Tutorial 7: Metagenome De Novo Assembly and Binning - KBase

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Background

What is Metagenome De Novo Assembly?

- Uses small DNA pieces called reads from environmental samples to rebuild the original genome sequence
- Assembles the genome using only the sequencing data, without a reference genome
- Outcome of this process is called 'contigs,' which are long pieces of DNA sequences
- Contigs are the reconstructed segments of the original genome
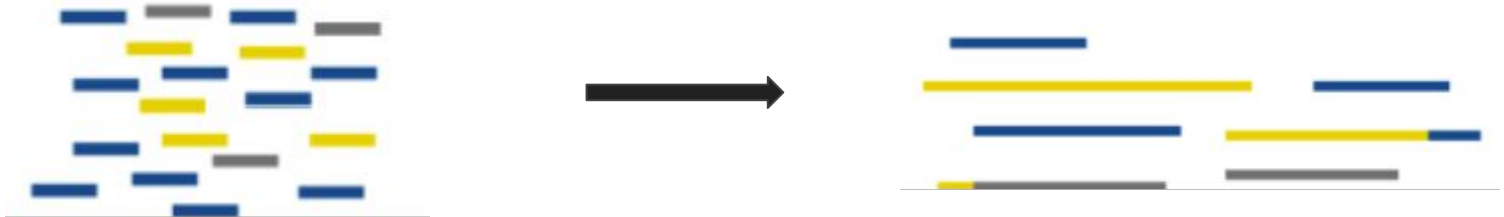
# What is Binning?

- The process of grouping contigs into bins
- Bin represents a potential genome from different organisms in the sample.
- This is done by analyzing patterns in the DNA sequences and their abundance
- Helps in organizing the complex data from environmental samples, making it easier to study the microbial diversity and the genetic makeup of individual species within the sample.
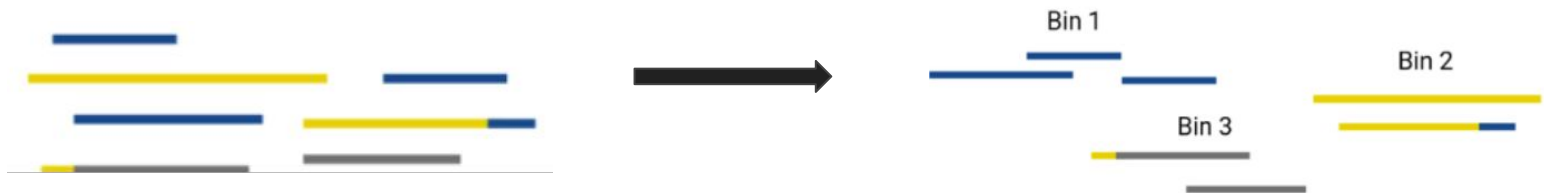
What is KBase?

- Web platform for performing De Novo Assembly and other bioinformatics tasks
- Provides tools and workflows to analyze environmental DNA samples
- Users can rebuild genomes from sequencing data using KBase's resources
- Makes it easier to study microbial communities and their functions
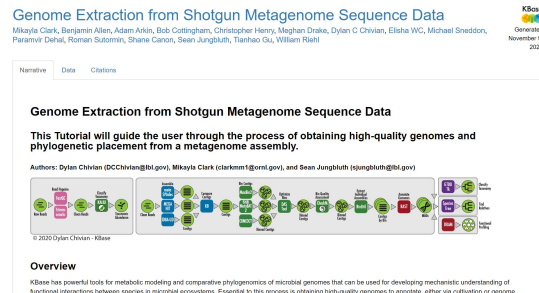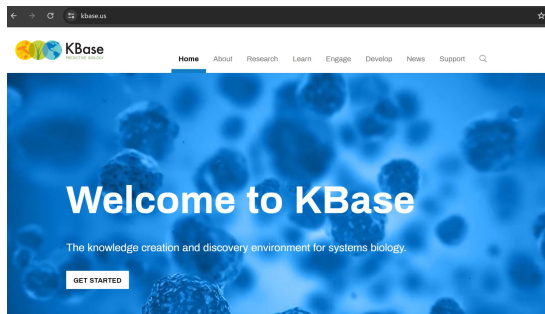
Workflows

- Read Hygiene
- Classify Taxonomy
- Assemble
- Compare Contigs
- Bin Contigs
- Optimize Binned Contigs by Consensus
- Bin Quality Assessment
- Extract Individual Assemblies
- Annotate Genomes
- Taxonomic Classification of MAGs

# Read Hygiene

- Paired-end reads in FASTQ format are imported
  - Paired-end reads?
    - Refer to sequences where both ends of a DNA or RNA fragment are sequenced.
- Improves the quality using FastQC and Trimmomatic
  - FastQC?
    - Provides a comprehensive analysis and visualization of the quality of data
  - Trimmomatic?
    - Removes adapter sequences, trimming low-quality sequence regions

Classify Taxonomy

- Kaiju is used to predict microbial composition based on protein similarity, and from this, a species tree is generated
  - Kaiju?
    - Tool designed for analyzing the taxonomic composition of microbial communities
    - Identifies prokaryotes, viruses, and eukaryotes quickly and accurately
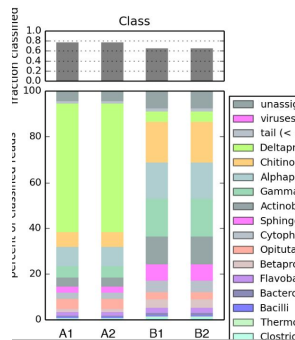
Assemble

- Reads are assembled to create scaffolding for the entire genome
- Multiple assembly apps are run to compare their results
    - using metaSPAdes, MEGAHIT, IDBA-UD
        - All uses De Bruijn graph approach
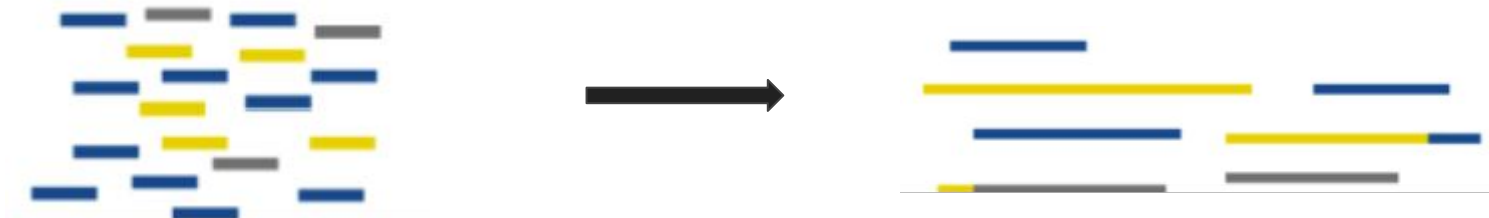
De Bruijn graph approach

1. **k-mer Splitting**

Breaking down DNA sequences (reads) into all possible subsequences of length k, known as k-mers

2. **Graph Construction**

Connecting nodes in the graph based on the continuity of k-mers

3. **Path Tracing and Assembly**

Finding paths from the start node to the end node in the De Bruijn graph and following these paths to assemble the sequence

Assemble

- Reads are assembled to create scaffolding for the entire genome
- Multiple assembly apps are run to compare their results
    - metaSPAdes vs. MEGAHIT vs. IDBA-UD
        - MEGAHIT is optimized for memory efficiency
        - IDBA-UD employs iterative k-mer size increases
        - Generally, MEGAHIT is faster and sound for large data but slightly inaccurate
        - metaSPAdes and IDBA-UD are more accurate but computationally expensive

Compare Contigs

- Sets of contigs generated using assemblers are compared with the quality metrics N50, L50
  - N50 refers to the minimum length of contigs, required to cover more than half of the total genome length
  - A high N50 value is desirable because it implies that the assembly has successfully captured longer regions of the genome without breaking them into smaller pieces
  - L50 signifies the minimum number of contigs required to cover half of the total genome length
  - A lower L50 value is preferred because it indicates that a significant portion of the genome can be represented with fewer, longer contigs
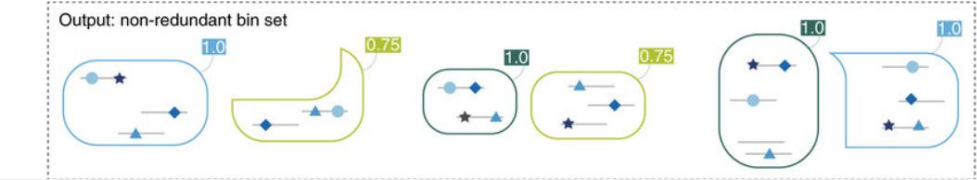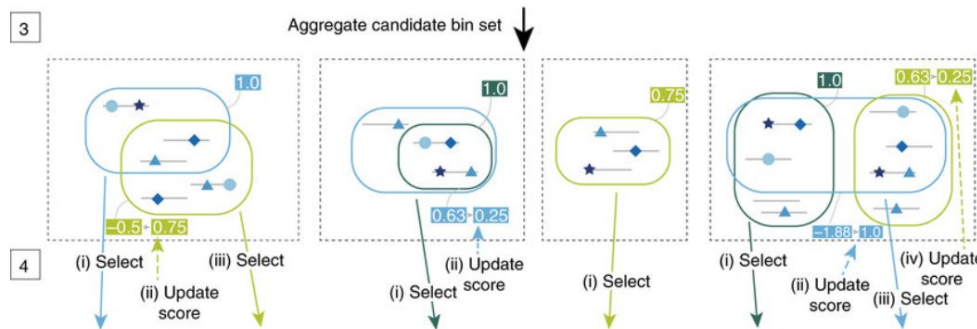
Bin Contigs

- Assembled contigs are clustered into bins, representing a hypothetical genome
    - MaxBin2
        - Employes k-mer frequencies and differential coverage
    - MetaBAT2
        - Utilizes tetranucleotide frequency and contig coverage patterns to group contigs
    - CONCOAT
        - Combines both the coverage and the nucleotide composition to cluster contigs

	c1	c2	...
AAAA	5	12	...
AAAT	0	6	...
AAAC	14	7	...
AAAG	3	8	...
...	...	...	...

Bin 1

Bin 2

Bin 3
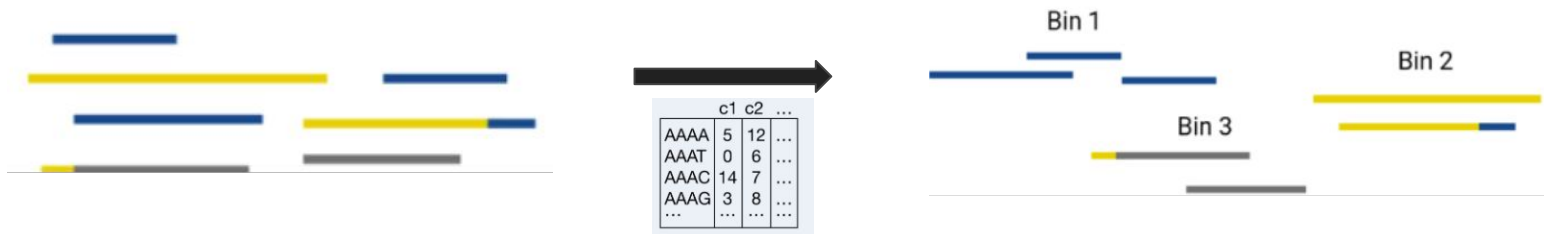
Optimize Binned Contigs by Consensus

- The quality of binned contigs is improved through consensus assignments using DAS-Tool, employing multiple methods
    - DAS-Tool?
        - Differential Abundance Score Tool is a comprehensive and advanced tool designed for integrating and refining the results of microbial genome reconstructions from metagenomic data

1 Input: 1 assembly [══════] + *n* bin sets [⬭, ⬭, ... , ⬭]

Bin set 1   Bin set 2   Bin set *n*

2 Predict single-copy genes [★, ◆, ▲, ... , ●] and score bins

−1.88   1.0   −0.5   0.63
1.0   0.63   1.0   0.75

Aggregate candidate bin set

3   1.0   1.0   0.75   1.0   0.63−0.25
−0.5−0.75   0.63−0.25   −1.88−1.0

4 (i) Select   (iii) Select   (i) Select   (ii) Update score   (i) Select   (i) Select   (ii) Update score   (iv) Update score   (iii) Select
(ii) Update score

Output: non-redundant bin set
1.0   0.75   1.0   0.75   1.0   1.0
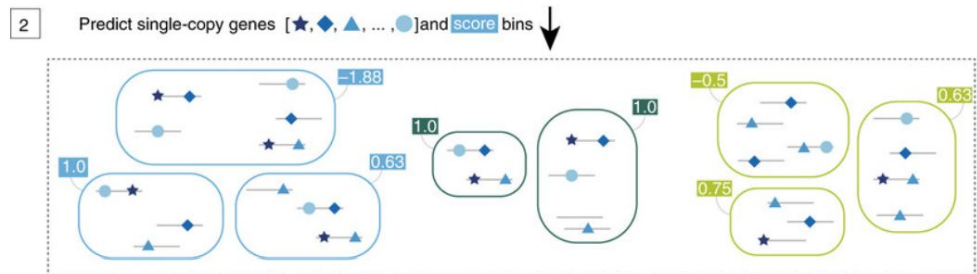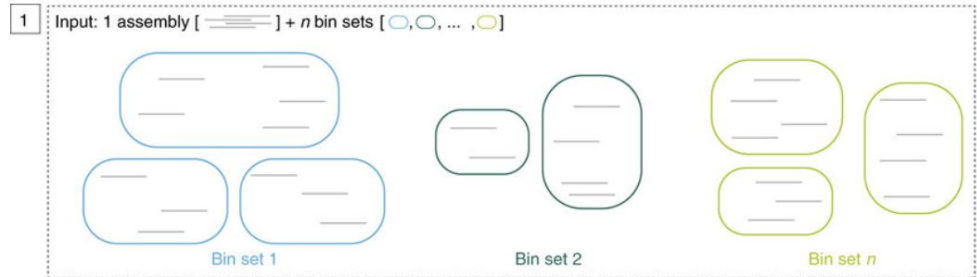
Bin Quality Assessment

- The quality of bins is evaluated using CheckM, and high-quality bins are filtered out
  - CheckM?
    - CheckM is a software tool designed for assessing the quality of microbial genomes recovered from isolates, single-cell sequencing, or metagenomic assemblies.

Extract Individual Assemblies

- High-quality bins are extracted as Assembly objects to be utilized in downstream applications.
    - BinUtil?
        - BinUtil is a tool designed for extracting high-quality bins as independent Assembly objects from metagenomic datasets
        - It facilitates the reconstruction and analysis of individual microbial genomes within complex communities
        - It plays a crucial role in downstream bioinformatics analyses, enabling detailed study of microbial diversity and function

Annotate Genomes

- High-quality bins are converted into annotated genomes using RASTtk
  - RASTtk?
    - RASTtk (Rapid Annotation using Subsystem Technology toolkit) is a software tool used for annotating bacterial and archaeal genomes
    - It uses a combination of similarity-based and model-based methods to predict genes and their functions

Taxonomic Classification of MAGs

- GTDB-Tk Classify is used to provide a phylogenetic classification for MAGs
  - GTDB-Tk Classify?
    - GTDB-Tk (Genome Taxonomy Database Toolkit) classifies microbial genomes based on the Genome Taxonomy Database
    - GTDB is a updated database that provides a standardized taxonomy for bacteria and archaea, which is based on genome phylogeny
  - MAG?
    - A Metagenome-Assembled Genome (MAG) is a collection of contigs derived from metagenomic datasets that are assembled and binned together
    - It represents the genome of a single microbial species or strain present within a complex microbial community

Kbase Usage

Kbase Narratives

The narrative is the workspace for kbase.

Here you can import data, utilize all the functions of kbase through apps, and view outputs.

Import and select from available data

Select apps and functions

Narrative interface: here you can see the apps you have selected and their outputs

Data Pre-processing

As a quick fix, we replace the incorrect nucleotides with a random ATCG, and reduce the corresponding quality scores to zero

Due to invalid characters preventing the upload of the fastq metagenome sample, we needed a data preprocessing tool.

```
81    ERROR on Line 93258: Invalid character ('S') in base sequence.
82    ERROR on Line 94738: Invalid character ('S') in base sequence.
83    ERROR on Line 105246: Invalid character ('W') in base sequence.
84    ERROR on Line 134718: Invalid character ('Y') in base sequence.
85    ERROR on Line 144210: Invalid character ('K') in base sequence.
86    ERROR on Line 168330: Invalid character ('R') in base sequence.
87    ERROR on Line 168330: Invalid character ('R') in base sequence.
```

```python
def remove_incorrect(fasta_dir):
    '''
    replace with random but with a quality of zero
    '''
    new_records = []
    bases = ['A', 'T', 'C', 'G']
    for record in SeqIO.parse(fasta_dir, "fastq"):
        #remove incorrect bases
        sequence = str(record.seq)
        index = re.search('[^ATCG]', sequence)

        if index is None:
            new_records.append(record)
            continue

        new_sequence = re.sub('[^ATCG]', random.choice(bases), sequence)
        record.seq = Seq(new_sequence)

        #remove annotations
        letter_annotations = record.letter_annotations
        record.letter_annotations = {}
        letter_annotations['phred_quality'][index.start()] = 0
        new_letter_annotations = {'phred_quality': letter_annotations['phred_quality']}
        record.letter_annotations = new_letter_annotations

        new_records.append(record)

    with open("incorrect_removed_" + fasta_dir, 'w') as output_handle:
        SeqIO.write(new_records, output_handle, "fastq")
print(len(new_records))
```

Importing data

For small files (>5gb), it is possible to upload data directly through the kbase interface.

Otherwise, it is necessary to use Globus

Our filetype is fastq interleaved

Compressed files can be uploaded, but they need to be uncompressed before importing to the narrative

Importing the selected data will automatically add this app which verifies and imports the file as a data object

Assess Read Quality

FastQC Assess Read Quality with FastQC - v0.12.1
A quality control application for high throughput sequence data.

Summary

✅ Basic Statistics

✅ Per base sequence quality

✅ Per sequence quality scores

✅ Per base sequence content

❌ Per sequence GC content

✅ Per base N content

✅ Sequence Length Distribution

✅ Sequence Duplication Levels

✅ Overrepresented sequences

✅ Adapter Content

❌ Per sequence GC content

GC distribution over all sequences

GC count per read
Theoretical Distribution

1200000

1000000

800000

600000

400000

200000

0

0 2 4 6 8 11 15 19 23 27 31 35 39 43 47 51 55 59 63 67 71 75 79 83 87 91 95 99

Mean GC content (%)

Position in read (bp)

astq

Classify Taxonomy - Kaiju

Assembly results

BEST ◻◻◻◻◻◻ WORST

ASSEMBLY	LONGEST CONTIG (bp)	Nx (Lx)		LENGTH (bp)	NUM CONTIGS	SUM LENGTH (bp)	Contig Length Histogram (1bp <= len < 10Kbp)	Contig Length Histogram (10Kbp <= len < 100Kbp)	Contig Length Histogram (len >= 100Kbp)
IDBA.contigs	1362924	N50:	73077	>= 10^6	2	2591304			
		L50:	(183)	>= 10^5	131	32044837			
		N75:	16441	>= 10^4	1205	60309663			
		L75:	(749)	>= 10^3	4050	72691075			
		N90:	5767	>= 500	4050	72691075			
		L90:	(1878)	>= 1	4050	72691075			
metaSPAdes.Assembly	1362255	N50:	89948	>= 10^6	2	2590385			
		L50:	(184)	>= 10^5	160	39747091			
		N75:	27588	>= 10^4	1276	73840281			
		L75:	(615)	>= 10^3	3620	83943041			
		N90:	8142	>= 500	3620	83943041			
		L90:	(1467)	>= 1	3620	83943041			
MEGAHIT.assembly	1362697	N50:	63929	>= 10^6	2	2591041			
		L50:	(222)	>= 10^5	140	32595372			
		N75:	9410	>= 10^4	1052	58036617			
		L75:	(1125)	>= 10^3	6441	78320115			
		N90:	3360	>= 500	6441	78320115			
		L90:	(3352)	>= 1	6441	78320115			

Bin Contigs

MaxBin2 **Bin Contigs using MaxBin2 - v2.2.4**
Group assembled metagenomic contigs into lineages (Bins) using depth-of-coverage, nucleotide composition, and marker genes.

CONCOCT **Bin Contigs using CONCOCT - v1.1**
Group assembled metagenomic contigs into lineages (Bins) using depth-of-coverage and nucleotide composition

JGI **MetaBAT2 Contig Binning - v1.7**
Bin metagenomic contigs

Overview

Bins: 21

Input Contigs: 3620

Binned Contigs: 3533 (97.6%)

Unbinned Contigs: 87 (2.4%)

Contigs Too Short: 0 (0.0%)

Summed Length of Binned Contigs: 83526449 (99.5%)

Summed Length of Unbinned Contigs: 416592 (0.5%)

Summed Length of Short Contigs: 0 (0.0%)

MaxBin

Overview

Binned contigs: 3081

Input contigs: 3620

Number of bins: 23

CONCOT

Overview   Summary

Binned contigs: 2602

Input contigs: 3620

Number of bins: 25

MetaBAT2

Refine Bins by Consensus
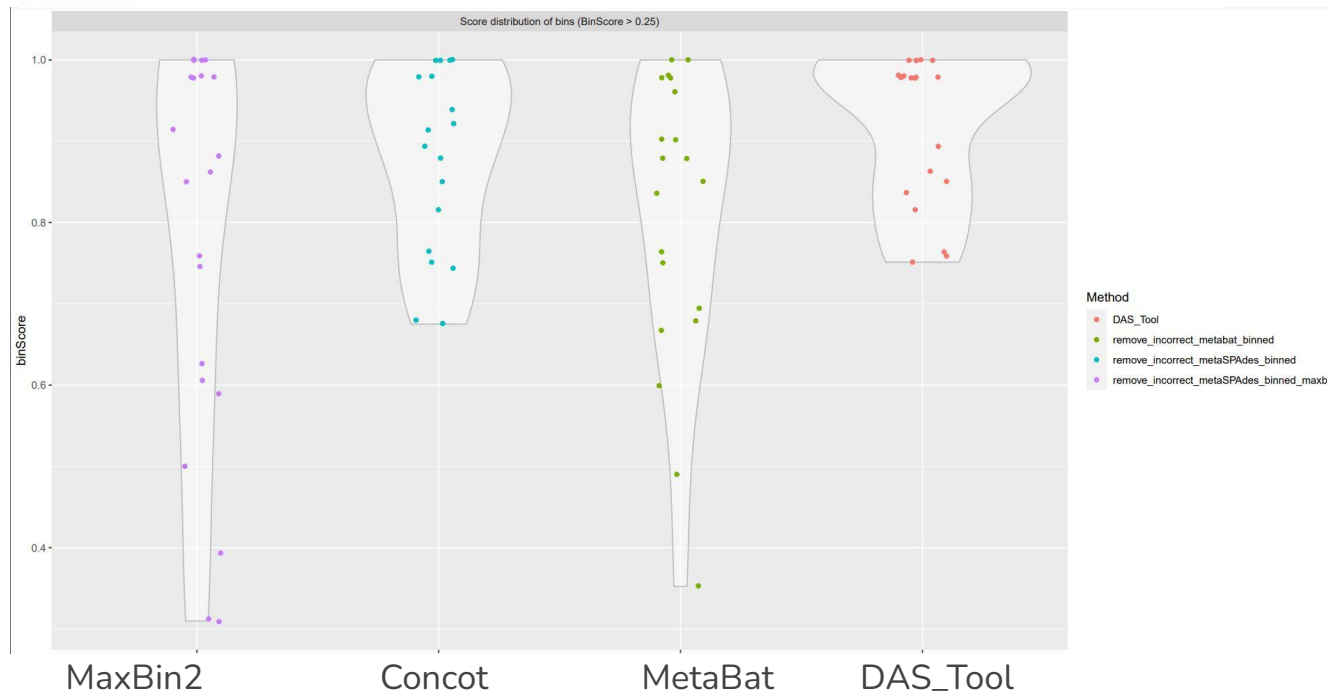

Optimize Bacterial or Archaeal Binned Contigs using DAS Tool - v1.1.2
Optimize bacterial or archaeal genome bins using a dereplication, aggregation and scoring strategy

Score distribution of bins (BinScore > 0.25)

Method
- DAS_Tool
- remove_incorrect_metabat_binned
- remove_incorrect_metaSPAdes_binned
- remove_incorrect_metaSPAdes_binned_maxb

MaxBin2 Concot MetaBat DAS_Tool

Bin Quality Assessment

CheckM — Assess Genome Quality with CheckM - v1.0.18
Runs the CheckM lineage workflow to assess the genome quality of isolates

CheckM — Filter Bins by Quality with CheckM - v1.0.18
Runs the CheckM lineage workflow to assess the genome quality of isolates,

BinUtil — Extract Bins as Assemblies from BinnedContigs - v1.0.2
Extract a bin as an Assembly from a BinnedContig dataset

From an initial 19 bins, we assess the quality and extract 12 for downstream use

The removed bins are marked in red. They do not satisfy the >95% completeness and <2% contamination threshold

Bin Name	Marker Lineage	# Genomes	# Markers	# Marker Sets	0	1	2	3	4	5+	Completeness	Contamination
bin.001	f__Rhodobacteraceae	84	568	330	2	562	4	0	0	0	99.6	0.5
bin.002	f__Rhodobacteraceae	84	568	330	6	560	2	0	0	0	98.56	0.3
bin.003	c__Betaproteobacteria	323	387	234	2	383	2	0	0	0	99.15	0.43
bin.004	o__Actinomycetales	488	309	185	0	308	1	0	0	0	100.0	0.18
bin.005	o__Burkholderiales	193	427	214	82	337	8	0	0	0	82.63	2.76
bin.006	o__Clostridiales	304	250	143	7	243	0	0	0	0	96.5	0.0
bin.007	p__Bacteroidetes	364	303	203	71	229	3	0	0	0	73.48	0.54
bin.008	c__Gammaproteobacteria	67	481	276	7	468	6	0	0	0	97.96	1.14
bin.009	c__Deltaproteobacteria	83	247	155	3	244	0	0	0	0	98.06	0.0
bin.010	o__Actinomycetales	148	572	276	6	559	7	0	0	0	98.99	0.71
bin.011	c__Bacilli	750	273	152	7	266	0	0	0	0	96.85	0.0
bin.012	k__Bacteria	924	151	101	2	130	19	0	0	0	98.68	10.89
bin.013	p__Bacteroidetes	364	302	203	14	285	3	0	0	0	98.81	0.99
bin.014	c__Alphaproteobacteria	468	388	250	39	345	4	0	0	0	91.86	1.13
bin.015	p__Firmicutes	100	295	158	1	289	5	0	0	0	99.37	2.22
bin.016	k__Bacteria	3167	126	75	30	96	0	0	0	0	81.25	0.0
bin.017	o__Pseudomonadales	185	813	308	5	805	3	0	0	0	99.02	0.32
bin.018	f__Xanthomonadaceae	55	659	290	152	493	14	0	0	0	79.54	2.72
bin.019	o__Burkholderiales	107	574	251	0	566	8	0	0	0	100.0	1.49
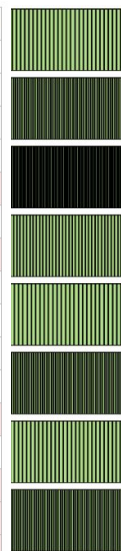
Created Object Name

Bin.001.fasta_assembly

Bin.002.fasta_assembly

Bin.003.fasta_assembly

Bin.004.fasta_assembly

Bin.006.fasta_assembly

Bin.008.fasta_assembly

Bin.009.fasta_assembly

Bin.010.fasta_assembly

Bin.011.fasta_assembly

Bin.013.fasta_assembly

Bin.017.fasta_assembly

Bin.019.fasta_assembly

extracted_bins.AssemblySet

bin.013
bin.010
bin.017
bin.003
bin.015
bin.001
bin.004
bin.019

S

Contamina

3   4

Annotate Metagenome

RAST Annotate Multiple Microbial Assemblies with RASTtk - v1.073
Annotate bacterial or archaeal assemblies and/or assembly sets using RASTtk (Rapid Annotations using Subsystems Technology toolkit).

Created Object Name

annotated_metagenome

Bin.001.fasta_assembly.RAST

Bin.002.fasta_assembly.RAST

Bin.003.fasta_assembly.RAST

Bin.004.fasta_assembly.RAST

Bin.006.fasta_assembly.RAST

Bin.008.fasta_assembly.RAST

Bin.009.fasta_assembly.RAST

Bin.010.fasta_assembly.RAST

Bin.011.fasta_assembly.RAST

Bin.013.fasta_assembly.RAST

Bin.017.fasta_assembly.RAST

Bin.019.fasta_assembly.RAST

```
The RAST algorithm was applied to annotating a genome sequence comprised of 141 contigs containing 4517581 nucleotides.
No initial gene calls were provided.
Standard features were called using: glimmer3; prodigal.
A scan was conducted for the following additional feature types: rRNA; tRNA; selenoproteins; pyrrolysoproteins; repeat regions; crispr.
The genome features were functionally annotated using the following algorithm(s): Kmers V2; Kmers V1; protein similarity.
In addition to the remaining original 0 coding features and 0 non-coding features, 4592 new features were called, of which 163 are non-coding.
Output genome has the following feature types:
        Coding gene                     4429
        Non-coding crispr_array            1
        Non-coding crispr_repeat          20
        Non-coding crispr_spacer          19
        Non-coding repeat                 79
        Non-coding rna                    44
Overall, the genes have 0 distinct functions.
The genes include 0 genes with a SEED annotation ontology across 0 distinct SEED functions.
The number of distinct functions can exceed the number of genes because some genes have multiple functions.
Bin.003.fasta_assembly succeeded!
```

Taxonomic Classification

Classify Microbes with GTDB-Tk - v2.3.2
Obtain objective taxonomic assignments for bacterial and archaeal genomes based on the Genome Taxonomy Database (GTDB)

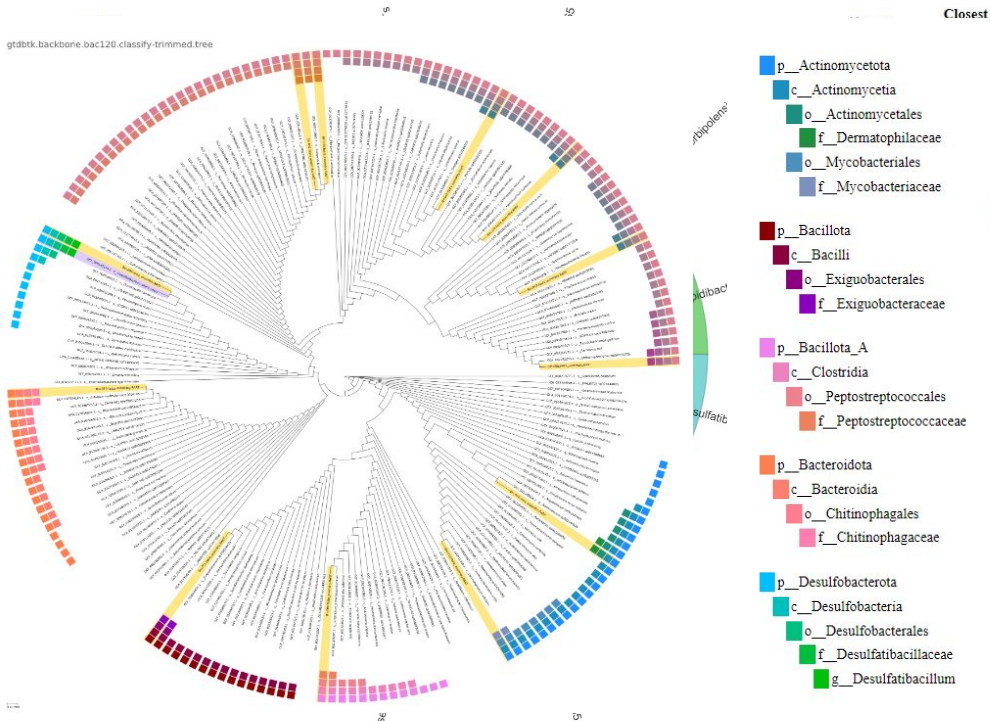| GTDB Trees | Krona Chart | Bacteria | Archaea | Bacteria Marker Summary | Archaea Marker Summary |

Find Relatives with Species Tree
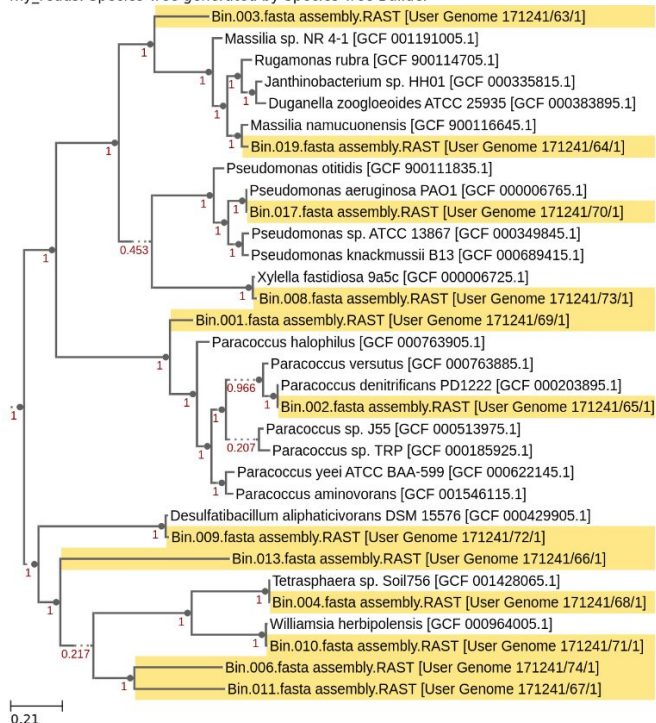

Species Tree
Insert Set of Genomes Into SpeciesTree - v2.2.0
Add a user-provided GenomeSet to a KBase SpeciesTree.

my_reads: Species Tree generated by Species Tree Builder

The highlighted genomes are the ones we have generated through our assembly and binning.

The current implementation has a tendency to overemphasize proximal genomes at the expense of phylogenetic diversity. This can be fixed through the following process:

Split bins into clades based on original tree

KB
Build GenomeSet - v1.7.6
Allows users to create a GenomeSet object.

Species Tree
Insert Set of Genomes Into SpeciesTree - v2.2.0
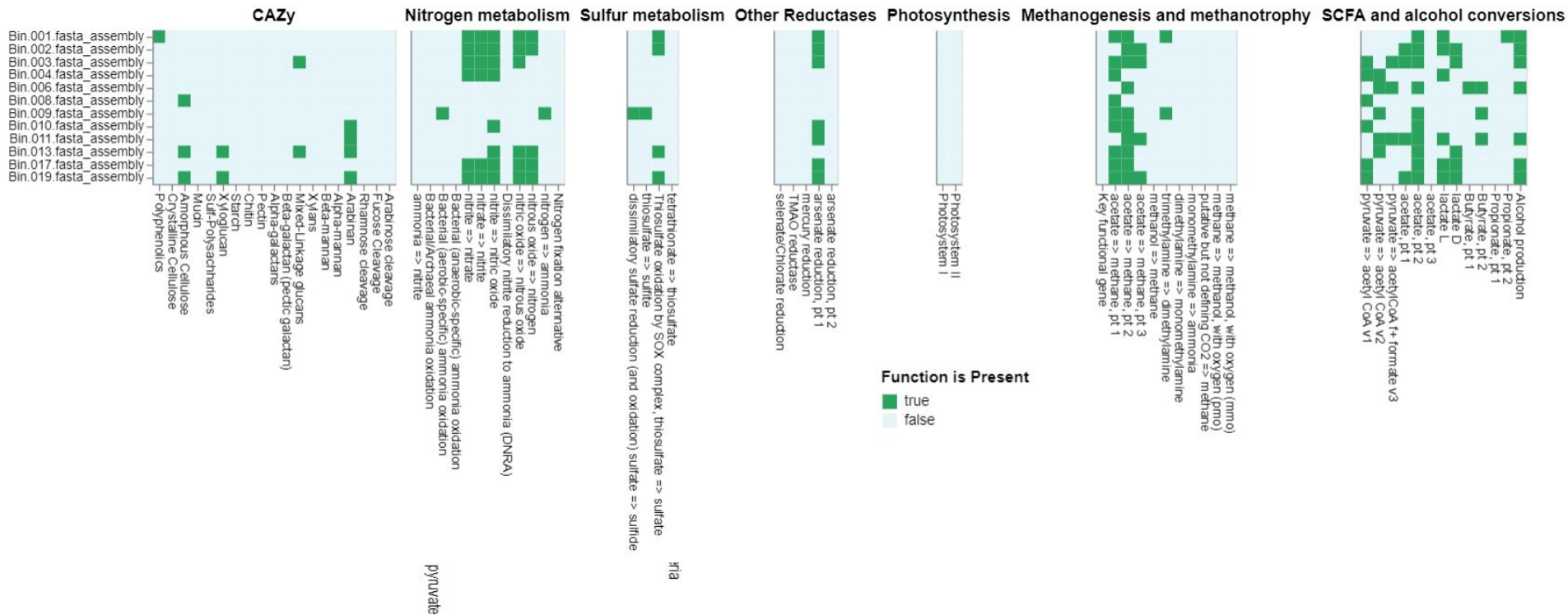Add a user-provided GenomeSet to a KBase SpeciesTree.

Repeat for each clade

Functional Classification of MAGs

Annotate and Distill Assemblies with DRAM
Annotate your assemblies, isolate genomes, or MAGs with DRAM and distill resulting annotations to create an interactive functional summary per

Reference

- Clark, M., Allen, B., Arkin, A., Cottingham, B., Henry, C., Drake, M., Chivian, D. C., Elisha, W. C., Sneddon, M., Dehal, P., Sutormin, R., Canon, S., Jungbluth, S., Gu, T., & Riehl, W. (2021, November 9). Genome Extraction from Shotgun Metagenome Sequence Data. KBase. https://kbase.us/n/33233/606/
- Sieber, C. M. K., Probst, A. J., Sharrar, A., Thomas, B. C., Hess, M., Tringe, S. G., & Banfield, J. F. (2018). Recovery of genomes from metagenomes via a dereplication, aggregation and scoring strategy. Nature Microbiology, 3(7). https://doi.org/10.1038/s41564-018-0171-1