original ASaiM workflow?

3. The abstract and introduction seem to emphasize the importance of studying the human microbiome and its connections to health and well-being, however, this manuscript uses a microbial community obtained from a biogas reactor. This is a very interesting community that warrants further research, but the introduction does not match the sample of interest. I suggest updating the introduction to focus less on the human microbiome and emphasize the importance of studying microbial communities in general. Alternatively, if you’d like to focus on human microbiomes, the example data used should reflect the human focus.

4. In your discussion, you say: “There are a few tools that can be used alternatively or in addition to the existing tools.” Does this mean that there are other tools like ASaiM-MT that can be used? Your introduction mainly listed tools that complete specific tasks and not an end to end workflow. If there are other workflow tools, a comparison should be highlighted in the discussion or introduction. What makes ASaiM-MT unique compared to other tools/workflows? What are its strengths? Is it more user-friendly than other tools? Expanding on this will help strengthen your conclusions about the tool itself.

**Minor comments:**

1. There are sections that can read like a “grocery list” of software names (eg. second paragraph of the introduction). I understand why you are listing them, but it could be useful to emphasize that each of these tools performs a single task and need to be put together into a workflow for a complete experiment. These tools are also often missing citations.

2. It can be difficult to understand which words are actually software names. Is it possible to type them in a monospace font? Or to bold the names?

3. At times the “Methods” section can get confusing, particularly when MT is being compared to the original ASaiM. My suggestion is to include numbers or roman numerals in the “in between” steps as described in Figure 2. That way you can also reference them in the text. For example, in “a) Preprocessing” there would be i. Input files, ii. Quality Control, iii. Adapter Trimming... etc. It might also be useful to include these headers (and numbers/letters) in your Methods section to let readers follow along.
4. “PMID 30298254” is used instead of a citation

Is the rationale for developing the new method (or application) clearly explained?
Partly

Is the description of the method technically sound?
Partly

Are sufficient details provided to allow replication of the method development and its use by others?
Yes

If any results are presented, are all the source data underlying the results available to ensure full reproducibility?
Yes

Are the conclusions about the method and its performance adequately supported by the findings presented in the article?
Partly

**Competing Interests:** No competing interests were disclosed.

**Reviewer Expertise:** Computational biology, bioinformatics, RNA-sequencing data pipeline development and analysis, microbiomes

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard, however I have significant reservations, as outlined above.

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**Author Response 25 Mar 2021**

**Subina Mehta**, University of Minnesota, USA

We greatly appreciate the reviewer’s comments and suggestions. We have incorporated the required changes in the updated version of the manuscript.

**Major comments:**

1. In your discussion, you highlighted a major limitation of ASaiM-MT: the fact that it can only handle a single sample and does not complete comparative analysis. However, you also mention that you have developed post-processing tools for this reason. It might be good to highlight that ASaiM-MT is essential (in particular for data pre-processing and identification of functional and taxonomic information) to be completed before statistical analysis. In addition, is it possible to process data in ASaiM-MT in batches? This might also lessen the perceived limitation of the tool.

**A:** We would like to thank the reviewer for the comment. We have edited the abstract and
mentioned that ASaiM-MT is essential for preprocessing and identification of taxonomy and functional information before performing comparative analysis. Also, Galaxy has a function of providing datasets as a collection rather than a single input which can be used to handle multiple datasets. In the Galaxy platform, we have other software tools such as MT2MQ and metaQuantome, which can perform comparative statistical analysis.

2. Sometimes it is not completely clear why ASaiM-MT is needed if ASaiM was developed for both meta-genomics and -transcriptomics. It is important to be specific and highlight the rationale for developing the ASaiM-MT workflow. An example: “For performing this action, the original ASaiM Shotgun workflow used the FASTQ-joiner to join the reads. However, in the ASaiM-MT version, we use the FASTQ interlacer. FASTQ interlacer joins the forward (/1) and the reverse reads (/2) using the sequence identifiers; sequences without designation will be named as single reads. The reason ASaiM-MT uses FASTQ-interlacer rather than FASTQ-joiner is because the joiner tool combines the forward and reverse read sequence together while the interlacer puts the forward and reverse read sequences in the same file while retaining the entity of each read along with an additional file with unpaired sequences.” Was this change made because FASTQ interlacer is more appropriate for RNA sequencing, or was it a change to improve on the original ASaiM workflow?

A: Thank you for this comment. We replaced the FASTQ-joiner tool with the FASTQ interlacer to improve on the original workflow. The FASTQ interlacer tool maintains the integrity of reads by maintaining the forward and the reverse sequence identifiers, as compared to joining them into a single read file, as is done by FASTQ-joiner tool.

3. The abstract and introduction seem to emphasize the importance of studying the human microbiome and its connections to health and well-being, however, this manuscript uses a microbial community obtained from a biogas reactor. This is a very interesting community that warrants further research, but the introduction does not match the sample of interest. I suggest updating the introduction to focus less on the human microbiome and emphasize the importance of studying microbial communities in general. Alternatively, if you’d like to focus on human microbiomes, the example data used should reflect the human focus.

A: We thank the reviewer for the comment and have edited the abstract and introduction to highlight the importance of microbiome research in ecology as well as clinical research.

4. In your discussion, you say: “There are a few tools that can be used alternatively or in addition to the existing tools.” Does this mean that there are other tools like ASaiM-MT that can be used? Your introduction mainly listed tools that complete specific tasks and not an end to end workflow. If there are other workflow tools, a comparison should be highlighted in the discussion or introduction. What makes ASaiM-MT unique compared to other tools/workflows? What are its strengths? Is it more user-friendly than other tools? Expanding on this will help strengthen your conclusions about the tool itself.

A: We thank the reviewer for the comment. The tools mentioned in the discussion offer an
alternative option to the tools that we have used in the ASaiM-MT workflow, especially if the users have a preference. However, although we consider these tools to be appropriate to metatranscriptomics research, we have not tested these tools.

ASaiM-MT workflow, due to its availability in Galaxy, offers an user-friendly option to the existing command line tools. The ASaiM-MT workflow has been tested with different datasets to ensure its compatibility. There is a systematic documentation available for the usage of the workflow in the Galaxy Training Network (GTN). Users can also ask questions to developers and users via the Gitter channel, if needed.

**Minor comments:**

1. There are sections that can read like a “grocery list” of software names (eg. second paragraph of the introduction). I understand why you are listing them, but it could be useful to emphasize that each of these tools performs a single task and need to be put together into a workflow for a complete experiment. These tools are also often missing citations.

   **A:** Thanks to the reviewer for the comment. The tools listed are alternatives to the existing tools in the workflow and have not been incorporated as a workflow. We have added citations to these tools.

2. It can be difficult to understand which words are actually software names. Is it possible to type them in a monospace font? Or to bold the names?

   **A:** Thanks to the reviewer for this comment. I have formatted the tool names by making them italic.

3. At times the “Methods” section can get confusing, particularly when MT is being compared to the original ASaiM. My suggestion is to include numbers or roman numerals in the “in between” steps as described in Figure 2. That way you can also reference them in the text. For example, in “a) Preprocessing” there would be i. Input files, ii. Quality Control, iii. Adapter Trimming... etc.. It might also be useful to include these headers (and numbers/letters) in your Methods section to let readers follow along.

   **A:** Thank you for the comment. I have reformatted the method section according to Figure 2.

4. “PMID 30298254” is used instead of a citation

   **A:** Thank you reviewer for pointing this out. I have made the required change.

**Competing Interests:** No competing interests were disclosed.
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