

# **ONE ARM Metagenomics Pipeline**

Prepared for and by: Project 4

Last updated September 19, 2024

# Contents

<b>I</b>	<b>Project overview</b>	<b>3</b>
1	HPC Preparation . . . . .	4
<b>II</b>	<b>Project 4 Scripts</b>	<b>8</b>
1	ARG-MGE.smk . . . . .	9
1.1	Pair merging . . . . .	9
1.2	Translation and Reverse Translation . . . . .	9
1.3	Iterative alignment of ARG-contigs . . . . .	9
1.4	Merging of paired reads . . . . .	10
1.5	Guided Metagenomic Assembly . . . . .	10
1.6	Confirmation of contigs with ARGs . . . . .	10
1.7	Standard Contig Quality Metrics . . . . .	11
1.8	Read-mapping . . . . .	11
	Read mapping parameters . . . . .	11
1.9	Taxonomic profiling of reads and contigs . . . . .	12
1.10	Detect structural variants (SVs) in contigs . . . . .	12
1.11	Sketching contigs followed by calculating Bray-Curtis diversity . . . . .	13
1.12	Determination of Transposons (TBA) . . . . .	14
1.13	Determination of putative plasmids . . . . .	14
1.14	Phage influence signatures . . . . .	14
1.15	Phage Signature Extraction and Phylogenetic Analysis . . . . .	15
1.16	Future Considerations . . . . .	15
2	metagenomics_general.smk . . . . .	16
2.1	Quality control (Pre-processing) . . . . .	16
2.2	Trimming . . . . .	16
2.3	Metagenomic Taxonomy . . . . .	17
2.4	Diversity analysis . . . . .	17
3	trim_randomizer.smk . . . . .	19
3.1	Random Parameter Generation . . . . .	19
3.2	Log Parameters to a TSV File . . . . .	19
3.3	Define Rules for Each Tool . . . . .	19
3.4	Interpretation and Analysis of Results (TBA) . . . . .	20
4	bootstrapping_rawreads.smk . . . . .	21

4.1	Directory setup of temporary files . . . . .	21
4.2	Sample Identification . . . . .	21
4.3	Bootstrapping . . . . .	21
4.4	Run <code>kraken_pipeline.bash</code> . . . . .	22
5	Binning.smk . . . . .	23
5.1	Universal Configurations . . . . .	23
5.2	FastUniq (Deduplication) . . . . .	23
5.3	Seqtk (FASTA Conversion) . . . . .	23
5.4	CD-HIT-EST (Clustering) at Identity: 90% . . . . .	24
5.5	MetaWRAP Binning and Reassembly . . . . .	24
5.6	DAS Tool (Bin Refinement) . . . . .	24
5.7	MAGpurify (Contamination Removal) . . . . .	24
5.8	MetaQUAST (Assembly Quality Assessment) . . . . .	25
5.9	dRep (Dereplication) . . . . .	26
5.10	CheckM2 . . . . .	26
6	<code>krakenpipeline.bash</code> . . . . .	26
<b>III Project Side Scripts</b>		<b>27</b>
1	. . . . .	28
<b>IV Project Main Scripts</b>		<b>29</b>
1	. . . . .	30
<b>V Daily entries</b>		<b>31</b>
1	September 20, 2024 . . . . .	32
2	September 19, 2024 . . . . .	33
3	September 18, 2024 . . . . .	34
4	September 17, 2024 . . . . .	35

## **Part I**

# **Project overview**

Metagenomic analysis of Antibiotic/Antimicrobial Resistance Genes (ARGs) in NCR (Metro Manila) hospitals, wastewaters, and surface waters.

☒ Done
 

- Essential

☐ Pending
 

- Optional

☐ Needs refinement
 

- Robust
- Deprecated

☐ Unexpected issues

☐ Drafted

☐ Moved

## 1 HPC Preparation

- ☐ Confirmation of Robustness of analysis
  - ☐ Agree upon the type of analyses
- ☐ SLURM request management
  - ☐ Send email for HPC usage requests
  - ☐ Setup Docker containers for SLURM
  - ☐ SLURM container for simulations
  - ☐ Automation of SLURM requests

## File Management

- ☐ BAM file parsing
- ☐ Interconversion between SAM and BAM

## Raw Read Processing

- ☐ Pipeline integration of raw.bash
- ☐ Data visualization

- ☐ Determination of optimal tool/s
- ☐ Determination of optimal parameters
- ☐ Tool combination randomization script

## Assembly & Binning

- ☐ Testing binning pipeline

### Creation

- ☐ metaSPAdes
- ☐ MASURCA
- ☐ PLASS
- ☐ AbySS

### Testing

- ☐ metaSPAdes
- ☐ MASURCA
- ☐ PLASS
- ☐ AbySS
- ☐ KMA-iterative
- ☐ Benchmarking between all of them
- ☒ Iterative assembly
- ☒ Contig quality checking

### Integration

- ☐ MetaWrap binning
  - Binning.smk
- ☐ Testing on datasets of higher depth for MetaWrap tests
- ☐ Refinement of bins

## Annotation

### ARG-related scripts

- RGI
  - ~~prokka\_ARG.bash,~~
  - ~~shortbred\_ARG.bash~~

- ShortBRED
  - shortbred\_ARG.bash

- AMRFinder

### Testing of Annotation Scripts

- ☐ RGI
- ☐ ShortBRED
  - shortbred\_ARG.bash
- ☐ AMRFinder

### Taxonomy related scripts

- ☐ MetaPhlan4

## Complete

### General Taxo-Metagenomics Pipeline

- 1 Creation of modular scripts
- 2
- 3 ✓ FasQC script for raw reads
  - 4 raw.bash
- 5 ✓ Trimmomatic script
  - 6 Trimmomatic.bash
  - 7 trimming-cleaning-checking.py
- 8 ✓ fastp
  - 9 fastp.bash
  - 10 trimming-cleaning-checking.py
- 11 ✓ Script to check if reads PASS or FAIL
  - 12 FastQC.bash
  - 13 trimming-cleaning-checking.py
- 14 ✓ FasQC script for trimmed reads
  - 15 FastQC-check.py
  - 16 trimming-cleaning-checking.py

- 17 ✓ Kraken2 script
  - 18 krakenpipeline.bash
- 19 ✓ Bracken script
  - 20 krakenpipeline.bash
- 21 ✓ In-house diversity calcs script
  - 22 calculate\_diversity.py
- 23 Test Run of Modular Scripts
- 24 ✓ FasQC script raw.bash
- 25 ✓ Trimmomatic script
  - 26 Trimmomatic.bash
  - 27 trimming-cleaning-checking.py
- 28 ✓ fastp
  - 29 fastp.bash
  - 30 trimming-cleaning-checking.py

- 31 ✓ Kraken2 script
  - 32 krakenpipeline.bash

- 33 ✓ Bracken script
  - 34 krakenpipeline.bash

- 35 ✓ In-house diversity calcs script
  - 36 calculate\_diversity.py

### Other Completed Pipeline Tasks

- 37
- 38 ✓ Integration with config folder
- 39 ✓ Testing with simulated datasets
- 40 ✓ Pipeline integration
  - 41 metagenomics\_general.smk
  - 42 ✓ FastQC
    - 43 trimming-cleaning-checking.py
  - 44 ✓ Trimmomatic
    - 45 trimming-cleaning-checking.py
  - 46 ✓ Kraken2
    - 47 krakenpipeline.bash
  - 48 ✓ Bracken
    - 49 krakenpipeline.bash
  - 50 △ QIIME2 krakenpipeline.bash
  - 51 ✓ Change QIIME2 to in-house diversity script
    - 52 calculate\_diversity.py

53 ✓ Aggregate diversity metrics  
 54 calculate\_diversity.py  
 55 ✓ Aggregate quality metrics  
 56 summary\_stat.bash  
 57 ✓ Remove fastp from Shellsript  
 58 ✓ Testing pipeline with simulated datasets  
 59 metagenomics\_general.smk

## File Management

60  
 61 ✓ Compression and decompression scripts  
 62 ✓ Demultiplexing script  
 63 ✓ Testing on multiple datasets

## Read Quality Control

## Creation

65  
 66 ✓ Parametric randomization script  
 67 trim\_randomizer.smk

- 68 1. Trimmomatic
- 69 2. CutAdapt
- 70 3. BBDuk
- 71 4. Sickle

72 5. fastp  
 73 ✓ Parsing QC metrics of all iterations  
 74 parseFastQC.py  
 75 ✓ Bootstrapping script  
 76 bootstrapping\_rawreads.smk

## Test-run

77  
 78 ✓ Parsing and aggregation of FastQC metrics  
 79 from multi-site data (N=10)  
 summary\_stat.bash  
 80 ✓ Randomization of trimming parameters from  
 81 different trimmers simultaneously (5 trim-  
 82 mers total)  
 trim\_randomizer.smk  
 83 ✓ Bootstrapping of randomization of trimming  
 84 parameters. 5 Trimmers, 1000 bootstraps  
 bootstrapping\_rawreads.smk

## Pipeline Preparation for HPC

85 ✓ Calculate using quotation from PGC  
 86 ✓ Set up config files  
 87 ✓ Taxo-metagenomics pipeline

# Script Descriptions

## Pipelines

This section covers all the scripts that were created during the Project.

*Listen, I'm eternally curious and I don't just want to settle for any random journal. No offense, but I'm aiming for something high-impact!*

These were created during off hours or during work hours, so some scripts might seem irrelevant at first, *but trust me, there's a conscientiousness or meticulousness to this madness.*

I have separated them into folders/repositories (shameless plug here: <https://github.com/GABallena>) based on their relevance to the project.

- **Project4** - Essential scripts directly related to the core analyses of the project.
- **Side** - Scripts that can potentially be used to increase the robustness of the paper.
- **Main** - Scripts related to file organization and data management, crucial for handling large datasets.



# **Project 4 Scripts**

# 1 ARG-MGE.smk

## 2 Stage: Draft General Purpose

3 This pipeline is designed for **comprehensive metagenomic analysis of ARGs**, it also includes  
4 placeholders for analyses of **mobile genetic elements (MGEs)**, and **plasmid detection**. It integrates  
5 tools for **read quality control, assembly, annotation, taxonomic profiling, and structural variant**  
6 **(SV) detection** to provide a high-resolution view of the genetic components in metagenomic samples.

## 8 Preprocessing

---

### 10 1.1 Pair merging

11 **Technical Notes:** PEAR (Paired-End reAd mergeR) compares paired-reads to correct infer the likely  
12 bases on its associated pair

13 **Rationale:** The main goal here is to ensure good read quality for downstream analyses and to maximize  
14 the amount of data by reducing gaps in the sequence and improving confidence of base calls within that  
15 region. seqtk then converts compressed FASTQ files to FASTA.

#### 16 Rationale

17 **Note:** Another tool, PandaSeq which does the same thing is used later, this usage of PEAR here is because  
18 it is more optimized for larger datasets - which in this case are trimmed reads. **Note:** Conversion is  
19 necessary here as some tools cannot read FASTQ files, and instead rely on FASTA formats.

### 20 1.2 Translation and Reverse Translation

21 **Technical Notes:** transeq from EMBOSS translates nucleotide sequences to protein sequences based  
22 on **standard genetic code**. backtranseq reverses this to allow iterative alignments with nucleotide  
23 sequences.

24 **Rationale:** This leverages conserved protein-level information, which is lost at the nucleotide level due  
25 to synonymous mutations - while also increasing the sensitivity to potential ARG proteins from k-mer  
26 alignment. See Nature Methods, doi: doi.org/10.1038/s41592-019-0437-4 (2019) for more details.

## 28 Metagenomic Assembly

---

### 30 1.3 Iterative alignment of ARG-contigs

31 **Technical Notes:** KMA (K-mer Alignment) is used find the reverse-translated ARGs with the proximity  
32 filtering option to determine the surrounding regions of ARGs. This is done iteratively for each gene  
33 increasing the ARG-associated database.

34 **Rationale:** Firstly, KMA is used because, unlike Bowtie2 and BWA-MEM, which were created specifically  
35 for Human metagenomics, KMA does not suffer (or suffers less) from multi-allelic databases. Secondly,  
36 iterative alignment using this process allows us to contextualize the region wherein ARGs reside - thereby  
37 narrowing our focus onto these local regions instead of looking at the global genomic context. **Notes:**

The script is designed to have a cap on the number of iterations KMA creates, increasing the database size.

## 1.4 Merging of paired reads

**Technical Notes:** PANDASeq is then used to further refine the paired-reads collated in the ARG-related genes database.

**Rationale:** PANDASeq was chosen as, while being slower than PEAR, it is more accurate. This merging step is included to ensure that only high-confidence reads are assembled. **Note:** We leverage the fact that the ARG-related genes database is smaller compared to the raw metagenomic reads database.

## 1.5 Guided Metagenomic Assembly

**Technical Notes:** metaSPAdes is then used to create contiguous sequences from these local regions by extending them using reads from the whole metagenomic pool. Additionally, Contigs are filtered by length here to remove possible artefacts.

**Rationale:** metaSPAdes was chosen as, while being slower than MEGAHIT, it is optimized in handling highly diverse and mixed microbial populations. CARD is used here because it is manually curated and updated regularly - in order to be included in the database, there must be clinical data (e.g. ASTs(Antimicrobial susceptibility testing)) involved in the study.

**Notably, this would decrease the sensitivity of our ARGs - and would mostly be biased towards those reported in the clinical setting.** To counter this, we could also incorporate other tools such as ResFinder, the NCBI AMR Database, and ARG-ANNOT

**Notes:** Filtering of contig length is handled by a python script called `minimum_length_CARD.py`.

**Notes:** Contigs are further extended using contigextender to form scaffolds

**Notes:** Might add other contig extending programs like GapFiller - which leverages mate-pair information

## Contig Quality Checks

---

## 1.6 Confirmation of contigs with ARGs

**Technical Notes:** RGI scan the contigs and check whether which contigs created by metaSPAdes have ARGs in them.

**Rationale:** metaSPAdes may have created contigs that DO NOT contain ARGs, and have instead assembled them into a more matching contig (a false-positive misassembly) - this can happen because of the different databases being used; also parallelization of methods like this increases robustness because it has been confirmed independently from different starting points (bottom-up vs top-down approach). This allows us to filter ARG-containing contigs.

**Notes:** RGI is the official scanner of CARD.

## 1.7 Standard Contig Quality Metrics

**Technical Notes:** A custom Python script `calculate_contig_quality.py` is created to do another round of checking contig quality for downstream analysis, R scripts (TBA) are used to visualize the data.

**Technical Notes:** The Python script will measure standard contig quality metrics e.g. N50, L50.

**Rationale: Note:**

## 1.8 Read-mapping

**Technical Notes:** Samtools is used here to map the raw reads from the larger database back to the assembled contigs and then calculates the coverage over the entire contig.

**Rationale:** Read-mapping is a quality control protocol used in metagenomics, to determine the quality of the assembly. High-coverage means that many of the k-mers align well with that region of the contig, while low coverage is evidence of inconsistent mapping and that the contigs should be refined, split, or discarded.

**Note** If there are persistent (after further refinement and reassemblies) sudden differences in coverage across a contig, that contig could be chimeric, meaning, it could be from two different populations.

**Extra note** A Python script is included in the pipeline that is determine visualize and check how the coverage changes over contig regions. In general, they could be interpreted as the following:

1. **Smooth, Uniform Coverage:** Typically shown by well-assembled contigs.
2. **Sharp Coverage Drop:** May need to be split or flagged for reassembly. May also be a misassembly point (chimeric contig) or caused by a structural variant.
3. **Coverage Gaps:** Regions with little to no read support; a strong indicator of misassembly.
4. **Gradual Drops:** Overlapping reads, repetitive or duplicate regions, partial HGT, sequence heterogeneity, or coverage differences due to a mixed population. Repeats and duplications can be filtered out using tools like RepeatMasker or BLAST (to be added to the script).
5. **Sharp Increase:** May be due to repetitive or duplicated regions, amplification bias from PCR, HGT, SV, chimeras.

### Read mapping parameters

**Technical Notes:** Four (4) Tools will be used in parallel to do the read mapping process BWA Bowtie KMA, and minimap2, their parameters have been adjusted to map reads at 95 % identity to the contigs.

Mapper	K-mer Length	Mismatch Penalty	Gap Opening Penalty	Gap Extension Penalty
BWA	21	5	7	2
BWA	31	4	6	2
BWA	51	3	5	1
Bowtie2	-	4,2	5,2	5,2
KMA	Default	95% identity	Automatic	Automatic
Minimap2	-	5	7,2	4,1

Table 1: Example table of mapper configurations without command example

**Rationale:** 95 % identity is used to increase sensitivity - as is standard for determining homologous sequences. This adjustment was made because k-mers are either attach or don't. Parallelization is used to increase robustness.

**Note** K-mer extension is used to increase the accuracy of mapping. BWA k-mer lengths can be adjusted, while KMA does it by default. The others, cannot be adjusted.

I chose to change the script to not allow gaps during this phase as we already used reverse translation earlier to correct for synonymous codons, and protein sequences are more important when it comes to ARG function, the new values are below. I also added protein-based read mapping.

Tool	K-mer Length	Mismatch Penalty	Gap Opening Penalty	Gap Extension Penalty
BWA	21, 31, 51	5, 4, 3	1000	1000
Bowtie2	-	4,2	1000,1000	1000,1000
KMA	-	95% identity	Automatic	Automatic
Minimap2	-	5	1000,1000	1000,1000

Table 2: Alignment parameters for ungapped alignments across BWA, Bowtie2, KMA, and Minimap2

**Rationale:** The main rationale for adding a protein-based read mapping protocol is because ARGs are primarily about their *protein-protein interactions* (biological relevance). This method also accounts for frameshifts with higher specificity to homologous regions.

**On a personal note:** This approach may also require further exploration into whether protein-protein interactions are altered—perhaps by investigating changes in binding sites. Which is a story for another day (*why do I do this to myself?*)

## 1.9 Taxonomic profiling of reads and contigs

**Technical Notes:** Kraken2 uses k-mer-based classification to assign taxonomy based on raw-reads. While SprayNPray complements this by assigning taxonomy at the contig level.

**Rationale:** By comparing their respective databases with our ARG-related databases, we will be able to connect our reads and/or contigs to their corresponding taxa, uncovering the microbial hosts responsible for carrying and potentially spreading ARGs in the environment.

## 1.10 Detect structural variants (SVs) in contigs

**Technical Notes:** Manta identifies large genomic rearrangements such as insertions, deletions, and duplications.

**Rationale:** Chimeric contigs may be due to systematic error or real biological signals. These chimeric contigs can be detected by Kraken2 and SprayNPray (i.e., when a portion of a contig is being assigned to different taxa). The rationale behind this step is to investigate whether structural variations are present — which may be evidence of horizontal gene transfer (HGT) events.

Tool	Input Files	Key Parameters
<b>tblastn</b>	<ul style="list-style-type: none"> <li>• Protein sequences: Translated protein sequences contigs</li> <li>• Nucleotide database: Cleaned sequence database</li> </ul>	<ul style="list-style-type: none"> <li>• -outfmt 6</li> <li>• -evaluate 1e-5</li> <li>• -gapopen 5</li> <li>• -gapextend 2</li> <li>• -matrix BLOSUM62</li> </ul>
<b>blastp</b>	<ul style="list-style-type: none"> <li>• Protein sequences: Translated protein sequences contigs</li> <li>• Protein database: Translated cleaned sequence database</li> </ul>	<ul style="list-style-type: none"> <li>• -outfmt 6</li> <li>• -evaluate 1e-5</li> <li>• -gapopen 5</li> <li>• -gapextend 2</li> <li>• -matrix BLOSUM62</li> </ul>

Table 3: Protein read-mapping parameters for tblastn and blastp

### 1.11 Sketching contigs followed by calculating Bray-Curtis diversity

**Technical Notes:** Mash Sketch uses a MinHash approach to generate a presence/absence profile of ARGs across contigs. This gives us a quick snapshot of what the contigs “look like” in terms of ARG content. For Bray-Curtis diversity, we calculate a dissimilarity matrix from the abundance data of ARGs, followed by a PCoA plot to visualize similarities between contigs based on their ARG profiles.

**Rationale:** The Mash Sketch helps rapidly identify the genetic makeup of contigs in terms of ARGs, which provides a foundation for further investigation. By applying Bray-Curtis diversity and using PCoA, we can group contigs based on their ARG similarity. If contigs with the same sketch group together, we can trace them back to their taxonomic IDs to identify the microbial hosts. However, if contigs have similar ARG profiles but belong to different taxa, this could serve as **evidence for Horizontal Gene Transfer (HGT)**. This dual approach allows us to trace ARG spread and potential HGT events in a metagenomic context.

**Note:** Bray-Curtis (dis-)similarity is most often used as a presence or absence diversity metric.

### Transposition

## 1.12 Determination of Transposons (TBA)

**Technical Notes:** Tools such as the following can be used:

- HMMER3 suite
- Tnppred - a transposon predictor tool

**Rationale:**

## Plasmids

---

## 1.13 Determination of putative plasmids

**Technical Notes:** The tools listed below will be used in parallel. Plasmids are considered valid when all 4 tools predict plasmid signatures in the contig.

- PlasPredict pipeline
- Recycler
- PlasmidFinder
- MOBSuite plasmid marker annotator

If plasmid signatures are present, plasmidSPAdes along with GapFiller will be used to check if the contig can circularize. oriTfinder will then be applied to contigs with fewer than 4 fragments. A Python script will calculate GC skews of the chimeric contig and compare it to its taxonomic counterparts. Another Python script will normalize the data according to 16S rRNA from trimmed reads. Lastly, plasmid percentage will be calculated based on reads mapped to putative plasmids over the total reads.

**Equation:**

$$\text{Plasmid Percentage} = \left( \frac{\text{Plasmid Reads}}{\text{Total Reads}} \right) \times 100$$

**Rationale:** oriTfinder looks for origin of transfer sites (oriT), which are characteristics of conjugative plasmid. This whole sub-pipeline is to look for evidence of conjugative plasmid transfer as the cause of these chimeric contigs. Normalization and percentage counts are used here to further check whether these "plasmids" align with our understanding of the average plasmid copy number.

**Note:** Will also be drafting a script to do sliding window analysis of GC-skews - as different characteristics of this curve can be interpreted in different ways.

## Phages

---

## 1.14 Phage influence signatures

**Technical Notes:** They will be determined using a variety of tools:

- VirSorter: Identifies viral signatures within microbial genomes and separates prophages from bacterial sequences.
- PHASTER: A web-based tool for phage search and annotation, identifying integrated prophages.
- VIBRANT: A tool that combines several approaches to identify and annotate phage elements in metagenomic sequences.

**Rationale:** This analysis aims to detect potential phage signatures in the chimeric contigs. Since phages are mobile genetic elements, their involvement in transferring ARGs through transduction is highly relevant. Phages, especially temperate phages, can integrate into bacterial genomes and excise themselves, sometimes carrying host genetic material, such as ARGs, with them. The integration and excision signatures detected in contigs will provide evidence of possible transduction events in our datasets, supporting the hypothesis of ARG dissemination via phages.

### 1.15 Phage Signature Extraction and Phylogenetic Analysis

**Technical Notes:** Phage-associated genes will be extracted from the chimeric contigs, followed by phylogenetic analysis to uncover evolutionary relationships. FastTree will be used to build a phylogenetic tree based on the extracted phage genes. For visualization, tools like iTOL or FigTree can be used to generate an interpretable phylogenetic tree.

**Rationale:** Phage genes embedded in chimeric contigs (their taxonomy) may serve as strong evidence of horizontal gene transfer (HGT) events. The aim here is to check the evolutionary origins of the phage genes found in our dataset and their potential involvement in the dissemination of ARGs.

### Homework?

---

### 1.16 Future Considerations

: Will continue improving this section (everything regarding HGT) by evaluating the results from these tools, aligning them with ARG presence, and refining the approach for identifying conjugation, transposon, and transduction events within chimeric contigs. This may also involve validating phage activity—*again, why do I do this to myself?*



## 2 metagenomics\_general.smk

**Stage: Done**

**General Purpose taxo-metagenomics**

This pipeline is designed for **The essentials in metagenomics**, which includes **quality checking, filtering, and trimming of raw reads to clean reads**. As well as the usual **taxo-metagenomic analysis**.

**Specifics:**

### 2.1 Quality control (Pre-processing)

**Raw trimming of raw metagenomic data**

**Technical Notes:** FastQC is used on **raw reads**

**Rationale:** This is mainly used as a point of comparison - determine whether the next step (trimming) was effective. This pre-processing step is the starting point in any and all metagenomics pipelines.

**Notes:** This whole quality control steps are interconnected with each other.

### 2.2 Trimming

**Technical Notes:** Trimmomatic is used on **raw reads**

**Rationale:** Trimming involves removing low quality bases (often depending on something called the Phred Score - which is just a measure of how "confident" we are that the base on that site was accurate), adapters, and filtering reads that go below a specific length threshold.

**Notes:** Journals often report the parameters on Trimmomatic (or any trimmer they decide to use); this is often so that the study is reproducible, should one decide to actually reproduce the study starting from scratch (raw reads).

**Notes:** There are many different trimmers each with their own strengths and weaknesses, Trimmomatic is just the most popular trimmer and is thus used here, though studies differ in the parameters they used for trimming - which often dictates how strict they are with what they define as "good enough".

**Perspective:** Why different people choose different trimmers depend on the **strengths and weaknesses** of the trimmer e.g. trimmers like "fastp" is used because it's fast making it suitable for very large datasets like deep sequencing. While some trimmers like Sickle has automatic adjustment over the entire sequence - which makes it useful for very ancient datasets where DNA is often highly-degraded. Other times, it's for **convenience** like Trim-Galore which combines FastQC and Cutadapt trimmer in a single command. Another good example is BBDuk which is part of a larger package called BBtools, BBDuk also has an built-in contamination detection - so it's particularly good at filtering out usual contaminants like sequences known to be from the human genome. so you can simply just use all the modules in that package for all-in-one processing. Other times, it's just **familiarity**.

**Quality checks of post-trimmed data**

**Technical Notes:** FastQC is used on **trimmed reads** to determine how effective the trimming process was.

**Rationale:** If the trimming process was effective, we should notice a better quality reads here, otherwise, we might have to adjust the trimming parameters.

**Notes:** Determination of whether the trimmed reads are "clean enough" is more of an art rather than actual science, though thresholds exist like Phred > 20 or Phred > 30 depending on how strict you are as a researcher.

## Processing cleaned reads

---

### 2.3 Metagenomic Taxonomy

#### Taxonomy from Cleaned Reads

**Technical Notes:** Kraken2 is used on **raw reads** to determine which species they came from.

**Rationale:** Taxonomy based on reads is standard on metagenomics instead of using assembled contigs because information is lost during the assembly process. By using cleaned reads we make sure that:

1. We are maximizing the amount of data
2. We are not being biased by low quality reads

**Notes:** There are many ways in which taxonomy is assigned to raw reads the Kraken-based packages use a curated database that links k-mers from your database to k-mers generated from their database (usually manually curated).

**Notes:** Others like MetaPhlan first create "markers" that are based off of known sequences, and then scan your raw reads for these markers, which it then checks for under what taxon/taxa that marker fell under.

### 2.4 Diversity analysis

#### Diversity analysis per site or sample

**Technical Notes:** Here Bracken - an extension of the Kraken packages or Qiime2 are often used to calculate diversity. Instead I opted to create my own Python script because:

1. I find that the diversity indices in either tools are limited to only the most often used i.e. the most popular indices - so it is not comprehensive
2. I've had problems integrating them into the pipeline because some dependency limitations, un-updated scripts, and a pre-processing step that requires converting all of Kraken2, Bracken files, then importing them to Qiime2 which is too time-consuming and inefficient - my script just automatically calculates from Bracken outputs and just puts out all the possible (non-phylogenetic-based-which you would need a phylogenetic tree to build first) indices out there.

**Rationale:** Taxonomy based on reads is standard on metagenomics instead of using assembled contigs because information is lost during the assembly process. By using cleaned reads we make sure that:

1. We are maximizing the amount of data
2. We are not being biased by low quality reads

76 **Notes:** There are many ways in which taxonomy is assigned to raw reads the Kraken-based packages  
77 use a curated database that links k-mers from your database to k-mers generated from their database  
78 (usually manually curated).

79 **Notes:** Others like MetaPhlan first create "markers" that are based off of known sequences, and then  
80 scan your raw reads for these markers, which it then checks for under what taxon/taxa that marker fell  
81 under.

82

### 3 trim\_randomizer.smk

**Stage: Done**

**Purpose: Randomization of trimming parameters**

This is a module that will be part of a bigger pipeline. The idea here to is randomize parameters in a variety of trimmers in this case Trimmomatic, fastp, CutAdapt, BBDuk, and Sickle - famous bioinformatics trimming tools. Trimmomatic and Sickle in particular are widely used in Illumina-based data.

#### 3.1 Random Parameter Generation

**Technical Notes:** Random parameters are generated for each trimming tool (Trimmomatic, Fastp, Cutadapt, BBDuk, Sickle). This is done using the `random` module, which creates random values for parameters such as quality scores, read length, adapter sequences, and error rates. These parameters are stored in the `generated_parameters` dictionary to ensure consistency across iterations for each sample.

**Rationale:** By randomizing parameters, the workflow allows for testing different parameter sets across multiple iterations to find optimal settings for trimming and quality control.

**Notes:** This approach helps with parameter exploration, particularly when you are unsure which trimming settings will give the best results. The randomness provides variability, which can highlight which parameters consistently lead to good results.

#### 3.2 Log Parameters to a TSV File

**Technical Notes:** Each set of generated parameters is logged into a separate TSV file (e.g., `trimmomatic_params.tsv`, `fastp_params.tsv`). The file headers are written only once, and parameters are appended as the trimming steps proceed. This is done in a structured way so that you can track the exact parameters used for each sample and iteration.

**Rationale:** Logging ensures reproducibility and transparency in bioinformatics workflows. Having a record of all the parameter values used in each iteration is crucial for comparing results and for future reference.

**Notes:** This practice is a standard in scientific workflows where random parameter generation is involved. It helps maintain a clear audit trail of the steps performed and aids in troubleshooting or refining workflows later.

#### 3.3 Define Rules for Each Tool

**Technical Notes:** The script uses Snakemake rules to define separate rules for each tool which include

1. Input folder (as the script is designed to go through all the FASTQ samples within the folder)
2. Output folder, where the processed files will be saved (e.g., trimmed paired and unpaired reads).  
The script is designed to keep all the `trimmed` reads in separate files.
3. Params: Fetches the parameters to be randomized.

**Rationale:** Using Snakemake here allows for parallel execution of the workflow. This parallelization is very important as the generation of random parameters is created using a random seed. Using a random seed like this allows us to replicate what the parameters that had the optimal results were by tracking down what seed was assigned as dictated in the TSV file.

**Note** keep in mind that this will create a large number of folders if you decide to iterate many times - as each iteration, per tool, will have its own folder full of trimmed reads, per sample/site.

### 3.4 Interpretation and Analysis of Results (TBA)

**Technical Notes:** Once all iterations have completed, the trimmed files can be analyzed to determine which parameters led to the best results in terms of quality and length distribution of reads. There are many different metrics that can be used to interpret the results, including (but not limited to)

1. Quality Scores - significant differences in quality among sites and across entire sequences (Phred score, Contamination, Adapter Removal, N Content, Length Distribution, etc.). Can be done using tools like FastQC
2. Visualization can be done using Rstudio or Python's matplotlib to visually look for differences in abnormalities.

**Rationale:** Evaluating read quality and assessing key metrics post-trimming helps to ensure that the data is suitable for downstream analyses. Optimal trimming should maximize the number of high-quality, usable reads while eliminating low-quality bases and adapter contamination.

#### A Personal Note

I have long pondered upon the idea of creating a graph with biological information on the y-axis and read length on the x-axis. I hypothesize that we will find a "sweet spot" wherein we can optimize the amount of information/read using such trimmers. Unfortunately, it is extremely difficult to define what a biological sequence really is, because technically you can generate any random sequence - whether protein or nucleotide. That's the main reason I have been stuck on this problem for quite a while, I've been trying to find the answer first.

The core difficulty here lies in defining what constitutes biological information in a meaningful, quantifiable way. Since any random sequence of nucleotides or proteins could be technically "valid" (false-positives).

When people are asked this question they often give **DESCRIPTIONS** of what a biological sequence is but not what **DEFINES** a biological sequence. I understand that there are many characteristics that can help in determining the signal from the noise, but I am just not satisfied with whether this "thing" checks most (if not all) the boxes - I need a non-subjective answer to this problem.

## 4 bootstrapping\_rawreads.smk

**Stage:** Further refinement

**Purpose:** Bootstrapping the taxo\_metagenomic pipeline itself

Bootstrapping is the process of randomly selecting from a pool of samples (with replacement) and using that in a specific process you want to bootstrap. This is a standard method used in molecular phylogenetics to determine the robustness of trees where a  $> 70$  support from bootstrapped data is considered robust enough.

### The principle of bootstrapping in phylogenetics

Phylogenetics uses this statistical technique because (in principle) it effectively means that removing parts of the entire sequence does not alter the topology of the tree.

Here I'm adapting this method with Kraken2's taxonomic profiling to see whether the taxonomic support of its k-mer assignment is also consistent even with changes in the sampling sites.

This is still WIP because I plan to go step further and start bootstrapping the raw reads themselves to see if changes in reads changes the topology of taxonomic assignment.

### 4.1 Directory setup of temporary files

**Technical Notes:** Snakemake starts by ensuring the existence of necessary directories. Most notably, the temporary bootstrap directories.

**Rationale:** Bootstrapping is sometimes done iteratively thousands of times, so making this a temporary directory helps manage space.

**Notes:**

### 4.2 Sample Identification

**Technical Notes:** The workflow identifies sample names by parsing filenames in the raw reads directory.

**Rationale:** Inclusion of these in the script allows the user to flexibly configure the naming convention and the directory in which they want to bootstrap.

**Notes:** Presently it looks for `_R1.fastq.gz` and its associated pair, `_R2.fastq.gz` in the `raw_reads` directory.

### 4.3 Bootstrapping

**Technical Notes:** This executes `bootstrap_reads.py` in the `scripts` directory to start bootstrapping the paired-end reads and outputs them in the temporary folders I mentioned earlier.

**Rationale:** Moving bootstrapping logic into a Python script leverages its ability to create randomizations from its libraries. Also it allows us to define a `seed` so it is reproducible (if you want to reproduce) the results anyway.

**Note:** The number of bootstraps and the fraction of samples you want to retain can be controlled in the `config.yaml` file in the configuration folder.

**On a personal note:** Before writing this part of the document (Sep 19, 2024, 3:00 AM), I decided to change the bootstrapping rule. It used to rely on a simplified approximation. The probability of not being selected after  $N$  independent draws from a sample of size  $N$  is given by:

$$P = \left(1 - \frac{1}{N}\right)^N$$

This is a well-known mathematical equation describing the probability of not selecting a sample at least once in  $N$  draws. As  $N \rightarrow \infty$ , this probability approaches:

$$\frac{1}{e}$$

Previously, I used the probability of a sample *being selected*, which is:

$$1 - \frac{1}{e}$$

This was used as an approximation for bootstrapping by shuffling and adjusting the sample size. However, the updated script now performs **actual bootstrapping**, sampling with replacement, which is a more accurate statistical method for resampling.

#### 4.4 Run `kraken_pipeline.bash`

**Technical Notes:** This Shellsript (or bash file) is used to automate the processing of all files from the bootstrapping. It runs them under Kraken2 then Bracken to generate taxonomy profiles for all of them. It also creates a log file for each replicate to provide traceability and error checking, helping diagnose any issues with specific replicates or samples.

**Rationale:** Read why I'm bootstrapping from the textbox above. The reason why I also included Bracken and measurements diversity metrics in the analysis per sampling replicate is so we can analyze how the topology of diversity also changes - similiar to how we look at topology of phylogenetic trees. The final process consolidates all the diversity metrics (alpha diversity) and matrices (beta diversity) into a TSV file.

**Note:** The decision to use Snakemake and Shellscripts here is so that the bootstrapping comes first before the pipeline is introduced. Otherwise Snakemake will run the entire thing in parallel, taking up so much memory because it runs Kraken2 for every single sample instead of doing it in batches - which takes up so much unnecessary time.

## 5 Binning.smk

**Stage: To test on higher coverage data**

**Binning pipeline to create high quality (Metagenome Assembled Genomes) MAGs**

**Specifics:** This pipeline passes through multiple quality checks during binning of using a variety of tools (both wrappers and modules) including MetaWrap, DasTool, CheckM2, MagPurify etc.

### 5.1 Universal Configurations

**Technical Notes:** The pipeline allows the user to configure settings they want for the binning process. By default, the settings are Minimum Contig Length (2500bp), Completeness (50%), Contamination (10%).

**Rationale:** The default settings were curated by me and the reason I chose them is because of the following

1. Longer contigs tend to represent more complete genomic fragments. Setting a threshold of 2500bp ensures better assembly quality. Others prefer a lower threshold for more sensitivity like 2000 bp.
2. Completeness 50% and Contamination 10% are actually based from a standard called the MIMAG standards.

**Notes:** Making configurations universal this way creates consistency across the script, i.e. when specifically asked by a specific tool, this returns a universal parameter.

**Notes:** Additionally, the user can specify the memory usage and number of threads they want to allocate per tool as well as other tool-specific parameters at the top of the script for ease of use. *I plan to add this to the configuration file soon once I have tested the file to be working at higher coverage - since you can't make high quality bins with low read counts - and my PC can't practically handle that sorry.*

### 5.2 FastUniq (Deduplication)

**Technical Notes:** FastUniq used to remove duplicate reads.

**Rationale:** Since we are focused on creating MAGs or genomes based on populations of genomes, removing duplicates is less risky during binning and is thus included. It also allows us to completely remove amplification bias from PCR reactions. Moreover, deduplication reduces redundancy and thereby memory usage downstream.

### 5.3 Seqtk (FASTA Conversion)

**Technical Notes:** Seqtk converts FASTQ to FASTA.

**Rationale:** This conversion prepares sequences for tools that require FASTA inputs, such as CD-HIT-EST.

**Notes:** I used to include a decompress-then-compress mechanism in the script to minimize memory storage but according to my calculations from the sequencing facility quotations, re-compressing may actually be more costly when done throughout the pipeline. Hence, it should be more cost-efficient to start compressing files once the bins are done.



## 5.4 CD-HIT-EST (Clustering) at Identity: 90%

**Technical Notes:** CD-HIT-EST clusters similar sequences at 90% identity.

**Rationale:** Clustering reduces redundancy in the contig data while maintaining closely related sequences.

**Notes:** I chose 90% ID because that's what I often see in published journals that is all. Perhaps, the 90% identity threshold balances removing duplicates while preserving diversity and that higher thresholds would result in fewer clusters but might oversimplify the data. *Fine, I'll make it my homework assignment why this specific threshold is used (I HAVE TO KNOW).*

## Bin Refinement

---

## 5.5 MetaWRAP Binning and Reassembly

**Technical Notes:** MetaWRAP is what is known as a wrapper program - meaning it makes use of other tools as its modules. For the binning process in particular it uses 3: MetaBAT2, MaxBin2, and CONCOCT. Each binning process goes through internal quality control checks and the one with the best bin-qualities are selected. It also has a reassembly feature wherein it reassembles the contigs again to try and further refine the bins.

**Rationale:** Using 3 bidders in parallel and choosing the best bins, quality checking, and then reassembling (not-so-good) bins make the binning process very robust, creating very refined bins *not sponsored by the way, talking as a fellow researcher.*

**Notes:** No moving forward, the process of further refinement of bins seems redundant. But do note that I have checked and validated that the process used in refining and checking by the tools used here cover different metrics - and therefore can be seen as parallel processes.

## 5.6 DAS Tool (Bin Refinement)

**Technical Notes:** DAS\_Tool is very similar to MetaWrap in that they both choose the best bins from a pool of bins from different bidders (in this pipeline DAS Tool is designed to **ALSO** use information from MetaBAT2, MaxBin2, and CONCOCT outputs to improve binning)

**Rationale:** However, as rationale for including it, is that it focuses more on single-copy gene (SCG) analysis. In contrast, in MetaWrap, bins are evaluated using completeness and contamination thresholds.

**Notes:** Notably, in the checking phase of this pipeline we will not be using DAS\_Tool for SCG analysis. It is optimized for refining bins not quality checking bins. Instead we will use a more updated and optimized software for the latter called BUSCO.

## 5.7 MAGpurify (Contamination Removal)

**Technical Notes:** MAGpurify is also a bin refiner, in a sense that it uses several modules to detect and **prune** contamination in genome bins. It also uses other metrics to define bin quality specifically it looks for differences in

### 1. Phylo-markers,

72 2. **Clade-markers**,

73 3. **Tetranucleotide-frequencies**,

74 4. **GC-content**,

75 5. and then removes **known-contaminants** from it's manually curated database (created back in  
76 2013)

77 **Rationale:** This step improves genome quality by removing low-confidence contigs or contamination  
78 from other taxa. Each module targets different contamination types (phylogenetic, clade, etc.).

79 **Notes:** Similar to DAS\_tool, MAGPurify is relatively old (in the bioinformatics world where new tools  
80 are being published every day). So in checking our bins we will be using more recently updated tools.

## 83 Quality checking

---

### 85 5.8 MetaQUAST (Assembly Quality Assessment)

86 **Technical Notes:** MetaQUAST assesses the quality of genome assemblies via the following:

- 87 1. N50 and L50 to determine contiguity
- 88 2. Number of contigs to determine fragmentation
- 89 3. GC content - since a single MAG should have a constant GC content across its entirety (usually)
- 90 4. Alignment to a reference sequence
- 91 **Additionally, it also detects** other metrics using modular tools
- 92 5. structural variations (requires GRIDSS)
- 93 6. presence of rRNA (requires SILVA)
- 94 7. Conserved gene sets (requires BUSCO)

95 **Rationale:** This step quantifies the completeness and accuracy of the assembled genomes, and is updated  
96 frequently.

97 **Notes:** Using reference genomes improves the accuracy of the assessment, but it's optional if references  
98 are unavailable.

#### A Personal Note

**Personal Note:** As of this writing, BUSCO has updated beyond the version required by QUAST (BUSCO od9). Unfortunately, this version is not available in the archives (you'll encounter a 404 error). Likewise, SILVA and GRIDSS frequently update. I recommend downloading each separately and manually linking their databases to avoid potential issues with QUAST's download management. *Good luck and have fun!*

## 5.9 dRep (Dereplication)

**Technical Notes:** dRep dereplicates genomes by clustering them based on similarity.

**Rationale:** Dereplication reduces redundancy in the assembled genome data, ensuring unique genome representations. Basically CD-HIT but for whole genomes.

**Notes:** FastANI is utilized for quick, precise clustering, and is part of the dRep package. It defaults to a 95% ANI (Average Nucleotide Identity) threshold, a common yet somewhat subjective metric used to determine microbial species boundaries. This is often sufficient for microbial genomes due to their high gene density, frequently organized in operons. However, it may not be as suitable for eukaryotic genomes, which are laden with repetitive elements.

### A Personal Note

**Personal Note:** I cannot find a newer version of either dRep or FastANI (both dating back to 2013, basically ancient in bioinformatics terms). There's a newer version called pyani, which is available in Bioconda, that might be a good replacement. However, I still need to reverse-engineer its source code to fully understand how it operates. Might be worth trying!

## 5.10 CheckM2

**Technical Notes:** CheckM2 is used to predict genome completeness and contamination using low-memory mode (essential for resource-limited systems like mine; remember to adjust this on HPC systems).

**Rationale:** CheckM2 is an updated version of CheckM, but many tools in this pipeline haven't been updated to recognize it. I haven't tested whether aliasing CheckM2 as CheckM would work, so it's added here as a final step to ensure the results meet MIMAG standards for completeness and contamination.

## 6 krakenpipeline.bash

**Stage: Done**

**The entire basic taxo-metagenomic pipeline with the usual tools**

**Specifics:** This pipeline passes through a loop between FastQC and Trimmomatic before progressing to Kraken2 then Bracken then calculate\_diversity.py.

7

## **Project Side Scripts**

6

<sup>1</sup> **1**

3

# Project Main Scripts

2

<sup>1</sup> **1**

## **Daily entries**



## 1 September 20, 2024

### To-do List

**Note:** This should've been yesterday's to-do list, but my landline was down, possibly due to the weather—it's been raining all day.

- ☐ Call for a group meeting regarding logistics (or when they will be available via Zoom call)
  - ☐ Discuss the issue about VMMC review fee
  - ☐ Raise concerns about data storage and management
  - ☐ Inquire if there are scripts I can develop to streamline the bureaucratic process
- ☐ Contact engineering/sanitation departments via landline
  - Talk about possible sampling sites and that we will present our authorization upon arrival
- ☐ VMMC
  - \* Tell them that we have a go-signal from the admin, but we have yet to pay for the fee
    - which can be held off for later
- ☐ Mary Johnston
  - \*
- ☐ ManilaMed
  - \* Tell them we already have a go signal from Ma'am Eula
- ☐ St. Lukes
  - \* Inquire where Engr. Valenzuela is there
  - \* Inquire the status of their renovation and if they are available for sampling this October already as talked about last year

### Completed Tasks

- ✓ raw API results now readable stored in TSV file
- ✓ Created symlinks in private (repo) to .git folders in public repos to enable tracking of Git and Github activities
- ✓ Resolved residual time-tracking activities within the local computer

### Scripts Worked On

- ✓ checkingAPI.sh (now updated to make the raw API results readable)

### Issues Encountered

- 

### Need to Troubleshoot

-

## 2 September 19, 2024

### To-do List

- ☐ Try to complete LaTeX documentation of scripts
  - ✓ Project 4 pipelines
    - ✓ ARG-MGE.smk
    - ✓ metagenomics\_pipeline.smk
    - ✓ Binning\_.smk
  - ➡ Project Main pipelines
  - ➡ Project Side pipelines
- ✓ Counter-check Project4 scripts with documentation to see redundancies

### Completed Tasks

- ✓ Deleted redundancies in Project 4 scripts
  - FastQC\_check.py
  - fastp.bash
  - README.md (since this document exists and is more comprehensive)
  - fastp metrics from summary\_stat.bash
- ✓ Decided to move to Project Side K-mer\_diversity.smk as it is not essential (yet).
- ✓ Finally resolved an issue with wakatime not integrating properly with TeXStudio
- ✓ I realize now that the entry dates were one-day behind.

### Scripts Worked On

- calculate\_contig\_quality.py now updated to calculate coverage statistics.

### Issues Encountered

- 

### Need to Troubleshoot

-

### 3 September 18, 2024

#### To-do List

- ☞ Call for a group meeting regarding logistics
  - ✓ Offer to use landline to minimize on-site visits and reduce transportation costs; as point-persons have not been replying via email as quickly lately
  - ✓ Discuss PGC calculations and talk about allocatable budget
  - ✓ On-site orientation with PGC engineering department (OETS) regarding sampling sites **taken care of by James and Roch**
- ☞ Raise concerns about data storage and management
- ☞ Inquire if there are scripts I can develop to streamline the bureaucratic process

#### Completed Tasks

- Setup wakatime in VS Code and Ubuntu to track my coding productivity
- Helped in receiving and moving the autoclave and fridge needed by the Program

#### Scripts Worked On

- `rawAPIformatter.sh`
- ✓ `checkingAPI.sh`
- `wakatime.sh`

#### Issues Encountered

- ✓ Connection blockage from wakatime API
- ☞ raw API results is unreadable

#### Need to Troubleshoot

- conky display of wakatime API for desktop

## 4 September 17, 2024

### To-do List

- ☞ Call for a group meeting regarding logistics
  - ☞ Offer to use landline to minimize on-site visits and reduce transportation costs; as point-persons have not been replying via email as quickly lately
  - ☞ Discuss PGC calculations and talk about allocatable budget
  - ☞ Raise concerns about data storage and management
  - ☞ Inquire if there are scripts I can develop to streamline the bureaucratic process

### Completed Tasks

- ✓ Updated GitHub repositories; created new repositories: `Documentation` and `Confidential`
- ✓ Created a new bash script to track progress across all repositories
- ✓ Set up this research diary to document daily progress and tasks
- ✓ Python script that calculates contig quality metrics L50, N50, but also L90, N90, GC skew etc.
- ✓ Python script that detects and calculates changes in coverage over contigs for read mapping
- ✓ Typesetting: explaining the process and rationale inside the ARG-MGE.smk pipeline
- ✓ Integration of the two scripts to the ARG-MGE pipeline

### Scripts Worked On

- `Project4/ARG_MGE.smk` - updated
- ✓ `Project4/calculate_contig_quality.py`
- ✓ `Project4/plot_and_detect_intermediate_coverage.py`

### Issues Encountered

- ✓ Could not access GitHub, needed to reset the SSH keys and update my Git histories for version control

### Need to Troubleshoot

- None identified at this time