

Computational Genomics Tutorial

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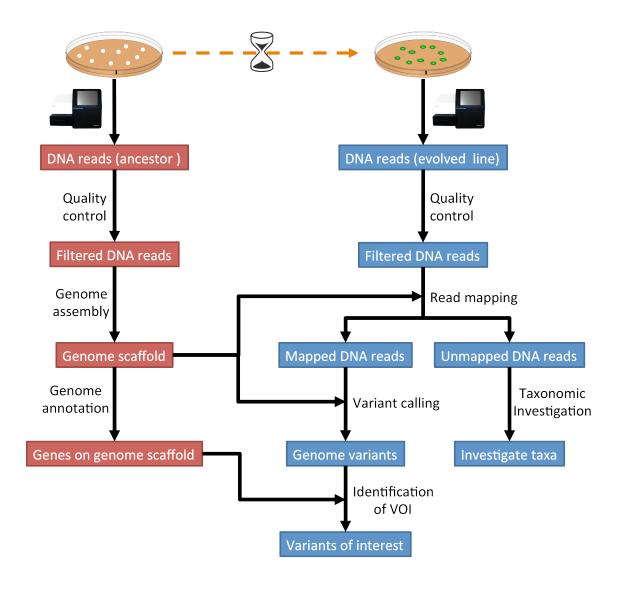
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This is an introductory tutorial for learning computational genomics mostly on the Linux command-line. You will learn how to analyse next-generation sequencing (NGS) data. The data you will be using is real research data. The final aim is to identify genome variations in evolved lines of wild yeast that can explain the observed biological phenotypes. Currently Sebastian¹ is teaching this material in the Massey University course Genetics and Evolution².

More information about other bioinformatics material and our research can be found on the webpages of the Schmeier Group³ (https://sschmeier.com).



Note: A online version of this tutorial can be accessed at https://genomics.sschmeier.com.

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¹ https://sschmeier.com

 $^{^2\} https://www.massey.ac.nz/massey/learning/programme-course/course.cfm?paper_code=203.341$

³ https://sschmeier.com

2 CONTENTS

ONE

INTRODUCTION

This is an introductory tutorial for learning genomics mostly on the Linux command-line. Should you need to refresh your knowledge about either Linux or the command-line, have a look here⁴.

In this tutorial you will learn how to analyse next-generation sequencing (NGS) data. The data you will be using is actual research data. The experiment follows a similar strategy as in what is called an "experimental evolution" experiment [KAWECKI2012], [ZEYL2006]. The final aim is to identify the genome variations in evolved lines of wild yeast that can explain the observed biological phenotype.

1.1 The workflow

The tutorial workflow is summarised in Fig. 1.1.

1.2 Learning outcomes

During this tutorial you will learn to:

- Check the data quality of an NGS experiment
- · Create a genome assembly of the ancestor based on NGS data
- Map NGS reads of evolved lines to the created ancestral reference genome
- Call genome variations/mutations in the evolved lines
- Annotate a newly derived reference genome
- Find variants of interest that may be responsible for the observed evolved phenotype

⁴ http://linux.sschmeier.com/

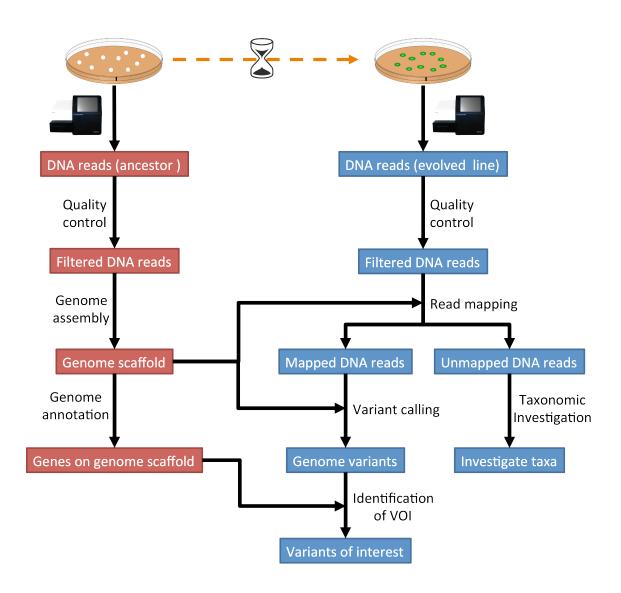


Fig. 1.1: The tutorial will follow this workflow.

TWO

TOOL INSTALLATION

2.1 Install the conda package manager

We will use the package/tool managing system conda⁷ to install some programs that we will use during the course. It is not installed by default, thus we need to install it first to be able to use it.

```
# download latest conda installer
curl -0 https://repo.continuum.io/miniconda/Miniconda3-latest-Linux-x86_64.sh

# run the installer
bash Miniconda3-latest-Linux-x86_64.sh

# delete the installer after successful run
rm Miniconda3-latest-Linux-x86_64.sh
```

Note: Should the conda installer download fail. Please find links to alternative locations on the *Downloads* (page 77) page.

2.1.1 Update .bashrc and .zshrc config-files

Before we are able to use conda⁸ we need to tell our shell where it can find the program. We add the right path to the conda⁹ installation to our shell config files:

```
echo 'export PATH="/home/manager/miniconda3/bin:$PATH"' >> ~/.bashrc
echo 'export PATH="/home/manager/miniconda3/bin:$PATH"' >> ~/.zshrc
```

Attention: The above assumes that your username is "manager", which is the default on a Biolinux install. Replace "manager" with your actual username. Find out with whoami.

So what is actually happening here? We are appending a line to a file (either .bashrc or .zshrc). If we are starting a new command-line shell, either file gets executed first (depending on which shell you are using, either bash or zsh shells). What this line does, is to put permanently the directory ~/miniconda3/bin first on your PATH variable. The PATH variable contains directories in which our computer looks for installed programs, one directory after the other until the program you requested is found (or not, then it will complain). Through the addition of the above line we make sure that the program conda can be found anytime we open a new shell.

Close shell/terminal, **re-open** new shell/terminal. Now, we should be able to use the conda¹⁰ command:

⁷ http://conda.pydata.org/miniconda.html

⁸ http://conda.pydata.org/miniconda.html

⁹ http://conda.pydata.org/miniconda.html

¹⁰ http://conda.pydata.org/miniconda.html

```
conda update conda
```

2.1.2 Installing conda channels to make tools available

Different tools are packaged in what conda¹¹ calls channels. We need to add some channels to make the bioinformatics and genomics tools available for installation:

```
# Install some conda channels
# A channel is where conda looks for packages
conda config --add channels defaults
conda config --add channels conda-forge
conda config --add channels bioconda
```

2.2 Create environments

We create a conda¹² environment for some tools. This is useful to work **reproducible** as we can easily re-create the tool-set with the same version numbers later on.

```
conda create -n ngs python=3
# activate the environment
conda activate ngs
```

So what is happening when you type conda activate ngs in a shell. The PATH variable (mentioned above) gets temporarily manipulated and set to:

```
$ conda activate ngs
# Lets look at the content of the PATH variable
(ngs) $ echo $PATH
/home/manager/miniconda3/envs/ngs/bin:/home/manager/miniconda3/bin:/usr/local/bin: ...
```

Now it will look first in your environment's bin directory but afterwards in the general conda bin (/home/manager/miniconda3/bin). So basically everything you install generally with conda (without being in an environment) is also available to you but gets overshadowed if a similar program is in /home/manager/miniconda3/envs/ngs/bin and you are in the ngs environment.

2.3 Install software

To install software into the activated environment, one uses the command conda install.

```
# install more tools into the environment
conda install package
```

Note: To tell if you are in the correct conda environment, look at the command-prompt. Do you see the name of the environment in round brackets at the very beginning of the prompt, e.g. (ngs)? If not, activate the ngs environment with conda activate ngs before installing the tools.

¹¹ http://conda.pydata.org/miniconda.html

¹² http://conda.pydata.org/miniconda.html

2.4 General conda commands

```
# to search for packages
conda search [package]

# To update all packages
conda update --all --yes

# List all packages installed
conda list [-n env]

# conda list environments
conda env list

# create new env
conda create -n [name] package [package] ...

# activate env
conda activate [name]

# deavtivate env
conda deactivate
```

THREE

QUALITY CONTROL

3.1 Preface

There are many sources of errors that can influence the quality of your sequencing run [ROBASKY2014]. In this quality control section we will use our skill on the command-line interface to deal with the task of investigating the quality and cleaning sequencing data [KIRCHNER2014].

Note: You will encounter some To-do sections at times. Write the solutions and answers into a text-file.

3.2 Overview

The part of the workflow we will work on in this section can be viewed in Fig. 3.1.

3.3 Learning outcomes

After studying this tutorial you should be able to:

- 1. Describe the steps involved in pre-processing/cleaning sequencing data.
- 2. Distinguish between a good and a bad sequencing run.
- 3. Compute, investigate and evaluate the quality of sequence data from a sequencing experiment.

3.4 The data

First, we are going to download the data we will analyse. Open a shell/terminal.

```
# create a directory you work in
mkdir analysis

# change into the directory
cd analysis

# download the data
curl -0 http://compbio.massey.ac.nz/data/203341/data.tar.gz

# uncompress it
tar -xvzf data.tar.gz
```

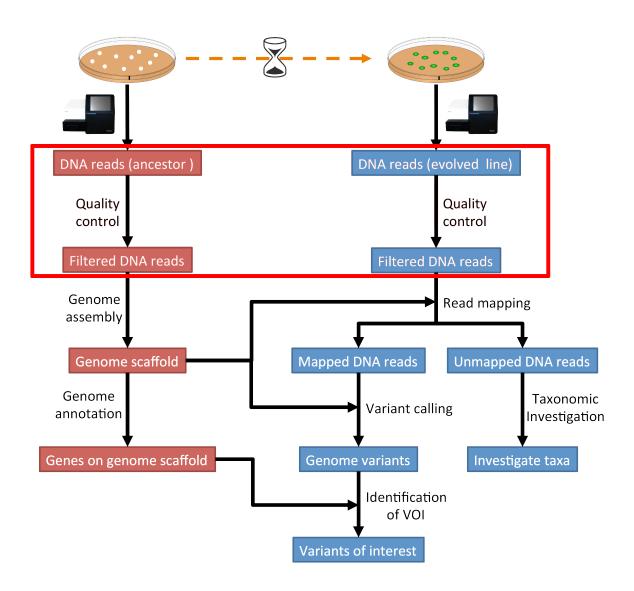


Fig. 3.1: The part of the workflow we will work on in this section marked in red.

Note: Should the download fail, download manually from *Downloads* (page 77).

The data is from a paired-end sequencing run data (see Fig. 3.2) from an Illumina¹³ MiSeq [GLENN2011]. Thus, we have two files, one for each end of the read.

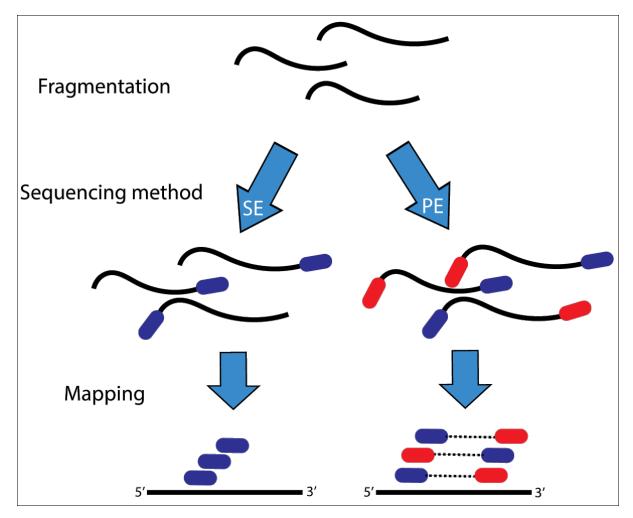


Fig. 3.2: Illustration of single-end (SE) versus paired-end (PE) sequencing.

If you need to refresh how Illumina¹⁴ paired-end sequencing works have a look at the Illumina technology webpage¹⁵ and this video¹⁶.

Attention: The data we are using is "almost" raw data as it came from the machine. This data has been post-processed in two ways already. All sequences that were identified as belonging to the PhiX genome have been removed. This process requires some skills we will learn in later sections. Illumina¹⁷ adapters have been removed as well already! The process is explained below but we are **not** going to do it.

3.4. The data 11

¹³ http://illumina.com

¹⁴ http://illumina.com

 $^{^{15}~{\}rm http://www.illumina.com/technology/next-generation-sequencing/paired-end-sequencing_assay.html}$

¹⁶ https://youtu.be/HMyCqWhwB8E

¹⁷ http://illumina.com

3.4.1 Investigate the data

Make use of your newly developed skills on the command-line to investigate the files in data folder.

Todo:

- 1. Use the command-line to get some ideas about the file.
- 2. What kind of files are we dealing with?
- 3. How many sequence reads are in the file?
- 4. Assume a genome size of 12MB. Calculate the coverage based on this formula: C = LN / G
- C: Coverage
- G: is the haploid genome length in bp
- L: is the read length in bp (e.g. 2x100 paired-end = 200)
- N: is the number of reads sequenced

3.5 The fastq file format

The data we receive from the sequencing is in fastq format. To remind us what this format entails, we can revisit the fastq wikipedia-page¹⁸!

A useful tool to decode base qualities can be found here¹⁹.

Todo: Explain briefly what the quality value represents.

3.6 The QC process

There are a few steps one need to do when getting the raw sequencing data from the sequencing facility:

- 1. Remove PhiX sequences
- 2. Adapter trimming
- 3. Quality trimming of reads
- 4. Quality assessment

3.7 PhiX genome

PhiX²⁰ is a nontailed bacteriophage with a single-stranded DNA and a genome with 5386 nucleotides. PhiX is used as a quality and calibration control for sequencing runs²¹. PhiX is often added at a low known concentration, spiked in the same lane along with the sample or used as a separate lane. As the concentration of the genome is known, one can calibrate the instruments. Thus, PhiX genomic sequences need to be removed before processing your data further as this constitutes a deliberate contamination [MUKHERJEE2015]. The steps involve mapping all reads to the "known" PhiX genome, and removing all of those sequence reads from the data.

¹⁸ https://en.wikipedia.org/wiki/FASTQ format

¹⁹ http://broadinstitute.github.io/picard/explain-qualities.html

²⁰ https://en.wikipedia.org/wiki/Phi_X_174

²¹ http://www.illumina.com/products/by-type/sequencing-kits/cluster-gen-sequencing-reagents/phix-control-v3.html

However, your sequencing provider might not have used PhiX, thus you need to read the protocol carefully, or just do this step in any case.

Attention: We are **not** going to do this step here, as this has been already done. Please see the *Read mapping* (page 25) section on how to map reads against a reference genome.

3.8 Adapter trimming

The process of sequencing DNA via Illumina²² technology requires the addition of some adapters to the sequences. These get sequenced as well and need to be removed as they are artificial and do not belong to the species we try to sequence. Generally speaking adapter trimming takes time.

Attention: The process of how to do this is explained here, however we are **not** going to do this as our sequences have been adapter-trimmed already.

First, we need to know the adapter sequences that were used during the sequencing of our samples. Normally, you should ask your sequencing provider, who should be providing this information to you. Illumina²³ itself provides a document²⁴ that describes the adapters used for their different technologies. Also the FastQC²⁵ tool, we will be using later on, provides a collection of contaminants and adapters²⁶.

Second, we need a tool that takes a list of adapters and scans each sequence read and removes the adapters. Install a tool called fastq-mcf²⁷ from the ea-utils suite²⁸ of tools that is able to do this.

```
# install
conda install ea-utils
```

Using the tool together with a adapter/contaminants list in fasta-file (here denoted as adapters.fa):

```
fastq-mcf -o cleaned.R1.fq.gz -o cleaned.R2.fq.gz adapaters.fa infile_R1.fastq infile_R2.fastq
```

• -o: Specifies the output-files. These are fastq-files for forward and reverse read, with adapters removed.

3.9 Sickle for dynamic trimming

We are using a simple program Sickle²⁹ for dynamic trimming of our sequencing reads to remove bad quality called bases from our reads.

```
conda activate ngs
conda install sickle-trim
```

Now we are going to run the program on our paired-end data:

```
# create a new directory
mkdir trimmed
```

(continues on next page)

²² http://illumina.com

²³ http://illumina.com

²⁴ https://support.illumina.com/downloads/illumina-customer-sequence-letter.html

²⁵ http://www.bioinformatics.babraham.ac.uk/projects/fastqc/

²⁶ https://github.com/csf-ngs/fastqc/blob/master/Contaminants/contaminant_list.txt

²⁷ https://github.com/ExpressionAnalysis/ea-utils/blob/wiki/FastqMcf.md

²⁸ https://expressionanalysis.github.io/ea-utils/

²⁹ https://github.com/najoshi/sickle

(continued from previous page)

```
# sickle parameters:
sickle --help
# as we are dealing with paired-end data you will be using "sickle pe"
sickle pe --help
# run sickle like this on the ancestor:
sickle pe -g -t sanger -f data/ancestor-R1.fastq.gz -r data/ancestor-R2.fastq.gz -o trimmed/
→ancestor-R1.trimmed.fastq.gz -p trimmed/ancestor-R2.trimmed.fastq.gz -s trimmed/ancestor-singles.
→fastq.gz
```

Todo:

1. Run Sickle³⁰ also on the evolved samples.

Hint: Should you not get the command togeter to trim the evolved samples, have a look at the coding solutions at Code: Sickle (page 75). Should you be unable to run Sickle³¹ at all to trim the data. You can download the trimmed dataset here³². Unarchive and uncompress the files with tar -xvzf trimmed. tar.gz.

3.10 Quality assessment of sequencing reads (FastQC)

3.10.1 Installing FastQC

```
conda activate ngs
conda install fastqc
# should now run the program
fastqc --help
```

```
FastQC - A high throughput sequence QC analysis tool
SYNOPSIS
        fastqc seqfile1 seqfile2 .. seqfileN
    fastqc [-o output dir] [--(no)extract] [-f fastq|bam|sam]
           [-c contaminant file] seqfile1 .. seqfileN
DESCRIPTION
    FastQC reads a set of sequence files and produces from each one a quality
    control report consisting of a number of different modules, each one of
    which will help to identify a different potential type of problem in your
    If no files to process are specified on the command line then the program
    will start {f as} an interactive graphical application. If files are provided
    on the command line then the program will run with no user interaction
```

(continues on next page)

³⁰ https://github.com/najoshi/sickle

³¹ https://github.com/najoshi/sickle

³² http://compbio.massey.ac.nz/data/203341/trimmed.tar.gz

(continued from previous page)

required. In this mode it is suitable for inclusion into a standardised analysis pipeline.

3.10.2 FastQC manual

FastQC³³ is a very simple program to run that provides inforation about sequence read quality.

From the webpage:

"FastQC aims to provide a simple way to do some quality control checks on raw sequence data coming from high throughput sequencing pipelines. It provides a modular set of analyses which you can use to give a quick impression of whether your data has any problems of which you should be aware before doing any further analysis."

The basic command looks like:

fastqc -o RESULT-DIR INPUT-FILE.[txt/fa/fq] ...

- -o RESULT-DIR is the directory where the result files will be written
- INPUT-FILE. [txt/fa/fq] is the sequence file to analyze, can be more than one file.

Hint: The result will be a HTML page per input file that can be opened in a web-browser.

Hint: The authors of FastQC³⁴ made some nice help pages explaining each of the plots and results you expect to see here³⁵.

3.10.3 Run FastQC on the untrimmed and trimmed data

Todo:

- 1. Create a directory for the results -> trimmed-fastqc
- 2. Run FastQC on all trimmed files.
- 3. Visit the FastQC³⁶ website and read about sequencing QC reports for good and bad Illumina³⁷ sequencing runs.
- 4. Compare your results to these examples (Fig. 3.3 to Fig. 3.5) of a particularly bad run (taken from the FastQC³⁸ website) and write down your observations with regards to your data.
- 5. What elements in these example figures (Fig. 3.3 to Fig. 3.5) indicate that the example is from a bad run?

Hint: Should you not get it right, try the commands in Code: FastQC (page 75).

³³ http://www.bioinformatics.babraham.ac.uk/projects/fastqc/

³⁴ http://www.bioinformatics.babraham.ac.uk/projects/fastqc/

³⁵ http://www.bioinformatics.babraham.ac.uk/projects/fastqc/Help/3%20Analysis%20Modules/

³⁶ http://www.bioinformatics.babraham.ac.uk/projects/fastqc/

³⁷ http://illumina.com

³⁸ http://www.bioinformatics.babraham.ac.uk/projects/fastqc/

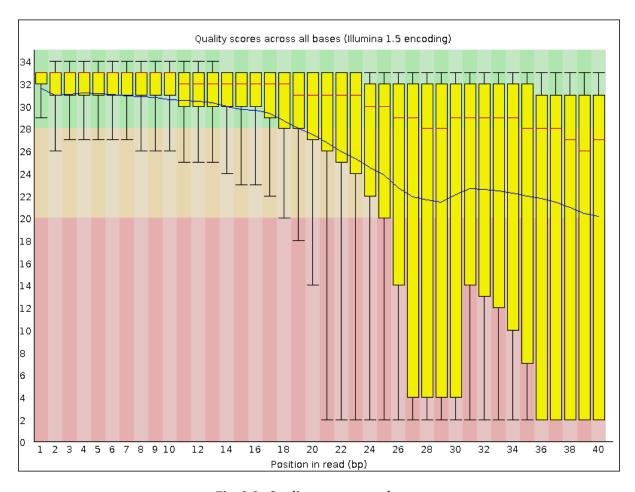


Fig. 3.3: Quality score across bases.

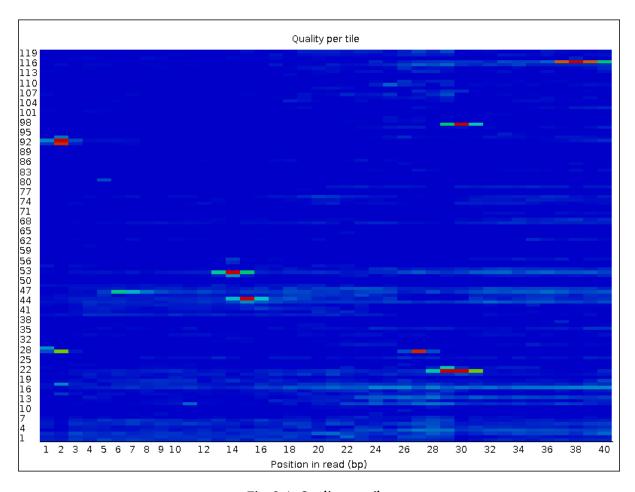


Fig. 3.4: Quality per tile.

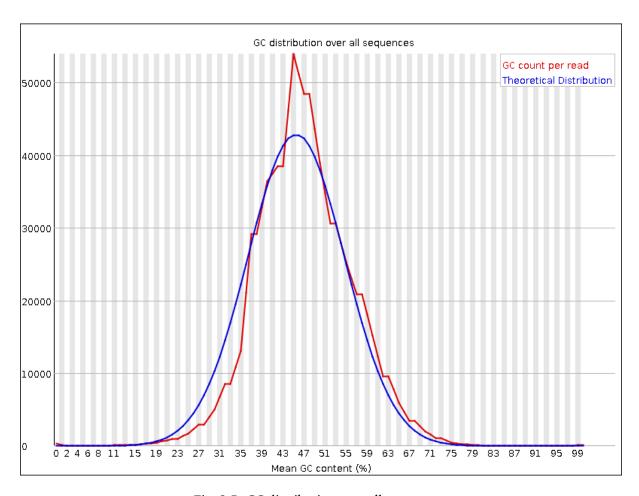


Fig. 3.5: GC distribution over all sequences.

FOUR

GENOME ASSEMBLY

4.1 Preface

In this section we will use our skill on the command-line interface to create a genome assembly from sequencing data.

Note: You will encounter some **To-do** sections at times. Write the solutions and answers into a text-file.

4.2 Overview

The part of the workflow we will work on in this section can be viewed in Fig. 4.1.

4.3 Learning outcomes

After studying this tutorial you should be able to:

- 1. Compute and interpret a whole genome assembly.
- 2. Judge the quality of a genome assembly.

4.4 Before we start

Lets see how our directory structure looks so far:

```
cd ~/analysis
ls -1F
```

```
data/
trimmed/
trimmed-fastqc/
```

4.4.1 Subsampling reads

Due to the size of the data sets you may find that the assembly takes a lot of time to complete, especially on older hardware. To mitigate this problem we can randomly select a subset of sequences we are going to use at this stage of the tutorial. To do this we will install another program:

```
conda activate ngs
conda install seqtk
```

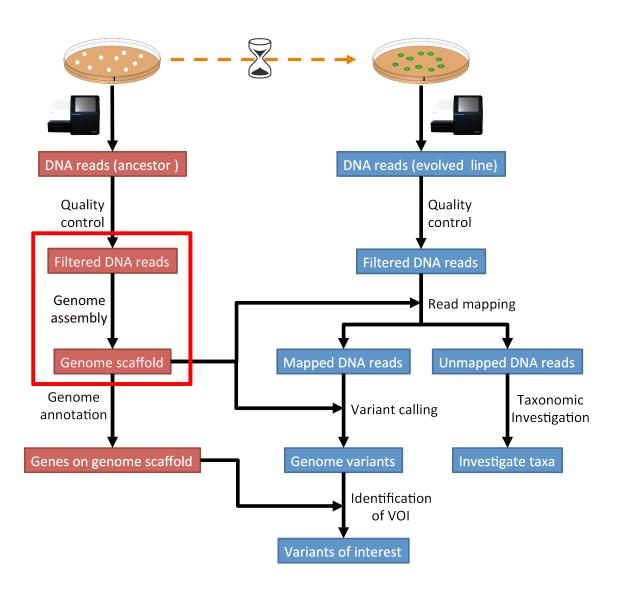


Fig. 4.1: The part of the workflow we will work on in this section marked in red.

Now that seqtk has been installed, we are going to sample 10% of the original reads:

```
# change directory
cd ~/analysis
# create directory
mkdir sampled

# sub sample reads
seqtk sample -s11 trimmed/ancestor-R1.trimmed.fastq.gz 0.1 | gzip > sampled/ancestor-R1.trimmed.
→fastq.gz
seqtk sample -s11 trimmed/ancestor-R2.trimmed.fastq.gz 0.1 | gzip > sampled/ancestor-R2.trimmed.
→fastq.gz
```

In the commands below you need to change the input directory from trimmed/ to sampled/.

Note: The -s options needs to be the same value for file 1 and file 2 to samples the reads that belong to each other. It specified the seed value for the random number generator.

Note: It should be noted that by reducing the amount of reads that go into the assembly, we are loosing information that could otherwise be used to make the assembly. Thus, the assembly will be likely "much" worse than when using the complete dataset.

4.5 Creating a genome assembly

We want to create a genome assembly for our ancestor. We are going to use the quality trimmed forward and backward DNA sequences and use a program called SPAdes⁴³ to build a genome assembly.

Todo:

1. Discuss briefly why we are using the ancestral sequences to create a reference genome as opposed to the evolved line.

4.5.1 Installing the software

We are going to use a program called SPAdes⁴⁴ fo assembling our genome. In a recent evaluation of assembly software, SPAdes⁴⁵ was found to be a good choice for fungal genomes [ABBAS2014]. It is also simple to install and use.

```
conda activate ngs
conda install spades
```

4.5.2 SPAdes usage

```
# change to your analysis root folder
cd ~/analysis
# first create a output directory for the assemblies
```

(continues on next page)

⁴³ http://bioinf.spbau.ru/spades

http://bioinf.spbau.ru/spades

⁴⁵ http://bioinf.spbau.ru/spades

(continued from previous page)

```
mkdir assembly

# to get a help for spades and an overview of the parameter type:
spades.py -h
```

The two files we need to submit to SPAdes⁴⁶ are two paired-end read files.

```
spades.py -o \ assembly/spades-default/ \ -1 \ trimmed/ancestor-R1.trimmed.fastq.gz \ -2 \ trimmed/ancestor-R2.trimmed.fastq.gz
```

Todo:

- 1. Run SPAdes⁴⁷ with default parameters on the ancestor
- 2. Read in the SPAdes⁴⁸ manual about about assembling with 2x150bp reads
- 3. Run SPAdes⁴⁹ a second time but use the options suggested at the SPAdes⁵⁰ manual section 3.4⁵¹ for assembling 2x150bp paired-end reads (are fungi multicellular?). Use a different output directory assembly/spades-150 for this run.

Hint: Should you not get it right, try the commands in Code: SPAdes assembly (trimmed data) (page 75).

4.6 Assembly quality assessment

4.6.1 Assembly statistics

Quast⁵² (QUality ASsesment Tool) [GUREVICH2013], evaluates genome assemblies by computing various metrics, including:

- N50: length for which the collection of all contigs of that length or longer covers at least 50% of assembly length
- NG50: where length of the reference genome is being covered
- NA50 and NGA50: where aligned blocks instead of contigs are taken
- missassemblies: misassembled and unaligned contigs or contigs bases
- genes and operons covered

It is easy with Quast⁵³ to compare these measures among several assemblies. The program can be used on their website⁵⁴.

```
conda install quast
```

Run Quast⁵⁵ with both assembly scaffolds.fasta files to compare the results.

```
46 http://bioinf.spbau.ru/spades
```

⁴⁷ http://bioinf.spbau.ru/spades

⁴⁸ http://bioinf.spbau.ru/spades

⁴⁹ http://bioinf.spbau.ru/spades

⁵⁰ http://bioinf.spbau.ru/spades

⁵¹ http://spades.bioinf.spbau.ru/release3.9.1/manual.html#sec3.4

⁵² http://quast.bioinf.spbau.ru/

⁵³ http://quast.bioinf.spbau.ru/

⁵⁴ http://quast.bioinf.spbau.ru/

⁵⁵ http://quast.bioinf.spbau.ru/

Note: Should you be unable to run SPAdes⁵⁶ on the data, you can manually download the assembly from *Downloads* (page 77). Unarchive and uncompress the files with tar -xvzf assembly.tar.gz.

 $quast -o \ assembly/quast \ assembly/spades-default/scaffolds.fasta \ assembly/spades-150/scaffolds.fasta \ assembly/spades-150/scaffolds.fasta$

Todo:

- 1. Compare the results of Quast⁵⁷ with regards to the two different assemblies.
- 2. Which one do you prefer and why?

4.7 Compare the untrimmed data

Todo:

- 1. To see if our trimming procedure has an influence on our assembly, run the same command you used on the trimmed data on the original untrimmed data.
- 2. Run Quast⁵⁸ on the assembly and compare the statistics to the one derived for the trimmed data set. Write down your observations.

Hint: Should you not get it right, try the commands in Code: SPAdes assembly (original data) (page 75).

4.8 Assemblathon

Todo: Now that you know the basics for assembling a genome and judging their quality, play with the SPAdes⁵⁹ parameters and the **trimmed data** to create the best assembly possible. We will compare the assemblies to find out who created the best one.

Todo:

- 1. Once you have your final assembly, rename your assembly directory int spades_final, e.g. mv assembly/spades-default assembly/spades_final.
- 2. Write down in your notes the command used to create your final assembly.
- 3. Write down in your notes the assembly statistics derived through Quast⁶⁰

 $^{^{56}}$ http://bioinf.spbau.ru/spades

⁵⁷ http://quast.bioinf.spbau.ru/

⁵⁸ http://quast.bioinf.spbau.ru/

⁵⁹ http://bioinf.spbau.ru/spades

⁶⁰ http://quast.bioinf.spbau.ru/

4.9 Further reading

4.9.1 Background on Genome Assemblies

- How to apply de Bruijn graphs to genome assembly. [COMPEAU2011]
- Sequence assembly demystified. [NAGARAJAN2013]

4.9.2 Evaluation of Genome Assembly Software

- GAGE: A critical evaluation of genome assemblies and assembly algorithms. [SALZBERG2012]
- Assessment of de novo assemblers for draft genomes: a case study with fungal genomes. [ABBAS2014]

4.10 Web links

- Lectures for this topic: Genome Assembly: An Introduction⁶¹
- SPAdes⁶²
- Quast⁶³
- Bandage⁶⁴ (Bioinformatics Application for Navigating De novo Assembly Graphs Easily) is a program that visualizes a genome assembly as a graph [WICK2015].

⁶¹ https://dx.doi.org/10.6084/m9.figshare.2972323.v1

⁶² http://bioinf.spbau.ru/spades

⁶³ http://quast.bioinf.spbau.ru/

⁶⁴ https://rrwick.github.io/Bandage/

FIVE

READ MAPPING

5.1 Preface

In this section we will use our skill on the command-line interface to map our reads from the evolved line to our ancestral reference genome.

Note: You will encounter some To-do sections at times. Write the solutions and answers into a text-file.

5.2 Overview

The part of the workflow we will work on in this section can be viewed in Fig. 5.1.

5.3 Learning outcomes

After studying this section of the tutorial you should be able to:

- 1. Explain the process of sequence read mapping.
- 2. Use bioinformatics tools to map sequencing reads to a reference genome.
- 3. Filter mapped reads based on quality.

5.4 Before we start

Lets see how our directory structure looks so far:

```
cd ~/analysis
# create a mapping result directory
mkdir mappings
ls -1F
```

```
assembly/
data/
mappings/
(sampled/)
trimmed/
trimmed-fastqc/
```

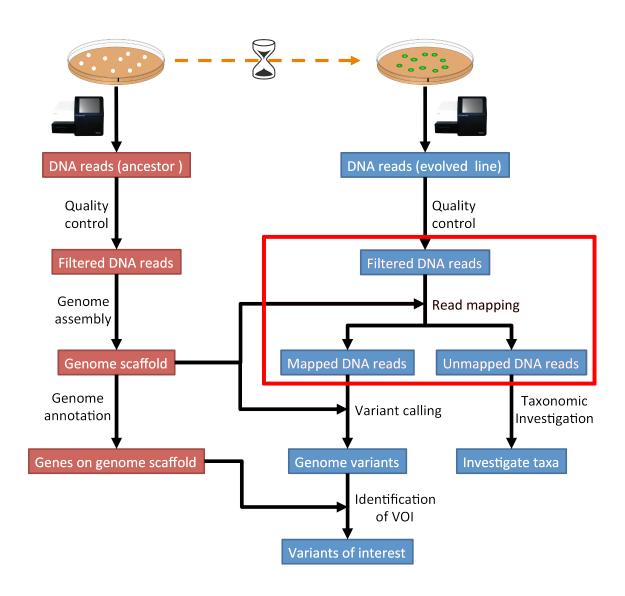


Fig. 5.1: The part of the workflow we will work on in this section marked in red.

Attention: If you sampled reads randomly for the assembly tutorial in the last section, please go and download first the assembly on the full data set. This can be found under *Downloads* (page 77). Unarchive and uncompress the files with tar -xvzf assembly.tar.gz.

5.5 Mapping sequence reads to a reference genome

We want to map the sequencing reads to the ancestral reference genome we created in the section *Genome assembly* (page 19). We are going to use the quality trimmed forward and backward DNA sequences of the evolved line and use a program called BWA⁷¹ to map the reads.

Todo:

1. Discuss briefly why we are using the ancestral genome as a reference genome as opposed to a genome for the evolved line.

5.5.1 Installing the software

We are going to use a program called BWA⁷² to map our reads to our genome.

It is simple to install and use.

```
conda activate ngs
conda install bedtools samtools bwa
```

5.6 BWA

5.6.1 Overview

BWA⁷³ is a short read aligner, that can take a reference genome and map single- or paired-end data to it [LI2009]. It requires an indexing step in which one supplies the reference genome and BWA⁷⁴ will create an index that in the subsequent steps will be used for aligning the reads to the reference genome. The general command structure of the BWA⁷⁵ tools we are going to use are shown below:

```
# bwa index help
bwa index

# indexing
bwa index path/to/reference-genome.fa

# bwa mem help
bwa mem

# single-end mapping, general command structure, adjust to your case
bwa mem path/to/reference-genome.fa path/to/reads.fq.gz > path/to/aln-se.sam

# paired-end mapping, general command structure, adjust to your case
bwa mem path/to/reference-genome.fa path/to/read1.fq.gz path/to/read2.fq.gz > path/to/aln-pe.sam
```

⁷¹ http://bio-bwa.sourceforge.net/

http://bio-bwa.sourceforge.net/

⁷³ http://bio-bwa.sourceforge.net/

⁷⁴ http://bio-bwa.sourceforge.net/

⁷⁵ http://bio-bwa.sourceforge.net/

5.6.2 Creating a reference index for mapping

Todo: Create an BWA⁷⁶ index for our reference genome assembly. Attention! Remember which file you need to submit to BWA⁷⁷.

Hint: Should you not get it right, try the commands in Code: BWA indexing (page 76).

Note: Should you be unable to run BWA⁷⁸ indexing on the data, you can download the index from *Downloads* (page 77). Unarchive and uncompress the files with tar -xvzf bwa-index.tar.gz.

5.6.3 Mapping reads in a paired-end manner

Now that we have created our index, it is time to map the filtered and trimmed sequencing reads of our evolved line to the reference genome.

Todo: Use the correct bwa mem command structure from above and map the reads of the evolved line to the reference genome.

Hint: Should you not get it right, try the commands in Code: BWA mapping (page 76).

5.7 Bowtie2 (alternative to BWA)

Attention: If the mapping did not succeed with BWA⁷⁹. We can use the aligner Bowtie2⁸⁰ explained in this section. If the mapping with BWA⁸¹ did work, you can jump this section. You can jump straight ahead to Section 5.8.

Install with:

conda install bowtie2

5.7.1 Overview

Bowtie 2^{82} is a short read aligner, that can take a reference genome and map single- or paired-end data to it [TRAPNELL2009]. It requires an indexing step in which one supplies the reference genome and Bowtie 2^{83} will create an index that in the subsequent steps will be used for aligning the reads to the reference genome. The general command structure of the Bowtie 2^{84} tools we are going to use are shown below:

⁷⁶ http://bio-bwa.sourceforge.net/

⁷⁷ http://bio-bwa.sourceforge.net/

⁷⁸ http://bio-bwa.sourceforge.net/

⁷⁹ http://bio-bwa.sourceforge.net/

⁸⁰ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

⁸¹ http://bio-bwa.sourceforge.net/

⁸² http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

⁸³ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

⁸⁴ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

```
# bowtie2 help
bowtie2-build

# indexing
bowtie2-build genome.fasta /path/to/index/prefix

# paired-end mapping
bowtie2 -X 1000 -x /path/to/index/prefix -1 read1.fq.gz -2 read2.fq.gz -S aln-pe.sam
```

• -X: Adjust the maximum fragment size (length of paired-end alignments + insert size) to 1000bp. This might be useful if you do not know the exact insert size of your data. The Bowtie2⁸⁵ default is set to 500 which is often considered too short⁸⁶.

5.7.2 Creating a reference index for mapping

Todo: Create an Bowtie2⁸⁷ index for our reference genome assembly. Attention! Remember which file you need to submit to Bowtie2⁸⁸.

Hint: Should you not get it right, try the commands in Code: Bowtie2 indexing (page 76).

Note: Should you be unable to run Bowtie2⁸⁹ indexing on the data, you can download the index from *Downloads* (page 77). Unarchive and uncompress the files with tar -xvzf bowtie2-index.tar.gz.

5.7.3 Mapping reads in a paired-end manner

Now that we have created our index, it is time to map the filtered and trimmed sequencing reads of our evolved line to the reference genome.

Todo: Use the correct bowtie2 command structure from above and map the reads of the evolved line to the reference genome.

Hint: Should you not get it right, try the commands in Code: Bowtie2 mapping (page 76).

Note: Bowtie2⁹⁰ does give very cryptic error messages without telling much why it did not want to run. The most likely reason is that you specified the paths to the files and result file wrongly. Check this first. Use tab completion a lot!

⁸⁵ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

 $^{^{86}\} http://lab.loman.net/2013/05/02/use-x-with-bowtie 2-to-set-minimum-and-maximum-insert-sizes-for-nexter a-libraries/li$

⁸⁷ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

⁸⁸ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

⁸⁹ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

⁹⁰ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

5.8 The sam mapping file-format

Bowtie2⁹¹ and BWA⁹² will produce a mapping file in sam-format. Have a look into the sam-file that was created by either program. A quick overview of the sam-format can be found here⁹³ and even more information can be found here⁹⁴. Briefly, first there are a lot of header lines. Then, for each read, that mapped to the reference, there is one line.

The columns of such a line in the mapping file are described in Table 5.1.

Col	Field	Description
1	QNAME	Query (pair) NAME
2	FLAG	bitwise FLAG
3	RNAME	Reference sequence NAME
4	POS	1-based leftmost POSition/coordinate of clipped sequence
5	MAPQ	MAPping Quality (Phred-scaled)
6	CIAGR	extended CIGAR string
7	MRNM	Mate Reference sequence NaMe ('=' if same as RNAME)
8	MPOS	1-based Mate POSition
9	ISIZE	Inferred insert SIZE
10	SEQ	query SEQuence on the same strand as the reference
11	QUAL	query QUALity (ASCII-33 gives the Phred base quality)
12	OPT	variable OPTional fields in the format TAG:VTYPE:VALUE

Table 5.1: The sam-file format fields.

One line of a mapped read can be seen here:

It basically defines, the read and the position in the reference genome where the read mapped and a quality of the map.

5.9 Mapping post-processing

5.9.1 Fix mates and compress

Because aligners can sometimes leave unusual SAM flag⁹⁵ information on SAM records, it is helpful when working with many tools to first clean up read pairing information and flags with SAMtools⁹⁶. We are going to produce also compressed bam output for efficient storing of and access to the mapped reads. Note, samtools fixmate expects **name-sorted** input files, which we can achieve with samtools sort -n.

```
samtools sort -n -O sam mappings/evolved-6.sam | samtools fixmate -m -O bam - mappings/evolved-6.

→fixmate.bam
```

- -m: Add ms (mate score) tags. These are used by markdup (below) to select the best reads to keep.
- -0 bam: specifies that we want compressed bam output from fixmate

⁹¹ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

⁹² http://bio-bwa.sourceforge.net/

⁹³ http://bio-bwa.sourceforge.net/bwa.shtml#4

⁹⁴ http://samtools.github.io/hts-specs/SAMv1.pdf

⁹⁵ http://bio-bwa.sourceforge.net/bwa.shtml#4

⁹⁶ http://samtools.sourceforge.net/

Attention: The step of sam to bam-file conversion might take a few minutes to finish, depending on how big your mapping file is.

We will be using the SAM flag⁹⁷ information later below to extract specific alignments.

Hint: A very useful tools to explain flags can be found here⁹⁸.

Once we have bam-file, we can also delete the original sam-file as it requires too much space.

rm mappings/evolved-6.sam

5.9.2 Sorting

We are going to use SAMtools⁹⁹ again to sort the bam-file into **coordinate order**:

```
# convert to bam file and sort
samtools sort -0 bam -o mappings/evolved-6.sorted.bam mappings/evolved-6.fixmate.bam
```

- -o: specifies the name of the output file.
- -0 bam: specifies that the output will be bam-format

5.9.3 Remove duplicates

In this step we remove duplicate reads. The main purpose of removing duplicates is to mitigate the effects of PCR amplification bias introduced during library construction. **It should be noted that this step is not always recommended.** It depends on the research question. In SNP calling it is a good idea to remove duplicates, as the statistics used in the tools that call SNPs sub-sequently expect this (most tools anyways). However, for other research questions that use mapping, you might not want to remove duplicates, e.g. RNA-seq.

 $samtools\ markdup\ -r\ -S\ mappings/evolved-6.sorted.bam\ mappings/evolved-6.sorted.dedup.bam$

Todo: Figure out what "PCR amplification bias" means.

Note: Should you be unable to do the post-processing steps, you can download the mapped data from *Downloads* (page 77).

5.10 Mapping statistics

5.10.1 Stats with SAMtools

Lets get an mapping overview:

samtools flagstat mappings/evolved-6.sorted.dedup.bam

⁹⁷ http://bio-bwa.sourceforge.net/bwa.shtml#4

⁹⁸ http://broadinstitute.github.io/picard/explain-flags.html

⁹⁹ http://samtools.sourceforge.net/

Todo: Look at the mapping statistics and understand their meaning¹⁰⁰. Discuss your results. Explain why we may find mapped reads that have their mate mapped to a different chromosome/contig? Can they be used for something?

For the sorted bam-file we can get read depth for at all positions of the reference genome, e.g. how many reads are overlapping the genomic position.

```
samtools depth mappings/evolved-6.sorted.dedup.bam | gzip > mappings/evolved-6.depth.txt.gz
```

Todo: Extract the depth values for contig 20 and load the data into R, calculate some statistics of our scaffold.

```
\verb| zcat mappings/evolved-6.depth.txt.gz | egrep \verb| '^NODE_20_' | gzip > mappings/NODE_20.depth.txt.gz | egrep | | egre
```

Now we quickly use some R^{101} to make a coverage plot for contig NODE20. Open a R^{102} shell by typing R on the command-line of the shell.

```
x <- read.table('mappings/NODE_20.depth.txt.gz', sep='\t', header=FALSE, strip.white=TRUE)

# Look at the beginning of x
head(x)

# calculate average depth
mean(x[,3])
# std dev
sqrt(var(x[,3]))

# mark areas that have a coverage below 20 in red
plot(x[,2], x[,3], col = ifelse(x[,3] < 20,'red','black'), pch=19, xlab='postion', ylab='coverage')

# to save a plot
png('mappings/covNODE20.png', width = 1200, height = 500)
plot(x[,2], x[,3], col = ifelse(x[,3] < 20,'red','black'), pch=19, xlab='postion', ylab='coverage')
dev.off()</pre>
```

The result plot will be looking similar to the one in Fig. 5.2

Todo: Look at the created plot. Explain why it makes sense that you find relatively bad coverage at the beginning and the end of the contig.

5.10.2 Stats with QualiMap

For a more in depth analysis of the mappings, one can use QualiMap¹⁰³ [OKO2015].

QualiMap¹⁰⁴ examines sequencing alignment data in SAM/BAM files according to the features of the mapped reads and provides an overall view of the data that helps to the detect biases in the sequencing and/or mapping of the data and eases decision-making for further analysis.

Installation:

```
100 https://www.biostars.org/p/12475/
101 https://www.r-project.org/
102 https://www.r-project.org/
103 http://qualimap.bioinfo.cipf.es/
104 http://qualimap.bioinfo.cipf.es/
```

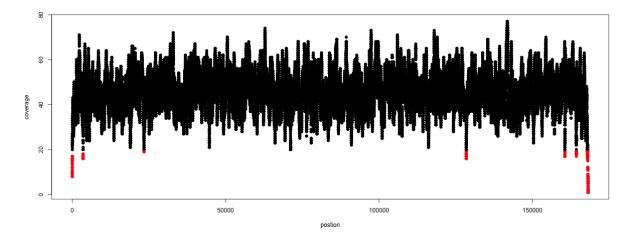


Fig. 5.2: A example coverage plot for a contig with highlighted in red regions with a coverage below 20 reads.

conda install qualimap

Run QualiMap¹⁰⁵ with:

qualimap bamqc -bam mappings/evolved-6.sorted.dedup.bam

This will create a report in the mapping folder. See this webpage¹⁰⁶ to get help on the sections in the report.

Todo: Install QualiMap¹⁰⁷ and investigate the mapping of the evolved sample. Write down your observations.

5.11 Sub-selecting reads

It is important to remember that the mapping commands we used above, without additional parameters to sub-select specific alignments (e.g. for Bowtie2¹⁰⁸ there are options like --no-mixed, which suppresses unpaired alignments for paired reads or --no-discordant, which suppresses discordant alignments for paired reads, etc.), are going to output all reads, including unmapped reads, multi-mapping reads, unpaired reads, discordant read pairs, etc. in one file. We can sub-select from the output reads we want to analyse further using SAMtools¹⁰⁹.

Todo: Explain what concordant and discordant read pairs are? Look at the Bowtie2¹¹⁰ manual.

5.11.1 Concordant reads

We can select read-pair that have been mapped in a correct manner (same chromosome/contig, correct orientation to each other, distance between reads is not stupid).

¹⁰⁵ http://qualimap.bioinfo.cipf.es/

¹⁰⁶ http://qualimap.bioinfo.cipf.es/doc_html/analysis.html#output

¹⁰⁷ http://qualimap.bioinfo.cipf.es/

¹⁰⁸ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

¹⁰⁹ http://samtools.sourceforge.net/

¹¹⁰ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

samtools view -h -b -f 3 mappings/evolved-6.sorted.dedup.bam > mappings/evolved-6.sorted.dedup. \hookrightarrow concordant.bam

- -h: Include the sam header
- -b: Output will be bam-format
- -f 3: Only extract correctly paired reads. -f extracts alignments with the specified SAM flag¹¹¹ set

Todo: Our final aim is to identify variants. For a particular class of variants, it is not the best idea to only focus on concordant reads. Why is that?

5.11.2 Quality-based sub-selection

In this section we want to sub-select reads based on the quality of the mapping. It seems a reasonable idea to only keep good mapping reads. As the SAM-format contains at column 5 the MAPQ value, which we established earlier is the "MAPping Quality" in Phred-scaled, this seems easily achieved. The formula to calculate the MAPQ value is: MAPQ = -10 * log10(p), where p is the probability that the read is mapped wrongly. However, there is a problem! While the MAPQ information would be very helpful indeed, the way that various tools implement this value differs. A good overview can be found here 112. Bottom-line is that we need to be aware that different tools use this value in different ways and the it is good to know the information that is encoded in the value. Once you dig deeper into the mechanics of the MAPQ implementation it becomes clear that this is not an easy topic. If you want to know more about the MAPQ topic, please follow the link above.

For the sake of going forward, we will sub-select reads with at least medium quality as defined by $Bowtie2^{113}$:

```
samtools view -h -b -q 20 mappings/evolved-6.sorted.dedup.bam > mappings/evolved-6.sorted.dedup.q20.

→bam
```

- -h: Include the sam header
- -q 20: Only extract reads with mapping quality >= 20

Hint: I will repeat here a recommendation given at the source $link^{114}$ above, as it is a good one: If you unsure what MAPQ scoring scheme is being used in your own data then you can plot out the MAPQ distribution in a BAM file using programs like the mentioned QualiMap¹¹⁵ or similar programs. This will at least show you the range and frequency with which different MAPQ values appear and may help identify a suitable threshold you may want to use.

5.11.3 Unmapped reads

We could decide to use Kraken2¹¹⁶ like in section *Taxonomic investigation* (page 37) to classify all unmapped sequence reads and identify the species they are coming from and test for contamination.

Lets see how we can get the unmapped portion of the reads from the bam-file:

```
111 http://bio-bwa.sourceforge.net/bwa.shtml#4
```

¹¹² https://sequencing.qcfail.com/articles/mapq-values-are-really-useful-but-their-implementation-is-a-mess/

¹¹³ http://bowtie-bio.sourceforge.net/bowtie2/index.shtml

¹¹⁴ https://sequencing.qcfail.com/articles/mapq-values-are-really-useful-but-their-implementation-is-a-mess/

¹¹⁵ http://qualimap.bioinfo.cipf.es/

¹¹⁶ https://www.ccb.jhu.edu/software/kraken2/

```
samtools view -b -f 4 mappings/evolved-6.sorted.dedup.bam > mappings/evolved-6.sorted.unmapped.bam
# count them
samtools view -c mappings/evolved-6.sorted.unmapped.bam
```

- -b: indicates that the output is BAM.
- -f INT: only include reads with this SAM flag¹¹⁷ set. You can also use the command samtools flags to get an overview of the flags.
- -c: count the reads

Lets extract the fastq sequence of the unmapped reads for read1 and read2.

```
\label{lem:bamToFastq} $$ -i $ mappings/evolved-6.sorted.unmapped.bam -fq mappings/evolved-6.sorted.unmapped.R1. $$ \rightarrow fastq -fq2 mappings/evolved-6.sorted.unmapped.R2.fastq $$
```

¹¹⁷ http://bio-bwa.sourceforge.net/bwa.shtml#4

CHAPTER

SIX

TAXONOMIC INVESTIGATION

6.1 Preface

We want to investate if there are sequences of other species in our collection of sequenced DNA pieces. We hope that most of them are from our species that we try to study, i.e. the DNA that we have extracted and amplified. This might be a way of quality control, e.g. have the samples been contaminated? Lets investigate if we find sequences from other species in our sequence set.

We will use the tool Kraken2¹²¹ to assign taxonomic classifications to our sequence reads. Let us see if we can id some sequences from other species.

Note: You will encounter some To-do sections at times. Write the solutions and answers into a text-file.

6.2 Overview

The part of the workflow we will work on in this section can be viewed in Fig. 6.1.

6.3 Before we start

Lets see how our directory structure looks so far:

```
cd ~/analysis
ls -1F
```

```
assembly/
data/
mappings/
trimmed/
trimmed-fastqc/
```

6.4 Kraken2

We will be using a tool called $Kraken2^{122}$ [WOOD2014]. This tool uses k-mers to assign a taxonomic labels in form of NCBI Taxonomy¹²³ to the sequence (if possible). The taxonomic label is assigned based on similar k-mer content of the sequence in question to the k-mer content of reference genome sequence. The result is a classification of the sequence in question to the most likely taxonomic label. If the k-mer

¹²¹ https://www.ccb.jhu.edu/software/kraken2/

¹²² https://www.ccb.jhu.edu/software/kraken2/

¹²³ https://www.ncbi.nlm.nih.gov/taxonomy

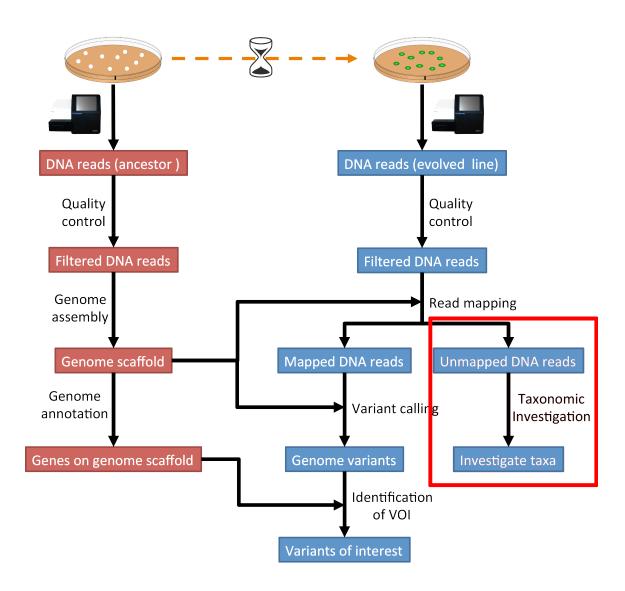


Fig. 6.1: The part of the workflow we will work on in this section marked in red.

content is not similar to any genomic sequence in the database used, it will not assign any taxonomic label.

6.4.1 Installation

Use conda in the same fashion as before to install Kraken²¹²⁴. However, we are going to install kraken into its own environment:

```
conda create --yes -n kraken kraken2 bracken
conda activate kraken
```

Now we create a directory where we are going to do the analysis and we will change into that directory too

```
# make sure you are in your analysis root folder
cd ~/analysis

# create dir
mkdir kraken
cd kraken
```

Now we need to create or download a $Kraken2^{125}$ database that can be used to assign the taxonomic labels to sequences. We opt for downloading the pre-build "minikraken" database from the $Kraken2^{126}$ website:

```
curl -0 https://www.ccb.jhu.edu/software/kraken2/dl/minikraken2_v2_8GB.tgz

# alternatively we can use wget
wget https://www.ccb.jhu.edu/software/kraken2/dl/minikraken2_v2_8GB.tgz

# once the download is finished, we need to extract the archive content:
tar -xvzf minikraken2_v2_8GB.tgz
```

Attention: Should the download fail. Please find links to alternative locations on the *Downloads* (page 77) page.

Note: The "minikraken" database was created from bacteria, viral and archaea sequences. What are the implications for us when we are trying to classify our sequences?

6.4.2 Usage

Now that we have installed Kraken2¹²⁷ and downloaded and extracted the minikraken database, we can attempt to investigate the sequences we got back from the sequencing provider for other species as the one it should contain. We call the Kraken2¹²⁸ tool and specify the database and fasta-file with the sequences it should use. The general command structure looks like this:

```
kraken2 --use-names --threads 4 --db minikraken2_v2_8GB --report example.report.txt example.fa >_

→example.kraken
```

6.4. Kraken2 39

¹²⁴ https://www.ccb.ihu.edu/software/kraken2/

¹²⁵ https://www.ccb.jhu.edu/software/kraken2/

¹²⁶ https://www.ccb.jhu.edu/software/kraken2/

¹²⁷ https://www.ccb.jhu.edu/software/kraken2/

¹²⁸ https://www.ccb.jhu.edu/software/kraken2/

However, we may have fastq-files, so we need to use --fastq-input which tells Kraken2¹²⁹ that it is dealing with fastq-formated files. In addition, we are dealing with paired-end data, which we can tell Kraken2¹³⁰ with the switch --paired. Here, we are investigating one of the unmapped paired-end read files of the evolved line.

This classification may take a while, depending on how many sequences we are going to classify. The resulting content of the file "evolved-6.kraken" looks similar to the following example:

```
7001326F:121:CBVVLANXX:1:1105:2240:12640
                                                                       816:9 171549:5 816:5
→171549:3 2:2 816:5 171549:4 816:34 171549:8 816:4 171549:2 816:10 A:35 816:10 171549:2 816:4_
→171549:8 816:34 171549:4 816:5 2:2 171549:3 816:5 171549:5 816:9
       7001326F:121:CBVVLANXX:1:1105:3487:12536
                                                      1339337 202
                                                                       1339337:67 A:35 1339337:66
U
       7001326F:121:CBVVLANXX:1:1105:5188:12504
                                                       0 251
                                                                       0:91 A:35 0:91
       7001326F:121:CBVVLANXX:1:1105:11030:12689
U
                                                       0
                                                              251
                                                                       0:91 A:35 0:91
       7001326F:121:CBVVLANXX:1:1105:7157:12806
U
                                                                       0:69 A:35 0:68
                                                       0
                                                              206
```

Each sequence classified by Kraken2¹³¹ results in a single line of output. Output lines contain five tabdelimited fields; from left to right, they are:

- 1. C/U: one letter code indicating that the sequence was either classified or unclassified.
- 2. The sequence ID, obtained from the FASTA/FASTQ header.
- 3. The taxonomy ID Kraken2¹³² used to label the sequence; this is **0** if the sequence is unclassified and otherwise should be the NCBI Taxonomy¹³³ identifier.
- 4. The length of the sequence in bp.
- 5. A space-delimited list indicating the lowest common ancestor (in the taxonomic tree) mapping of each k-mer in the sequence. For example, 562:13 561:4 A:31 0:1 562:3 would indicate that:
 - the first 13 k-mers mapped to taxonomy ID #562
 - the next 4 k-mers mapped to taxonomy ID #561
 - the next 31 k-mers contained an ambiguous nucleotide
 - the next k-mer was not in the database
 - the last 3 k-mers mapped to taxonomy ID #562

Note: The Kraken2¹³⁴ manual can be accessed here¹³⁵.

6.4.3 Investigate taxa

We can use the webpage NCBI TaxIdentifier 136 to quickly get the names to the taxonomy identifier. However, this is impractical as we are dealing potentially with many sequences. Kraken 2^{137} has some scripts that help us understand our results better.

```
129 https://www.ccb.jhu.edu/software/kraken2/
```

¹³⁰ https://www.ccb.jhu.edu/software/kraken2/

¹³¹ https://www.ccb.jhu.edu/software/kraken2/

¹³² https://www.ccb.jhu.edu/software/kraken2/

¹³³ https://www.ncbi.nlm.nih.gov/taxonomy

¹³⁴ https://www.ccb.jhu.edu/software/kraken2/

https://www.ccb.jhu.edu/software/kraken2/index.shtml?t=manual

¹³⁶ https://www.ncbi.nlm.nih.gov/Taxonomy/TaxIdentifier/tax_identifier.cgi

¹³⁷ https://www.ccb.jhu.edu/software/kraken2/

Because we used the Kraken2¹³⁸ switch --report FILE, we have got also a sample-wide report of all taxa found. This is much better to get an overview what was found.

The first few lines of an example report are shown below.

```
514312
              514312
                                      unclassified
16.44
      101180
              0
                              1
                      R
16.44 101180
              0
                      R1
                              131567
                                        cellular organisms
16.44 101180 2775
                      D
                              2
                                          Bacteria
13.99 86114
              1
                      D1
                              1783270
                                            FCB group
13.99 86112
                                              Bacteroidetes/Chlorobi group
                      D2
                              68336
              0
13.99 86103
                      Р
                              976
                                                Bacteroidetes
              8
13.94 85798
              2
                      С
                              200643
                                                  Bacteroidia
13.94 85789
              19
                      0
                              171549
                                                    Bacteroidales
13.87 85392
                              815
                                                      Bacteroidaceae
```

The output of kraken-report is tab-delimited, with one line per taxon. The fields of the output, from left-to-right, are as follows:

- 1. Percentage of reads covered by the clade rooted at this taxon
- 2. Number of reads covered by the clade rooted at this taxon
- 3. Number of reads assigned directly to this taxon
- 4. A rank code, indicating (U)nclassified, (D)omain, (K)ingdom, (P)hylum, (C)lass, (O)rder, (F)amily, (G)enus, or (S)pecies. All other ranks are simply "-".
- 5. NCBI Taxonomy¹³⁹ ID
- 6. The indented scientific name

Note: If you want to compare the taxa content of different samples to another, one can create a report whose structure is always the same for all samples, disregarding which taxa are found (obviously the percentages and numbers will be different).

We can cerate such a report using the option --report-zero-counts which will print out all taxa (instead of only those found). We then sort the taxa according to taxa-ids (column 5), e.g. sort -n -k5.

The report is not ordered according to taxa ids and contains all taxa in the database, even if they have not been found in our sample and are thus zero. The columns are the same as in the former report, however, we have more rows and they are now differently sorted, according to the NCBI Taxonomy¹⁴⁰ id.

6.4.4 Bracken

Bracken¹⁴¹ stands for Bayesian Reestimation of Abundance with KrakEN, and is a statistical method that computes the abundance of species in DNA sequences from a metagenomics sample [LU2017]. Bracken¹⁴² uses the taxonomy labels assigned by Kraken²¹⁴³ (see above) to estimate the number of reads originating from each species present in a sample. Bracken¹⁴⁴ classifies reads to the best matching location in the taxonomic tree, but does not estimate abundances of species. Combined with the Kraken classifier, Bracken¹⁴⁵ will produces more accurate species- and genus-level abundance estimates than Kraken²¹⁴⁶ alone.

6.4. Kraken2 41

¹³⁸ https://www.ccb.jhu.edu/software/kraken2/

¹³⁹ https://www.ncbi.nlm.nih.gov/taxonomy

¹⁴⁰ https://www.ncbi.nlm.nih.gov/taxonomy

¹⁴¹ https://ccb.jhu.edu/software/bracken/index.shtml

¹⁴² https://ccb.jhu.edu/software/bracken/index.shtml

¹⁴³ https://www.ccb.jhu.edu/software/kraken2/

¹⁴⁴ https://ccb.jhu.edu/software/bracken/index.shtml

https://ccb.jhu.edu/software/bracken/index.shtml

¹⁴⁶ https://www.ccb.jhu.edu/software/kraken2/

The use of Bracken¹⁴⁷ subsequent to Kraken²¹⁴⁸ is optional but might improve on the Kraken²¹⁴⁹ results.

Installation

We installed Bracken¹⁵⁰ already together with Kraken²¹⁵¹ above, so it should be ready to be used.

However, we need to download some additional files for the Kraken2¹⁵² database.

```
cd minikraken2_v2_8GB
wget https://ccb.jhu.edu/software/bracken/dl/minikraken2_vminikraken2_v2_8GB/database100mers.kmer_
distrib
cd -
```

Usage

Now, we can use Bracken¹⁵³ on the Kraken²¹⁵⁴ results to improve them.

The general structure of the Bracken¹⁵⁵ command look like this:

```
bracken -d DB -i kraken2.report -o bracken.species.txt -l S
```

- -1 S: denotes the level we want to look at. S stands for species but other levels are available.
- -d DB: specifies the Kraken2¹⁵⁶ database that should be used.

Let us apply Bracken¹⁵⁷ to the example above:

```
bracken -d minikraken2_v2_8GB -i evolved-6.kraken -1 S -o evolved-6.bracken
```

The species-focused result-table looks similar to this:

```
name
       taxonomy_id
                       taxonomy_lvl
                                      kraken_assigned_reads
                                                             added_reads
                                                                            new_est_reads _
→fraction_total_reads
Streptococcus sp. oral taxon 431
                                      712633 S
                                                     2
                                                             0
                                                                     2
                                                                            0.00001
                                                                    0.00000
Neorhizobium sp. NCHU2750 1825976 S 0
                                                     0
                                                             0
Pseudomonas sp. MT-1 150396 S 0
Ahniella affigens 2021234 S 1
                                             0
                                                             0.00000
                                                     0
                                             0
                                                     1
                                                             0.00000
Sinorhizobium sp. CCBAU 05631 794846 S
                                             0
                                                     0
                                                                     0.00000
                                                             0
Cohnella sp. 18JY8-7 2480923 S
                                      1
                                             0
                                                     1
                                                             0.00000
Bacillus velezensis
                                                     8
                                                             0.00002
                      492670 S
Actinoplanes missouriensis
                              1866
                                                                     0.00002
```

The important column is the new_est_reads, which gives the newly estimated reads.

6.5 Centrifuge

We can also use another tool by the same group called Centrifuge¹⁵⁸ [KIM2017]. This tool uses a novel indexing scheme based on the Burrows-Wheeler transform (BWT) and the Ferragina-Manzini (FM) index,

```
147 https://ccb.jhu.edu/software/bracken/index.shtml
```

¹⁴⁸ https://www.ccb.jhu.edu/software/kraken2/

¹⁴⁹ https://www.ccb.jhu.edu/software/kraken2/

¹⁵⁰ https://ccb.jhu.edu/software/bracken/index.shtml

¹⁵¹ https://www.ccb.jhu.edu/software/kraken2/

¹⁵² https://www.ccb.jhu.edu/software/kraken2/

¹⁵³ https://ccb.jhu.edu/software/bracken/index.shtml

¹⁵⁴ https://www.ccb.jhu.edu/software/kraken2/

¹⁵⁵ https://ccb.jhu.edu/software/bracken/index.shtml

¹⁵⁶ https://www.ccb.jhu.edu/software/kraken2/

¹⁵⁷ https://ccb.jhu.edu/software/bracken/index.shtml

¹⁵⁸ http://www.ccb.jhu.edu/software/centrifuge/index.shtml

optimized specifically for the metagenomic classification problem to assign a taxonomic labels in form of NCBI Taxonomy¹⁵⁹ to the sequence (if possible). The result is a classification of the sequence in question to the most likely taxonomic label. If the search sequence is not similar to any genomic sequence in the database used, it will not assign any taxonomic label.

Note: I would normally use Kraken2¹⁶⁰ and only prefer Centrifuge¹⁶¹ if memory and/or speed are an issue .

6.5.1 Installation

Use conda in the same fashion as before to install Centrifuge¹⁶²:

```
conda create --yes -n centrifuge centrifuge
conda activate centrifuge
```

Now we create a directory where we are going to do the analysis and we will change into that directory too.

```
# make sure you are in your analysis root folder
cd ~/analysis

# create dir
mkdir centrifuge
cd centrifuge
```

Now we need to create or download a Centrifuge¹⁶³ database that can be used to assign the taxonomic labels to sequences. We opt for downloading the pre-build database from the Centrifuge¹⁶⁴ website:

```
curl -0 ftp://ftp.ccb.jhu.edu/pub/infphilo/centrifuge/data/p_compressed.tar.gz

# alternatively we can use wget
wget ftp://ftp.ccb.jhu.edu/pub/infphilo/centrifuge/data/p_compressed.tar.gz

# once the download is finished, we need to extract the archive content
# It will extract a few files from the archive and may take a moment to finish.
tar -xvzf p_compressed.tar.gz
```

Attention: Should the download fail. Please find links to alternative locations on the *Downloads* (page 77) page.

Note: The database we will be using was created from bacteria and archaea sequences only. What are the implications for us when we are trying to classify our sequences?

6.5.2 Usage

Now that we have installed Centrifuge¹⁶⁵ and downloaded and extracted the pre-build database, we can attempt to investigate the sequences we got back from the sequencing provider for other species as the

```
159 https://www.ncbi.nlm.nih.gov/taxonomy
160 https://www.ccb.jhu.edu/software/kraken2/
161 http://www.ccb.jhu.edu/software/centrifuge/index.shtml
162 http://www.ccb.jhu.edu/software/centrifuge/index.shtml
163 http://www.ccb.jhu.edu/software/centrifuge/index.shtml
164 http://www.ccb.jhu.edu/software/centrifuge/index.shtml
165 http://www.ccb.jhu.edu/software/centrifuge/index.shtml
```

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one it should contain. We call the Centrifuge¹⁶⁶ tool and specify the database and fasta-file with the sequences it should use. The general command structure looks like this:

```
centrifuge -x p_compressed -U example.fa --report-file report.txt -S results.txt
```

However, if we do not have fastq-files we may have to use the -f option, which tells Centrifuge¹⁶⁷ that it is dealing with a fasta-formated file. Here, we are investigating one of the unmapped paired-end read files of the evolved line.

```
centrifuge -x p_compressed -U ../mappings/evolved-6.sorted.unmapped.R1.fastq --report-file evolved-

→6-R1-report.txt -S evolved-6-R1-results.txt
```

This classification may take a moment, depending on how many sequences we are going to classify. The resulting content of the file evolved-6-R1-results.txt looks similar to the following example:

readID	seqID	taxID	score	2ndBestScore	hitLength	queryLe	ngth	numMa	tches
M02810:	197:0000	00000-A	/55U:1:1	101:15316:8461	cid 1747	1747	1892	0	<u>.</u>
→ 103	135	1							
M02810:	197:0000	00000-A\	/55U:1:1	101:15563:3249	cid 161879	161879	18496	0	ш
→ 151	151	1							
M02810:	197:0000	00000-A\	/55U:1:1	101:19743:5166	cid 564 564	10404	10404	117	<u>.</u>
→ 151	2								
M02810:	197:0000	00000-A	/55U:1:1	101:19743:5166	cid 562 562	10404	10404	117	u u
→ 151	2								

Each sequence classified by Centrifuge¹⁶⁸ results in a single line of output. Output lines contain eight tab-delimited fields; from left to right, they are according to the Centrifuge¹⁶⁹ website:

- 1. The read ID from a raw sequencing read.
- 2. The sequence ID of the genomic sequence, where the read is classified.
- 3. The taxonomic ID of the genomic sequence in the second column.
- 4. The score for the classification, which is the weighted sum of hits.
- 5. The score for the next best classification.
- 6. A pair of two numbers: (1) an approximate number of base pairs of the read that match the genomic sequence and (2) the length of a read or the combined length of mate pairs.
- 7. A pair of two numbers: (1) an approximate number of base pairs of the read that match the genomic sequence and (2) the length of a read or the combined length of mate pairs.
- 8. The number of classifications for this read, indicating how many assignments were made.

6.5.3 Investigate taxa

Centrifuge report

The command above creates a Centrifuge¹⁷⁰ report automatically for us. It contains an overview of the identified taxa and their abundances in your supplied sequences (normalised to genomic length):

name	taxID	taxRank	genomeSi	ize	numReads	numUniq	ueReads	abundance
Pseudomo	nas aer	uginosa	287	species	22457305	1	0	0.0
Pseudomo	onas flu	orescens	294	species	14826544	1	1	0.0
Pseudomo	nas put	ida	303	species	6888188 1	1	0.0	

¹⁶⁶ http://www.ccb.jhu.edu/software/centrifuge/index.shtml

¹⁶⁷ http://www.ccb.jhu.edu/software/centrifuge/index.shtml

¹⁶⁸ http://www.ccb.jhu.edu/software/centrifuge/index.shtml

http://www.ccb.jhu.edu/software/centrifuge/index.shtml

¹⁷⁰ http://www.ccb.jhu.edu/software/centrifuge/index.shtml

```
Ralstonia pickettii 329 species 6378979 3 2 0.0
Pseudomonas pseudoalcaligenes 330 species 4691662 1 1 0.0171143
```

Each line contains seven tab-delimited fields; from left to right, they are according to the Centrifuge¹⁷¹ website:

- 1. The name of a genome, or the name corresponding to a taxonomic ID (the second column) at a rank higher than the strain.
- 2. The taxonomic ID.
- 3. The taxonomic rank.
- 4. The length of the genome sequence.
- 5. The number of reads classified to this genomic sequence including multi-classified reads.
- 6. The number of reads uniquely classified to this genomic sequence.
- 7. The proportion of this genome normalized by its genomic length.

Kraken-like report

If we would like to generate a report as generated with the former tool Kraken2¹⁷², we can do it like this:

centrifuge-kreport -x p_compressed evolved-6-R1-results.txt > evolved-6-R1-kreport.txt						
0.00	0	0	U	0	unclassified	
78.74	163	0	-	1	root	
78.74	163	0	-	131567	cellular organisms	
78.74	163	0	D	2	Bacteria	
54.67	113	0	Р	1224	Proteobacteria	
36.60	75	0	С	1236	Gammaproteobacteria	
31.18	64	0	0	91347	Enterobacterales	
30.96	64	0	F	543	Enterobacteriaceae	
23.89	49	0	G	561	Escherichia	
23.37	48	48	S	562	Escherichia coli	
0.40	0	0	S	564	Escherichia fergusonii	
0.12	0	0	S	208962	Escherichia albertii	
3.26	6	0	G	570	Klebsiella	
3.14	6	6	S	573	Klebsiella pneumoniae	
0.12	0	0	S	548	[Enterobacter] aerogenes	
2.92	6	0	G	620	Shigella	
1.13	2	2	S	623	Shigella flexneri	
0.82	1	1	S	624	Shigella sonnei	
0.50	1	1	S	1813821	Shigella sp. PAMC 28760	
0.38	0	0	S	621	Shigella boydii	

This gives a similar (not the same) report as the $Kraken 2^{173}$ tool. The report is tab-delimited, with one line per taxon. The fields of the output, from left-to-right, are as follows:

- 1. Percentage of reads covered by the clade rooted at this taxon
- 2. Number of reads covered by the clade rooted at this taxon
- 3. Number of reads assigned directly to this taxon
- 4. A rank code, indicating (U)nclassified, (D)omain, (K)ingdom, (P)hylum, (C)lass, (O)rder, (F)amily, (G)enus, or (S)pecies. All other ranks are simply "-".

6.5. Centrifuge 45

¹⁷¹ http://www.ccb.jhu.edu/software/centrifuge/index.shtml

¹⁷² https://www.ccb.jhu.edu/software/kraken2/

¹⁷³ https://www.ccb.jhu.edu/software/kraken2/

- 5. NCBI Taxonomy ID
- 6. The indented scientific name

6.6 Visualisation (Krona)

We use the Krona¹⁷⁴ tools to create a nice interactive visualisation of the taxa content of our sample [ONDOV2011]. Fig. 6.2 shows an example (albeit an artificial one) snapshot of the visualisation Krona¹⁷⁵ provides. Fig. 6.2 is a snapshot of the interactive web-page similar to the one we try to create.

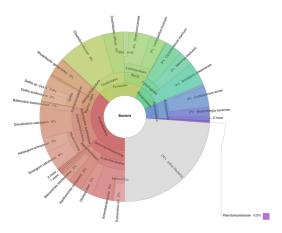


Fig. 6.2: Example of an Krona output webpage.

6.6.1 Installation

Install Krona¹⁷⁶ with:

```
source activate ngs
conda install krona
```

First some house-keeping to make the Krona¹⁷⁷ installation work. Do not worry to much about what is happening here.

```
# we delete a symbolic link that is not correct
rm -rf ~/miniconda3/envs/ngs/opt/krona/taxonomy

# we create a directory in our home where the krona database will live
mkdir -p ~/krona/taxonomy

# now we make a symbolic link to that directory
ln -s ~/krona/taxonomy ~/miniconda3/envs/ngs/opt/krona/taxonomy
```

6.6.2 Build the taxonomy

We need to build a taxonomy database for Krona¹⁷⁸. However, if this fails we will skip this step and just download a pre-build one. Lets first try to build one.

```
174 https://github.com/marbl/Krona/wiki
```

¹⁷⁵ https://github.com/marbl/Krona/wiki

¹⁷⁶ https://github.com/marbl/Krona/wiki

¹⁷⁷ https://github.com/marbl/Krona/wiki

¹⁷⁸ https://github.com/marbl/Krona/wiki

```
ktUpdateTaxonomy.sh ~/krona/taxonomy
```

Now, if this fails, we download a pre-build taxonomy database for krona:

```
# Download pre-build database
curl -0 http://compbio.massey.ac.nz/data/taxonomy.tab.gz

# we unzip the file
gzip -d taxonomy.tab.gz

# we move the unzipped file to our taxonomy directory we specified in the step before.
mv taxonomy.tab ~/krona/taxonomy
```

Attention: Should this also fail we can download a pre-build database on the *Downloads* (page 77) page via a browser.

6.6.3 Visualise

Now, we use the tool ktImportTaxonomy from the Krona¹⁷⁹ tools to create the html web-page. We first need build a two column file (read_id<tab>tax_id) as input to the ktImportTaxonomy tool. We will do this by cutting the columns out of either the Kraken2¹⁸⁰ or Centrifuge¹⁸¹ results:

```
# Kraken2
cd kraken
cat evolved-6.kraken | cut -f 2,3 > evolved-6.kraken.krona
ktImportTaxonomy evolved-6.kraken.krona
firefox taxonomy.krona.html

# Centrifuge
cd centrifuge
cat evolved-6-R1-results.txt | cut -f 1,3 > evolved-6-R1-results.krona
ktImportTaxonomy evolved-6-R1-results.krona
firefox taxonomy.krona.html
```

What happens here is that we extract the second and third column from the Kraken2¹⁸² results. Afterwards, we input these to the Krona¹⁸³ script, and open the resulting web-page in a bowser. Done!

¹⁷⁹ https://github.com/marbl/Krona/wiki

¹⁸⁰ https://www.ccb.jhu.edu/software/kraken2/

¹⁸¹ http://www.ccb.jhu.edu/software/centrifuge/index.shtml

¹⁸² https://www.ccb.jhu.edu/software/kraken2/

¹⁸³ https://github.com/marbl/Krona/wiki

CHAPTER

SEVEN

VARIANT CALLING

7.1 Preface

In this section we will use our genome assembly based on the ancestor and call genetic variants in the evolved line [NIELSEN2011].

7.2 Overview

The part of the workflow we will work on in this section can be viewed in Fig. 7.1.

7.3 Learning outcomes

After studying this tutorial section you should be able to:

#. Use tools to call variants based on a reference genome. #, Be able to describe what influences the calling of variants.

7.4 Before we start

Lets see how our directory structure looks so far:

```
cd ~/analysis
ls -1F
```

```
assembly/
data/
kraken/
mappings/
trimmed/
trimmed-fastqc/
```

7.5 Installing necessary software

Tools we are going to use in this section and how to intall them if you not have done it yet.

```
# activate the env
conda activate ngs
# Install these tools into the conda environment
```

(continues on next page)

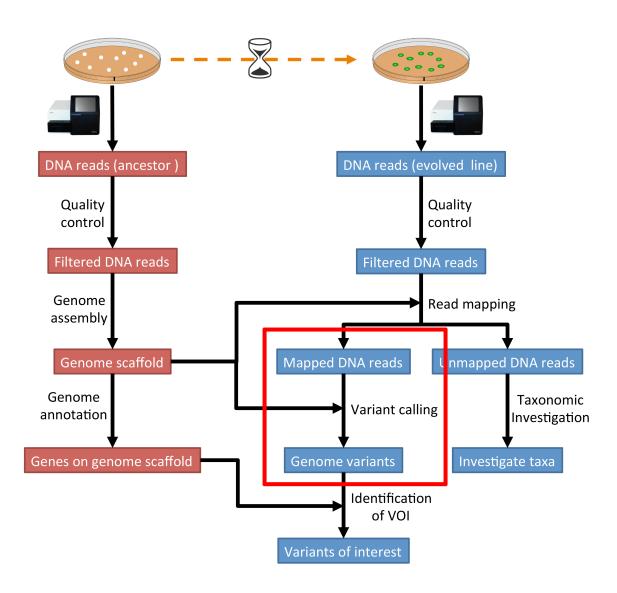


Fig. 7.1: The part of the workflow we will work on in this section marked in red.

(continued from previous page)

```
# if not already installed
conda install samtools
conda install bamtools
conda install freebayes
conda install bedtools
conda install vcflib
conda install rtg-tools
conda install bcftools
```

7.6 Preprocessing

We first need to make an index of our reference genome as this is required by the SNP caller. Given a scaffold/contig file in fasta-format, e.g. scaffolds.fasta which is located in the directory assembly/ spades_final, use SAMtools¹⁸⁸ to do this:

```
samtools faidx assembly/spades_final/scaffolds.fasta
```

Furthermore we need to pre-process our mapping files a bit further and create a bam-index file (.bai) for the bam-file we want to work with:

```
bamtools index -in mappings/evolved-6.sorted.dedup.q20.bam
```

Lets also create a new directory for the variants:

```
mkdir variants
```

7.7 Calling variants

7.7.1 SAMtools mpileup

We use the sorted filtered bam-file that we produced in the mapping step before.

```
# We first pile up all the reads and then call variants samtools mpileup -u -g -f assembly/spades_final/scaffolds.fasta mappings/evolved-6.sorted.dedup.q20.

→bam | bcftools call -v -m -O z -o variants/evolved-6.mpileup.vcf.gz
```

SAMtools¹⁸⁹ mpileup parameter:

- -u: uncompressed output
- -g: generate genotype likelihoods in BCF format
- -f FILE: faidx indexed reference sequence file

BCFtools¹⁹⁰ view parameter:

- -v: output variant sites only
- -m: alternative model for multiallelic and rare-variant calling
- -o: output file-name
- -0 z: output type: 'z' compressed VCF

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¹⁸⁸ http://samtools.sourceforge.net/

¹⁸⁹ http://samtools.sourceforge.net/

¹⁹⁰ http://www.htslib.org/doc/bcftools.html

7.7.2 Freebayes

As an alternative we can do some variant calling with another tool called freebayes¹⁹¹. Given a reference genome scaffold file in fasta-format, e.g. scaffolds.fasta and the index in .fai format and a mapping file (.bam file) and a mapping index (.bai file), we can call variants with freebayes¹⁹² like so:

```
# Now we call variants and pipe the results into a new file freebayes -f assembly/spades_final/scaffolds.fasta mappings/evolved-6.sorted.dedup.q20.bam | gzip >_ → variants/evolved-6.freebayes.vcf.gz
```

7.8 Post-processing

7.8.1 Understanding the output files (.vcf)

Lets look at a vcf-file:

```
# first 10 lines, which are part of the header
zcat variants/evolved-6.mpileup.vcf.gz | head
```

```
##fileformat=VCFv4.2
##FILTER=<ID=PASS,Description="All filters passed">
##samtoolsVersion=1.3.1+htslib=1.3.1
##samtoolsCommand=samtools mpileup -g -f assembly/spades_final/scaffolds.fasta -o variants/evolved-
-6.mpileup.bcf mappings/evolved-6.sorted.q20.bam
##reference=file://assembly/spades_final/scaffolds.fasta
##contig=<ID=NODE_1_length_1419525_cov_15.3898,length=1419525>
##contig=<ID=NODE_2_length_1254443_cov_15.4779,length=1254443>
##contig=<ID=NODE_3_length_972329_cov_15.3966,length=972329>
##contig=<ID=NODE_4_length_951685_cov_15.4231,length=951685>
##contig=<ID=NODE_5_length_925222_cov_15.39,length=925222>
##contig=<ID=NODE_6_length_916533_cov_15.4426,length=916533>
```

Lets look at the variants:

```
# remove header lines and look at top 4 entires
zcat variants/evolved-6.mpileup.vcf.gz | egrep -v '##' | head -4
```

```
#CHROM POS
                                        QUAL
                                                 FILTER INFO
                                                                 FORMAT mappings/evolved-6.sorted.
→q20.bam
NODE_1_length_1419525_cov_15.3898
                                         24721
                                                                 С
                                                                                          DP=12; VDB=0.
\Rightarrow 205941; SGB=-0.680642; MQ0F=0; AC=2; AN=2; DP4=0,0,12,0; MQ=40
                                                                GT:PL
                                                                         1/1:191,36,0
NODE_1_length_1419525_cov_15.3898
                                        157033 .
                                                         AAGAGAGAGAGAGAGAGAGA
→ AAGAGAGAGAGAGAGAGAGA 39.3328 .
                                            INDEL; IDV=6; IMF=0.146341; DP=41; VDB=0.0813946; SGB=-0.
→616816; MQSB=1; MQ0F=0; ICB=1; HOB=0.5; AC=1; AN=2; DP4=13,17,3,3; MQ=42
                                                                        GT:PL 0/1:75,0,255
NODE_1_length_1419525_cov_15.3898
                                        162469 . T C
                                                                         19.609 .
                                                                                          DP=16 · VDR=0
→045681; SGB=-0.511536; RPB=0.032027; MQB=0.832553; BQB=0.130524; MQ0F=0; ICB=1; HOB=0.5; AC=1; AN=2; DP4=13,
\rightarrow 0,3,0;MQ=39
                     GT:PL
                             0/1:54,0,155
```

The fields in a vcf-file are described in he table (Table 7.1) below:

¹⁹¹ https://github.com/ekg/freebayes

¹⁹² https://github.com/ekg/freebayes

- 11 ·		ccc1	C C 1	C	C 1 1
Table	·/ I•	Tho	vcf-file	tormat	HAIde
Iable	/	1111	vci-ine	TOTHIAL	neids.

Col	Field	Description
1	CHROM	Chromosome name
2	POS	1-based position. For an indel, this is the position preceding the indel.
3	ID	Variant identifier. Usually the dbSNP rsID.
4	REF	Reference sequence at POS involved in the variant. For a SNP, it is a single base.
5	ALT	Comma delimited list of alternative seuqence(s).
6	QUAL	Phred-scaled probability of all samples being homozygous reference.
7	FILTER	Semicolon delimited list of filters that the variant fails to pass.
8	INFO	Semicolon delimited list of variant information.
9	FORMAT	Colon delimited list of the format of individual genotypes in the following fields.
10+	Sample(s)	Individual genotype information defined by FORMAT.

7.8.2 Statistics

Now we can use it to do some statistics and filter our variant calls.

First, to prepare out vcf-file for querying we need to index it with tabix:

```
tabix -p vcf variants/evolved-6.mpileup.vcf.gz
```

• -p vcf: input format

We can get some quick stats with rtg vcfstats:

```
rtg vcfstats variants/evolved-6.mpileup.vcf.gz
```

Example output from rtg vcfstats:

```
: variants/evolved-6.mpileup.vcf.gz
Location
Failed Filters
                                 : 0
Passed Filters
                                 : 516
SNPs
                                 : 399
MNPs
                                 : 0
Insertions
                                 : 104
Deletions
                                 : 13
Indels
                                : 0
Same as reference : 0
SNP Transitions/Transversions: 1.87 (286/153)
Total Het/Hom ratio : 3.20 (393/123)
SNP Het/Hom ratio
                               : 8.98 (359/40)

      SNP Het/Hom ratio
      : - (0/0)

      MNP Het/Hom ratio
      : 0.30 (24/80)

      Insertion Het/Hom ratio
      : 33 (10/3)

Deletion Het/Hom ratio
                               : 3.33 (10/3)
Indel Het/Hom ratio
                               : - (0/0)
Insertion/Deletion ratio : 8.00 (104/13)
Indel/SNP+MNP ratio : 0.29 (117/399)
```

However, we can also run BCFtools¹⁹³ to extract more detailed statistics about our variant calls:

```
bcftools stats -F assembly/spades_final/scaffolds.fasta -s - variants/evolved-6.mpileup.vcf.gz >_ 

→variants/evolved-6.mpileup.vcf.gz.stats
```

- -s -: list of samples for sample stats, "-" to include all samples
- -F FILE: faidx indexed reference sequence file to determine INDEL context

Now we take the stats and make some plots (e.g. Fig. 7.2) which are particular of interest if having multiple samples, as one can easily compare them. However, we are only working with one here:

7.8. Post-processing

¹⁹³ http://www.htslib.org/doc/bcftools.html

```
mkdir variants/plots
plot-vcfstats -p variants/plots/ variants/evolved-6.mpileup.vcf.gz.stats
```

• -p: The output files prefix, add a slash at the end to create a new directory.

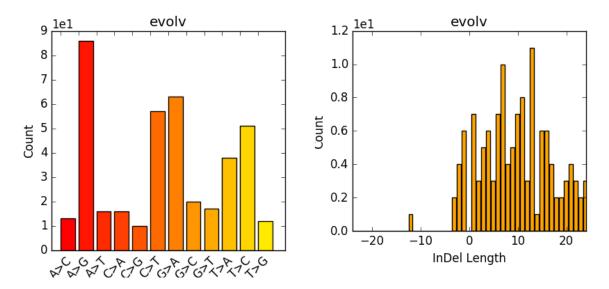


Fig. 7.2: Example of plot-vcfstats output.

7.8.3 Variant filtration

Variant filtration is a big topic in itself [OLSEN2015]. There is no consens yet and research on how to best filter variants is ongoing.

We will do some simple filtration procedures here. For one, we can filter out low quality reads.

Here, we only include variants that have quality > 30.

```
# use rtg vcfffilter
rtg vcffilter -q 30 -i variants/evolved-6.mpileup.vcf.gz -o variants/evolved-6.mpileup.q30.vcf.gz
```

- -i FILE: input file
- -o FILE: output file
- -q FLOAT: minimal allowed quality in output.

or use vcflib¹⁹⁴:

```
# or use vcflib
zcat variants/evolved-6.mpileup.vcf.gz | vcffilter -f "QUAL >= 30" | gzip > variants/evolved-6.

→mpileup.q30.vcf.gz z
```

• -f "QUAL >= 30": we only include variants that have been called with quality >= 30.

Ouick stats for the filtered variants:

```
# look at stats for filtered
rtg vcfstats variants/evolved-6.mpileup.q30.vcf.gz
```

¹⁹⁴ https://github.com/vcflib/vcflib#vcflib

freebayes¹⁹⁵ adds some extra information to the vcf-files it creates. This allows for some more detailed filtering. This strategy will NOT work on the SAMtools¹⁹⁶ mpileup called variants Here we filter, based on some recommendation form the developer of freebayes¹⁹⁷:

```
zcat variants/evolved-6.freebayes.vcf.gz | vcffilter -f "QUAL > 1 & QUAL / AO > 10 & SAF > 0 & SAR_ \rightarrow 0 & RPR > 1 & RPL > 1" | gzip > variants/evolved-6.freebayes.filtered.vcf.gz
```

- QUAL > 1: removes really bad sites
- QUAL / AO > 10: additional contribution of each obs should be 10 log units (\sim Q10 per read)
- SAF > 0 & SAR > 0: reads on both strands
- RPR > 1 & RPL > 1: at least two reads "balanced" to each side of the site

Todo: Look at the statistics. One ratio that is mentioned in the statistics is transition transversion ratio (ts/tv). Explain what this ratio is and why the observed ratio makes sense.

This strategy used here will do for our purposes. However, several more elaborate filtering strategies have been explored, e.g. here 198.

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¹⁹⁵ https://github.com/ekg/freebayes

¹⁹⁶ http://samtools.sourceforge.net/

¹⁹⁷ https://github.com/ekg/freebayes

¹⁹⁸ https://github.com/ekg/freebayes#observation-filters-and-qualities

CHAPTER

EIGHT

GENOME ANNOTATION

8.1 Preface

In this section you will predict genes and assess your assembly using Augustus²⁰¹ and BUSCO²⁰².

Attention: The annotation process will take up to 90 minutes. Start it as soon as possible.

Note: You will encounter some **To-do** sections at times. Write the solutions and answers into a text-file.

8.2 Overview

The part of the workflow we will work on in this section can be viewed in Fig. 8.1.

8.3 Learning outcomes

After studying this section of the tutorial you should be able to:

- 1. Explain how annotation completeness is assessed using orthologues
- 2. Use bioinformatics tools to perform gene prediction
- 3. Use genome-viewing software to graphically explore genome annotations and NGS data overlays

8.4 Before we start

Lets see how our directory structure looks so far:

```
cd ~/analysis
ls -1F
```

assembly/
data/
kraken/
mappings/
trimmed/

(continues on next page)

²⁰¹ http://augustus.gobics.de

²⁰² http://busco.ezlab.org

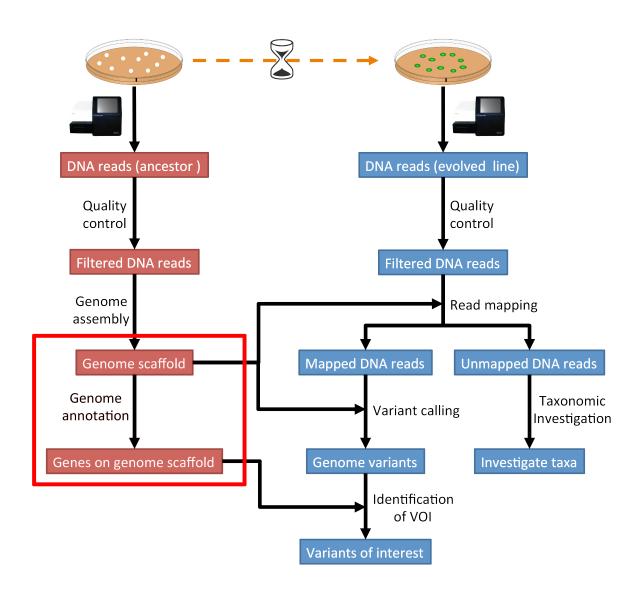


Fig. 8.1: The part of the workflow we will work on in this section marked in red.

(continued from previous page)

```
trimmed-fastqc/
variants/
```

8.5 Installing the software

```
# activate the env
conda activate ngs
conda install busco
```

This will install both the Augustus²⁰³ [STANKE2005] and the BUSCO²⁰⁴ [SIMAO2015] software, which we will use (separately) for gene prediction and assessment of assembly completeness, respectively.

Make a directory for the annotation results:

```
mkdir annotation cd annotation
```

We need to get the database that BUSCO²⁰⁵ will use to assess orthologue presence absence in our genome annotation. We will use wget for this:

```
wget http://busco.ezlab.org/datasets/saccharomycetales_odb9.tar.gz
# unpack the archive
tar -xzvf saccharomycetales_odb9.tar.gz
```

Note: Should the download fail, download manually from *Downloads* (page 77).

We also need to place the configuration file for this program in a location in which we have "write" privileges. Do this recursively for the entire config directory, placing it into your current annotation directory:

```
cp -r ~/miniconda3/envs/ngs/config/ ./
```

We next need to specify the path to this config file so that the program knows where to look now that we have changed the location (note that this is all one line below):

```
export AUGUSTUS_CONFIG_PATH="~/analysis/annotation/config/"n
```

We next check that we have actually changed the path correctly. Entering this into the command should result in the file location being output on the next line of the command prompt.

```
echo $AUGUSTUS_CONFIG_PATH
```

8.6 Assessment of orthologue presence and absence

BUSCO²⁰⁶ will assess orthologue presence absence using blastn²⁰⁷, a rapid method of finding close matches in large databases (we will discuss this in lecture). It uses blastn²⁰⁸ to make sure that it does

```
<sup>203</sup> http://augustus.gobics.de
```

²⁰⁴ http://busco.ezlab.org

²⁰⁵ http://busco.ezlab.org

²⁰⁶ http://busco.ezlab.org

²⁰⁷ https://blast.ncbi.nlm.nih.gov/Blast.cgi?PAGE_TYPE=BlastSearch

²⁰⁸ https://blast.ncbi.nlm.nih.gov/Blast.cgi?PAGE_TYPE=BlastSearch

not miss any part of any possible coding sequences. To run the program, we give it

- A fasta format input file
- · A name for the output files
- The name of the lineage database against which we are assessing orthologue presence absence (that we downloaded above)
- An indication of the type of annotation we are doing (genomic, as opposed to transcriptomic or previously annotated protein files).

```
busco -i ../assembly/spades_final/scaffolds.fasta -o file_name_of_your_choice -l ./

→saccharomycetales_odb9 -m geno
```

Note: This should take about 90 minutes to run. So in the meantime do the next step.

8.7 Annotation

We will use Augustus²⁰⁹ to perform gene prediction. This program implements a hidden markov model (HMM) to infer where genes lie in the assembly you have made. To run the program you need to give it:

- Information as to whether you would like the genes called on both strands (or just the forward or reverse strands)
- A "model" organism on which it can base it's HMM parameters on (in this case we will use S. cerevisiae)
- The location of the assembly file
- A name for the output file, which will be a .gff (general feature format) file.
- We will also tell it to display a progress bar as it moves through the genome assembly.

```
augustus \ --progress=true \ --strand=both \ --species=saccharomyces\_cerevisiae\_S288C \ ../assembly/spades\_ \\ \leftarrow final/scaffolds.fasta > your\_new\_fungus.gff
```

Note: Should the process of producing your annotation fail, you can download a annotation manually from *Downloads* (page 77). Remember to unzip the file.

8.8 Interactive viewing

We will use the software IGV^{210} to view the assembly, the gene predictions you have made, and the variants that you have called, all in one window.

8.9 Installing IGV²¹¹

We will not install this software using conda²¹². Instead, make a new directory in your home directory entitled "software", and change into this directory. You will have to download the software from the Broad Institute:

²⁰⁹ http://augustus.gobics.de

²¹⁰ http://software.broadinstitute.org/software/igv/

²¹¹ http://software.broadinstitute.org/software/igv/

²¹² http://conda.pydata.org/miniconda.html

```
mkdir software
cd software
wget http://data.broadinstitute.org/igv/projects/downloads/2.4/IGV_2.4.10.zip

# unzip the software:
unzip IGV_2.4.10.zip

# and change into that directory.
cd IGV_2.4.10.zip

# To run the interactive GUI, you will need to run the bash script in that directory:
bash igv.sh
```

Note: Should the download fail, download manually from *Downloads* (page 77).

This will open up a new window. Navigate to that window and open up your genome assembly:

- Genome -> load Genome from File
- · Load your assembly, not your gff file.

Load the tracks:

- File -> Load from file
- Load your vcf file from last week
- Load your gff file from this week.

At this point you should be able to zoom in and out to see regions in which there are SNPs or other types of variants. You can also see the predicted genes. If you zoom in far enough, you can see the sequence (DNA and protein).

If you have time and interest, you can right click on the sequence and copy it. Open a new browser window and go to the blastn homepage. There, you can blast your gene of interest (GOI) and see if blast can assign a function to it.

The end goal of this lab will be for you to select a variant that you feel is interesting (e.g. due to the gene it falls near or within), and hypothesize as to why that mutation might have increased in frequency in these evolving yeast populations.

8.10 Assessment of orthologue presence and absence (2)

Hopefully your BUSCO²¹³ analysis will have finished by this time. Navigate into the output directory you created. There are many directories and files in there containing information on the orthologues that were found, but here we are only really interested in one: the summary statistics. This is located in the short_summary*.txt file. Look at this file. It will note the total number of orthologues found, the number expected, and the number missing. This gives an indication of your genome completeness.

Todo: Is it necessarily true that your assembly is incomplete if it is missing some orthologues? Why or why not?

²¹³ http://busco.ezlab.org

CHAPTER

NINE

ORTHOLOGY AND PHYLOGENY

9.1 Preface

In this section you will use some software to find orthologue genes and do phylogenetic reconstructions.

9.2 Learning outcomes

After studying this tutorial you should be able to:

- 1. Use bioinformatics software to find orthologues in the NCBI database.
- 2. Use bioinformatics software to perform sequence alignment.
- 3. Use bioinformatics software to perform phylogenetic reconstructions.

9.3 Before we start

Lets see how our directory structure looks so far:

```
cd ~/analysis
ls -1F
```

```
annotation/
assembly/
data/
kraken/
mappings/
trimmed/
trimmed-fastqc/
variants/
```

Make a directory for the phylogeny results (in your analysis directory):

```
mkdir phylogeny
```

Download the fasta file of the *S. cerevisiase* TEF2 gene to the phylogeny folder:

```
cd phylogeny
curl -O http://compbio.massey.ac.nz/data/203341/s_cerev_tef2.fas
```

Note: Should the download fail, download manually from *Downloads* (page 77).

9.4 Installing the software

```
# activate the env
conda activate ngs
conda install blast
```

This will install a BLAST²¹⁶ executable that you can use to remotely query the NCBI database.

```
conda install muscle
```

This will install MUSCLE²¹⁷, alignment program that you can use to align nucleotide or protein sequences.

We will also install RAxML-NG²¹⁸, a phylogenetic tree inference tool, which uses maximum-likelihood (ML) optimality criterion. However, there is no conda repository for it yet. Thus, we need to download it manually.

```
wget
https://github.com/amkozlov/raxml-ng/releases/download/0.5.1/raxml-ng_v0.5.1b_linux_x86_64.zip
unzip raxml-ng_v0.5.1b_linux_x86_64.zip
rm raxml-ng_v0.5.1b_linux_x86_64.zip
```

9.5 Finding orthologues using BLAST

We will first make a $BLAST^{219}$ database of our current assembly so that we can find the orthologous sequence of the *S. cerevisiae* gene. To do this, we run the command makeblastdb:

```
# create blast db
makeblastdb in ../assembly/spades_final/scaffolds.fasta dbtype nucl
```

To run BLAST²²⁰, we give it:

- -db: The name of the database that we are BLASTing
- -query: A fasta format input file
- A name for the output files
- Some notes about the format we want

First, we blast without any formatting:

```
blastn db ../assembly/spades_final/scaffolds.fasta query s_cerev_tef2.fas > blast.out
```

This should output a file with a set of BLAST²²¹ hits similar to what you might see on the BLAST²²² web site.

Read through the output (e.g. using nano) to see what the results of your BLAST²²³ run was.

Next we will format the output a little so that it is easier to deal with.

```
216 https://blast.ncbi.nlm.nih.gov/Blast.cgi
217 http://www.ebi.ac.uk/Tools/msa/muscle/
218 https://github.com/amkozlov/raxml-ng
219 https://blast.ncbi.nlm.nih.gov/Blast.cgi
220 https://blast.ncbi.nlm.nih.gov/Blast.cgi
221 https://blast.ncbi.nlm.nih.gov/Blast.cgi
222 https://blast.ncbi.nlm.nih.gov/Blast.cgi
223 https://blast.ncbi.nlm.nih.gov/Blast.cgi
224 https://blast.ncbi.nlm.nih.gov/Blast.cgi
```

blastn db ../assembly/spades_final/scaffolds.fasta query s_cerev_tef2.fas evalue 1e-100 outfmt "6_ \rightarrow length sseq" > blast_formatted.out

This will yield a file that has only the sequences of the subject, so that we can later add those to other fasta files. However, the formatting is not perfect. To adjust the format such that it is fasta format, open the file in an editor (e.g. nano) and edit the first line so that it has a name for your sequence. You should know the general format of a fasta-file (e.g. the first line start with a ">").

Hint: To edit in vi editor, you will need to press the escape key and "a" or "e". To save in vi, you will need to press the escape key and "w" (write). To quit vi, you will need to press the escape key and "q" (quit).

Next, you have to replace the dashes (signifying indels in the BLAST 224 result). This can easily be done in vi: Press the escape key, followed by: :%s/\-//g

Now we will BLAST²²⁵ a remote database to get a list of hits that are already in the NCBI database.

Note: It turns out you may not be able to access this database from within BioLinux. In such a case, download the file named blast.fas and place it into your ~/analysis/phylogeny/ directory.

```
curl -O http://compbio.massey.ac.nz/data/203341/blast_u.fas
```

Append the fasta file of your yeast sequence to this file, using whatever set of commands you wish/know.

Note: Should the download fail, download manually from *Downloads* (page 77).

9.6 Performing an alignment

We will use MUSCLE²²⁶ to perform our alignment on all the sequences in the BLAST²²⁷ fasta file. This syntax is very simple (change the filenames accordingly):

```
muscle in infile.fas out your_alignment.aln
```

9.7 Building a phylogeny

We will use RAxML-NG²²⁸ to build our phylogeny. This uses a maximum likelihood method to infer parameters of evolution and the topology of the tree. Again, the syntx of the command is fairly simple, except you must make sure that you are using the directory in which RAxML-NG²²⁹ sits.

The arguments are:

- -s: an alignment file
- -m: a model of evolution. In this case we will use a general time reversible model with gamma distributed rates (GTR+GAMMA)
- -n: outfile-name

²²⁴ https://blast.ncbi.nlm.nih.gov/Blast.cgi

²²⁵ https://blast.ncbi.nlm.nih.gov/Blast.cgi

²²⁶ http://www.ebi.ac.uk/Tools/msa/muscle/

²²⁷ https://blast.ncbi.nlm.nih.gov/Blast.cgi

²²⁸ https://github.com/amkozlov/raxml-ng

²²⁹ https://github.com/amkozlov/raxml-ng

• -p: specify a random number seed for the parsimony inferences

raxmlHPC -s your_alignment.aln -m GTRGAMMA n yeast_tree p 12345

9.8 Visualizing the phylogeny

We will use the online software Interactive Tree of Life $(iTOL)^{230}$ to visualize the tree. Navigate to this homepage. Open the file containing your tree (*bestTree.out), copy the contents, and paste into the web page (in the Tree text box).

You should then be able to zoom in and out to see where your yeast taxa is. To find out the closest relative, you will have to use the NCBI taxa page²³¹.

Todo: Are you certain that the yeast are related in the way that the phylogeny suggests? Why might the topology of this phylogeny not truly reflect the evolutionary history of these yeast species?

http://itol.embl.de/upload.cgi

²³¹ https://www.ncbi.nlm.nih.gov/Taxonomy/TaxIdentifier/tax_identifier.cgi

TEN

VARIANTS-OF-INTEREST

10.1 Preface

In this section we will use our genome annotation of our reference and our genome variants in the evolved line to find variants that are interesting in terms of the observed biology.

Note: You will encounter some To-do sections at times. Write the solutions and answers into a text-file.

10.2 Overview

The part of the workflow we will work on in this section can be viewed in Fig. 10.1.

10.3 Learning outcomes

After studying this section of the tutorial you should be able to:

- 1. Identify variants of interests.
- 2. Understand how the variants might affect the observed biology in the evolved line.

10.4 Before we start

Lets see how our directory structure looks so far:

```
cd ~/analysis
ls -1F
```

```
annotation/
assembly/
data/
kraken/
mappings/
phylogeny/
SolexaQA/
SolexaQA++
trimmed/
trimmed-fastqc/
trimmed-solexaqa/
variants/
```

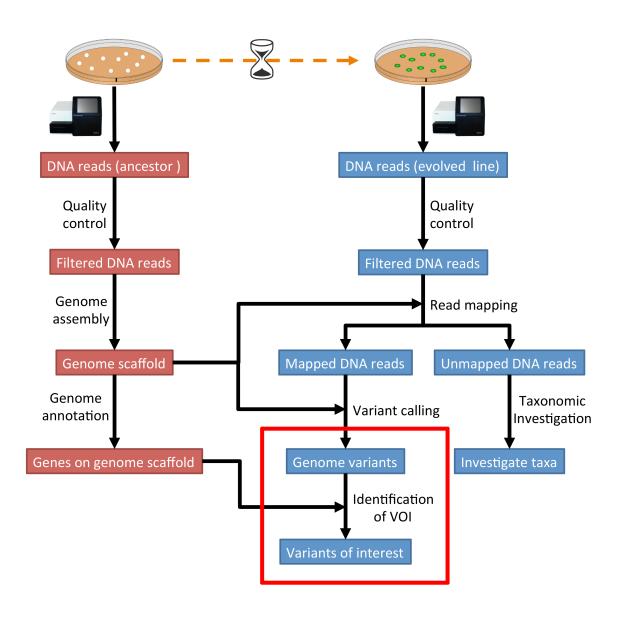


Fig. 10.1: The part of the workflow we will work on in this section marked in red.

10.5 General comments for identifying variants-of-interest

Things to consider when looking for variants-of-interest:

- The quality score of the variant call.
 - Do we call the variant with a higher then normal score?
- The mapping quality score.
 - How confident are we that the reads were mapped at the position correctly?
- The location of the SNP.
 - SNPs in larger contigs are probably more interesting than in tiny contigs.
 - Does the SNP overlap a coding region in the genome annotation?
- The type of SNP.
 - substitutions vs. indels

10.6 SnpEff

We will be using SnpEff²³² to annotate our identified variants. The tool will tell us on to which genes we should focus further analyses.

10.6.1 Installing software

Tools we are going to use in this section and how to install them if you not have done it yet.

```
# activate the env
conda activate ngs

# Install these tools into the conda environment
# if not already installed
conda install snpeff
conda install genometools-genometools
```

Make a directory for the results (in your analysis directory) and change into the directory:

```
mkdir voi

# change into the directory
cd voi
```

10.6.2 Prepare SnpEff database

We need to create our own config-file for SnpEff²³³. Where is the snpEff.config:

```
find ~ -name snpEff.config
/home/manager/miniconda3/envs/ngs/share/snpeff-4.3.1m-0/snpEff.config
```

This will give you the path to the snpEff.config. It might be looking a bit different then the one shown here, depending on the version of SnpEff²³⁴ that is installed.

Make a local copy of the snpEff.config and then edit it with an editor of your choice:

```
<sup>232</sup> http://snpeff.sourceforge.net/index.html
```

²³³ http://snpeff.sourceforge.net/index.html

²³⁴ http://snpeff.sourceforge.net/index.html

```
cp /home/manager/miniconda3/envs/ngs/share/snpeff-4.3.1m-0/snpEff.config .
nano snpEff.config
```

Make sure the data directory path in the snpEff.config looks like this:

```
data.dir = ./data/
```

There is a section with databases, which starts like this:

Add the following two lines in the database section underneath these header lines:

```
# my yeast genome
yeastanc.genome : WildYeastAnc
```

Now, we need to create a local data folder called ./data/yeastanc.

```
# create folders
mkdir -p ./data/yeastanc
```

Copy our genome assembly to the newly created data folder. The name needs to be sequences.fa or yeastanc.fa:

```
cp ../assembly/spades_final/scaffolds.fasta ./data/yeastanc/sequences.fa
gzip ./data/yeastanc/sequences.fa
```

Copy our genome annotation to the data folder. The name needs to be genes.gff (or genes.gtf for gtf-files).

```
cp ../annotation/your_new_fungus.gff ./data/yeastanc/genes.gff
gzip ./data/yeastanc/genes.gff
```

Now we can build a new SnpEff²³⁵ database:

```
snpEff build -c snpEff.config -gff3 -v yeastanc > snpEff.stdout 2> snpEff.stderr
```

Note: Should this fail, due to gff-format of the annotation, we can try to convert the gff to gtf:

```
# using genometools
gt gff3_to_gtf ../annotation/your_new_fungus.gff -o ./data/yeastanc/genes.gtf
gzip ./data/yeastanc/genes.gtf
```

Now, we can use the gtf annotation top build the database:

²³⁵ http://snpeff.sourceforge.net/index.html

```
snpEff build -c snpEff.config -gtf22 -v yeastanc > snpEff.stdout 2> snpEff.stderr
```

10.6.3 SNP annotation

Now we can use our new SnpEff²³⁶ database to annotate some variants, e.g.:

SnpEff²³⁷ adds ANN fields to the vcf-file entries that explain the effect of the variant.

Note: If you are unable to do the annotation, you can download an annotated vcf-file from *Downloads* (page 77).

10.6.4 Example

Lets look at one entry from the original vcf-file and the annotated one. We are only interested in the 8th column, which contains information regarding the variant. SnpEff²³⁸ will add fields here:

When expecting the second entry, we find that SnpEff²³⁹ added annotation information starting with ANN=T|missense_variant|.... If we look a bit more closely we find that the variant results in a amino acid change from a threonine to a serine (c.664T>A|p.Ser222Thr). The codon for serine is TCN and for threonine is ACN, so the variant in the first nucleotide of the codon made the amino acid change.

A quick protein $BLAST^{240}$ of the CDS sequence where the variant was found (extracted from the genes. gff.gz) shows that the closest hit is a translation elongation factor from a species called Candida dubliniensis²⁴¹ another fungi.

```
<sup>236</sup> http://snpeff.sourceforge.net/index.html
```

10.6. SnpEff 71

 $^{^{237}\} http://snpeff.sourceforge.net/index.html$

²³⁸ http://snpeff.sourceforge.net/index.html

²³⁹ http://snpeff.sourceforge.net/index.html

²⁴⁰ https://blast.ncbi.nlm.nih.gov/Blast.cgi

²⁴¹ https://en.wikipedia.org/wiki/Candida_dubliniensis

²⁴² https://blast.ncbi.nlm.nih.gov/Blast.cgi

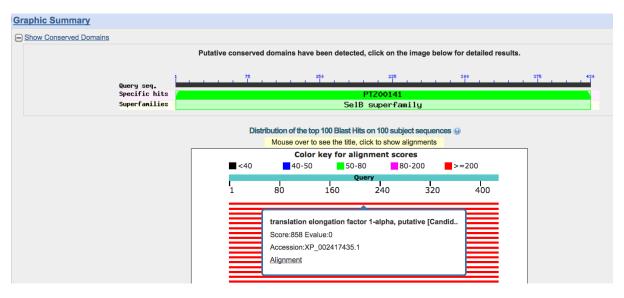


Fig. 10.2: Results of a BLAST²⁴² search of the CDS.

ELEVEN

QUICK COMMAND REFERENCE

11.1 Shell commands

```
# Where in the directory tree am I?
# List the documents and sub-directories in the current directory
# a bit nicer listing with more information
ls -laF
# Change into your home directory
# Change back into the last directory
cd -
# Change one directory up in the tree
# Change explicitly into a directory "temp"
cd temp
# Quickly show content of a file "temp.txt"
# exist the view with "q", navigate line up and down with "k" and "j"
less temp.text
# Show the beginning of a file "temp.txt"
head temp.txt
# Show the end of a file "temp.txt"
tail temp.txt
```

11.2 General conda commands

```
# To update all packages
conda update --all --yes

# List all packages installed
conda list [-n env]

# conda list environments
conda env list
```

(continues on next page)

(continued from previous page)

```
# create new env
conda create -n [name] package [package] ...

# activate env
conda activate [name]

# deavtivate env
conda deactivate
```

TWELVE

CODING SOLUTIONS

12.1 QC

12.1.1 Code: Sickle

12.1.2 Code: FastQC

Create directory:

```
mkdir trimmed-fastqc
```

Run FastQC:

 $fastqc - o \ trimmed-fastqc \ trimmed/ancestor-R1.trimmed.fastq.gz \ trimmed/ancestor-R2.trimmed.fastq.gz \\ {\color{red} \hookrightarrow } trimmed/evolved-6-R1.trimmed.fastq.gz \ trimmed/evolved-6-R2.trimmed.fastq.gz \\ }$

Open html webpages:

```
firefox trimmed-fastqc/\*.html
```

12.2 Assembly

12.2.1 Code: SPAdes assembly (trimmed data)

```
spades.py -o assembly/spades-150/ -k 21,33,55,77 --careful -1 trimmed/ancestor-R1.trimmed.fastq.gz -\rightarrow2 trimmed/ancestor-R2.trimmed.fastq.gz
```

12.2.2 Code: SPAdes assembly (original data)

```
spades.py -o assembly/spades-original/ -k 21,33,55,77 --careful -1 data/ancestor-R1.fastq.gz -2 \rightarrow data/ancestor-R2.fastq.gz
```

12.3 Mapping

12.3.1 Code: Bowtie2 indexing

Build the index:

 $bowtie 2-build\ assembly/spades_final/scaffolds.fasta\ assembly/spades_final/scaffolds$

12.3.2 Code: Bowtie2 mapping

Map to the genome. Use a max fragemnt length of 1000 bp:

 $bowtie 2 - X 1000 - x \ assembly/spades_final/scaffolds - 1 \ trimmed/evolved-6-R1.trimmed.fsatq.gz - 2_ \\ \rightarrow trimmed/evolved-6-R2.trimmed.fastq.gz - S \ mappings/evolved-6.sam$

12.3.3 Code: BWA indexing

Index the genome assembly:

bwa index assembly/spades_final/scaffolds.fasta

12.3.4 Code: BWA mapping

Run bwa mem:

trimmed data

bwa mem assembly/spades_final/scaffolds.fasta trimmed/evolved-6-R1.trimmed.fastq.gz trimmed/evolved- \leftrightarrow 6-R2.trimmed.fastq.gz > mappings/evolved-6.sam

raw data

 $bwa \ mem \ assembly/spades_final/scaffolds.fasta \ data/evolved-6-R1.fastq.gz \ data/evolved-6-R2.fastq.gz \\ \\ \rightarrow> \ mappings/evolved-6.raw.sam$

THIRTEEN

DOWNLOADS

13.1 Tools

- Miniconda installer [EXTERNAL²⁴³ | INTERNAL²⁴⁴ | DROPBOX²⁴⁵]
- Minikraken database [EXTERNAL²⁴⁶]
- Bracken file [EXTERANAL²⁴⁷]
- Centrifuge database [EXTERNAL²⁴⁸ | INTERNAL²⁴⁹ | DROPBOX²⁵⁰]
- Krona taxonomy database [INTERNAL²⁵¹ | DROPBOX²⁵²]
- SolexaQA++ [EXTERNAL²⁵³ | INTERNAL²⁵⁴ | DROPBOX²⁵⁵]
- BUSCO Saccharomycetales_odb9 database [EXTERNAL²⁵⁶ | INTERNAL²⁵⁷ | DROPBOX²⁵⁸]
- IGV [EXTERNAL²⁵⁹ | INTERNAL²⁶⁰ | DROPBOX²⁶¹]
- RAxML-NG [EXTERNAL²⁶² | INTERNAL²⁶³ | DROPBOX²⁶⁴]

13.2 Data

• Quality control (page 9): Raw data-set [INTERNAL²⁶⁵ | DROPBOX²⁶⁶]

```
<sup>243</sup> https://repo.continuum.io/miniconda/Miniconda3-latest-Linux-x86 64.sh
```

²⁴⁴ http://compbio.massey.ac.nz/data/203341/Miniconda3-latest-Linux-x86_64.sh

²⁴⁵ https://www.dropbox.com/s/tz2wocdzjr4grdy/Miniconda3-latest-Linux-x86_64.sh?dl=0

²⁴⁶ https://www.ccb.jhu.edu/software/kraken2/dl/minikraken2_v2_8GB.tgz

²⁴⁷ https://ccb.jhu.edu/software/bracken/dl/minikraken2_v2/database100mers.kmer_distrib

²⁴⁸ ftp://ftp.ccb.jhu.edu/pub/infphilo/centrifuge/data/p_compressed.tar.gz

²⁴⁹ http://compbio.massey.ac.nz/data/203341/p_compressed.tar.gz

²⁵⁰ https://www.dropbox.com/s/a4u6u37ngk2nawx/p_compressed.tar.gz?dl=0

²⁵¹ http://compbio.massey.ac.nz/data/203341/taxonomy.tab.gz

 $^{^{252}\} https://www.dropbox.com/s/cwf1qc5zyq65yvn/taxonomy.tab.gz?dl=0$

²⁵³ https://downloads.sourceforge.net/project/solexaqa/src/SolexaQA%2B%2B_v3.1.7.1.zip?r=https%3A%2F%2Fsourceforge.net%2Fprojects%2Fsolexaqa%2Ffiles%2F&ts=1495062885&use mirror=iweb

²⁵⁴ http://compbio.massey.ac.nz/data/203341/SolexaQA.tar.gz

²⁵⁵ https://www.dropbox.com/s/r9a7hg0tlwe6pk4/SolexaQA.tar.gz?dl=0

²⁵⁶ http://busco.ezlab.org/datasets/saccharomycetales_odb9.tar.gz

²⁵⁷ http://compbio.massey.ac.nz/data/203341/saccharomycetales_odb9.tar.gz

²⁵⁸ https://www.dropbox.com/s/7ow5yi6s5a0ente/saccharomycetales_odb9.tar.gz?dl=0

²⁵⁹ http://data.broadinstitute.org/igv/projects/downloads/IGV_2.3.92.zip

²⁶⁰ http://compbio.massey.ac.nz/data/203341/IGV_2.3.92.zip

²⁶¹ https://www.dropbox.com/s/bpucaolxhwf78le/IGV_2.3.92.zip?dl=0

²⁶² https://github.com/amkozlov/raxml-ng/releases/download/0.3.0/raxml-ng_v0.3.0b_linux_x86_64.zip

http://compbio.massey.ac.nz/data/203341/raxml-ng_v0.3.0b_linux_x86_64.zip

²⁶⁴ https://www.dropbox.com/s/iliws53ri5z4y69/raxml-ng_v0.3.0b_linux_x86_64.zip?dl=0

²⁶⁵ http://compbio.massey.ac.nz/data/203341/data.tar.gz

²⁶⁶ https://www.dropbox.com/s/70gcfqzrqugwcn5/data.tar.gz?dl=0

- Quality control (page 9): Trimmed data-set [INTERNAL²⁶⁷ | DROPBOX²⁶⁸]
- Genome assembly (page 19): Assembled data-set [INTERNAL²⁶⁹ | DROPBOX²⁷⁰]
- Read mapping (page 25): Mapping index (bowtie2) [INTERNAL²⁷¹ | DROPBOX²⁷²]
- Read mapping (page 25): Mapping index (bwa) [INTERNAL²⁷³ | DROPBOX²⁷⁴]
- Read mapping (page 25): Mapped data [INTERNAL²⁷⁵ | DROPBOX²⁷⁶]
- Genome annotation (page 57): GFF annotation file [INTERNAL²⁷⁷ | DROPBOX²⁷⁸]
- Orthology and Phylogeny (page 63): S. cerevisiase TEF2 gene file [INTERNAL²⁷⁹ | DROPBOX²⁸⁰]
- Orthology and Phylogeny (page 63): BLAST file [INTERNAL²⁸¹ | DROPBOX²⁸²]
- Variants-of-interest (page 67): SnpEff annotated vcf-file [INTERNAL²⁸³ | DROPBOX²⁸⁴]

```
<sup>267</sup> http://compbio.massey.ac.nz/data/203341/trimmed.tar.gz
```

 $^{^{268}\} https://www.dropbox.com/s/o6ioadoxfppbjrv/trimmed.tar.gz?dl{=}0$

http://compbio.massey.ac.nz/data/203341/assembly.tar.gz

 $^{^{270}}$ https://www.dropbox.com/s/vlyn2fxgkiml5m8/assembly.tar.gz?dl=0

http://compbio.massey.ac.nz/data/203341/bowtie2-index.tar.gz

²⁷² https://www.dropbox.com/s/dcbdsxl5bjhmif8/bowtie2-index.tar.gz?dl=0

http://compbio.massey.ac.nz/data/203341/bwa-index.tar.gz

²⁷⁴ https://www.dropbox.com/s/yidw27u56iast9z/bwa-index.tar.gz?dl=0

²⁷⁵ http://compbio.massey.ac.nz/data/203341/evolved-6.sorted.dedup.bam

 $^{^{276}\} https://www.dropbox.com/s/k1qn63rwnojhmrz/evolved-6.sorted.dedup.bam?dl=0$

²⁷⁷ http://compbio.massey.ac.nz/data/203341/your_new_fungus.gff.gz

²⁷⁸ https://www.dropbox.com/s/6bo9g8h3q6h1x8x/your_new_fungus.gff.gz?dl=0

²⁷⁹ http://compbio.massey.ac.nz/data/203341/s_cerev_tef2.fas

²⁸⁰ https://www.dropbox.com/s/ooxl2q0vp0bzmq3/s_cerev_tef2.fas?dl=0

²⁸¹ http://compbio.massey.ac.nz/data/203341/blast_u.fas

²⁸² https://www.dropbox.com/s/ul4dzx44lfoewzx/blast_u.fas?dl=0

²⁸³ http://compbio.massey.ac.nz/data/203341/evolved-6.freebayes.filtered.anno.vcf

²⁸⁴ https://www.dropbox.com/s/67m45v5fghdh0d3/evolved-6.freebayes.filtered.anno.vcf?dl=0

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