# **Detecting variation**

Today, we'll be comparing reads from a 2018 Trunk River metagenome to a MAG (metagenome-assembled genome) from the same metagenome to examine variation in the population. Since these steps require some computing power and time, I've pre-computed the full analysis and pulled out a subset as an example.

## Step 0: Data acquisition

Make a new working directory in your group folder. Copy the starting data to the new folder and decompress it.

```
In [ ]: cp /datahaus/variants/mini_example/student_data.zip .
In [ ]: unzip student_data.zip
```

# Step 1: read mapping

The first step is to map the metagenome reads to the reference "genome" - the contigs from the metagenomic assembly. Though we're only interested in one MAG for now, I've mapped reads to contigs from all the MAGs.

Why map all the reads to all the contigs? Why not just map to the MAG we're interested in?

Knowing which reads map where, I've extracted a subset of reads that map to the MAG from bin28. We'll work with that for set for now.

First, build a bowtie index of the fasta file containing the bin contigs (near\_lem\_2018\_bin.28.fa). near\_lem\_bin28 will be the base name of the generated index files.

```
In [ ]: bowtie2-build near_lem_2018_bin.28.fa near_lem_bin28
```

Now map the metagenome reads ( <code>near\_lem\_2018\_R1\_bin28.fastq.gz</code> and <code>near\_lem\_2018\_R2\_bin28.fastq.gz</code> ) to this reference using bowtie2. This may take  $\sim$ 15 minutes; <code>screen</code> might be useful.

```
In [ ]: bowtie2 --local -p 4 -x near_lem_bin28 \
    -1 near_lem_2018_R1_bin28.fastq.gz -2 near_lem_2018_R2_bin28.fastq.gz \
    -S near_lem_bin28_vs_near_lem_bin28.sam --no-unal
```

Then convert the .sam file to a .bam file.

## Step 2: Gene Predictions

InStrain can give us a gene-level analysis of variation if we provide it a gene file. We'll use prodigal to predict genes. (If you tried prokka during the genome assembly & annotation workshop, prodigal is what prokka uses to predict genes).

```
In [ ]: prodigal -a near_lem_bin28.faa -d near_lem_bin28.fna \
  -f gff -i near_lem_2018_bin.28.fa -o near_lem_bin28.gff
```

#### Step 3: Find variable positions

Now, use inStrain's profile function to compare the reads to the reference genome (~15 mins for a single genome, longer for metagenomes).

```
In [ ]: inStrain profile -o instrain_near_lem_bin28 -p 4 \
   -g near_lem_bin28.fna \
   near_lem_bin28_vs_near_lem_bin28.bam near_lem_2018_bin.28.fa
```

InStrain's output will go to the directory <a href="instrain\_near\_lem\_bin28">instrain\_near\_lem\_bin28</a>. The <a href="output">output</a> folder contains several .tsv files, which are described in detail here: <a href="https://instrain.readthedocs.io/en/latest/example\_output.html">https://instrain.readthedocs.io/en/latest/example\_output.html</a>.

Note: all positions in these files are numbered starting from 0. Be careful when comparing to 1-based genome positions!

Briefly, \*\_SNVs.tsv contains a list of all single-nucleotide variants (SNVs). \*gene\_info.tsv, \*scaffold\_info.tsv, and \*genome\_info.tsv contain summaries of variation at the gene, scaffold, and genome (in our case, MAG) level. \*linkage.tsv contains info about how frequently SNVs occur on the same read.

What type of variation would linkage information be useful for detecting?

These files are all tab-separated text files and can be downloaded to your computer and opened in your favorite spreadsheet program.

#### Step 4: Exploration

Bin28 has been identified as purple nonsulfur bacterium from the genus Rhodovulum.

Some questions to ask:

- How many SNVs are present in the genome?
- Which genes in the Rhodovulum MAG have the most SNVs? The least? What do they do? (BLAST is your friend)
- Are there regions with multiple highly-linked SNVs?