



Computational Genomics Tutorial

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CONTENTS

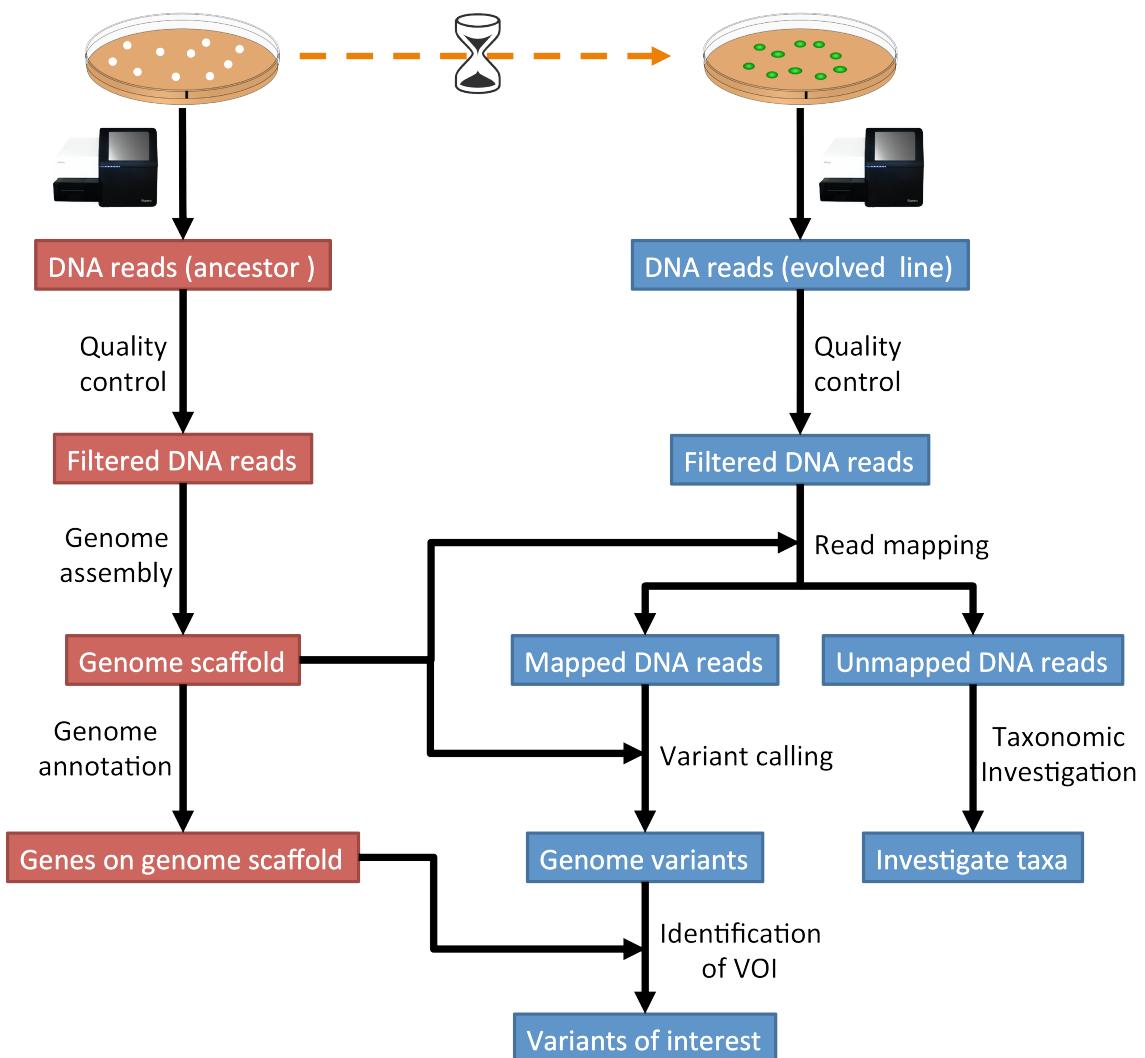
1	Introduction	3
1.1	The workflow	3
1.2	Learning outcomes	3
2	Tool installation	5
2.1	Install the conda package manager	5
2.2	Create environments	6
2.3	Install software	6
2.4	General conda commands	6
3	Quality control	9
3.1	Preface	9
3.2	Overview	9
3.3	Learning outcomes	9
3.4	The data	9
3.5	The fastq file format	12
3.6	The QC process	12
3.7	PhiX genome	12
3.8	Adapter trimming	13
3.9	Quality assessment of sequencing reads (SolexaQA++)	13
3.10	Sickle for dynamic trimming (alternative to SolexaQA++)	15
3.11	Quality assessment of sequencing reads (FastQC)	18
4	Genome assembly	23
4.1	Preface	23
4.2	Overview	23
4.3	Learning outcomes	23
4.4	Before we start	23
4.5	Creating a genome assembly	25
4.6	Assembly quality assessment	26
4.7	Compare the untrimmed data	27
4.8	Assemblathon	27
4.9	Further reading	27
4.10	Web links	28
5	Read mapping	29
5.1	Preface	29
5.2	Overview	29
5.3	Learning outcomes	29
5.4	Before we start	29
5.5	Mapping sequence reads to a reference genome	29
5.6	Bowtie2	31
5.7	BWA	32
5.8	The sam mapping file-format	33
5.9	Mapping post-processing	34

5.10 Mapping statistics	35
5.11 Sub-selecting reads	37
6 Taxonomic investigation	39
6.1 Preface	39
6.2 Overview	39
6.3 Before we start	39
6.4 Kraken2	39
6.5 Centrifuge	44
6.6 Visualisation (Krona)	47
7 Variant calling	51
7.1 Preface	51
7.2 Overview	51
7.3 Learning outcomes	51
7.4 Before we start	51
7.5 Installing necessary software	51
7.6 Preprocessing	53
7.7 Calling variants	53
7.8 Post-processing	54
8 Genome annotation	59
8.1 Preface	59
8.2 Overview	59
8.3 Learning outcomes	59
8.4 Before we start	59
8.5 Installing the software	61
8.6 Assessment of orthologue presence and absence	61
8.7 Annotation	62
8.8 Interactive viewing	62
8.9 Installing IGV	62
8.10 Assessment of orthologue presence and absence (2)	63
9 Orthology and Phylogeny	65
9.1 Preface	65
9.2 Learning outcomes	65
9.3 Before we start	65
9.4 Installing the software	66
9.5 Finding orthologues using BLAST	66
9.6 Performing an alignment	67
9.7 Building a phylogeny	67
9.8 Visualizing the phylogeny	68
10 Variants-of-interest	69
10.1 Preface	69
10.2 Overview	69
10.3 Learning outcomes	69
10.4 Before we start	69
10.5 General comments for identifying variants-of-interest	69
10.6 SnpEff	71
11 Quick command reference	75
11.1 Shell commands	75
11.2 General conda commands	75
12 Coding solutions	77
12.1 QC	77
12.2 Assembly	78
12.3 Mapping	78

13 Downloads	79
13.1 Tools	79
13.2 Data	79
Bibliography	85

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>.

¹ <https://sschmeier.com>

² https://www.massey.ac.nz/massey/learning/programme-course/course.cfm?paper_code=203.341

³ <https://sschmeier.com>

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 [\[KAWECKI2012\]](#) (page 85), [\[ZEYL2006\]](#) (page 85). 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/>

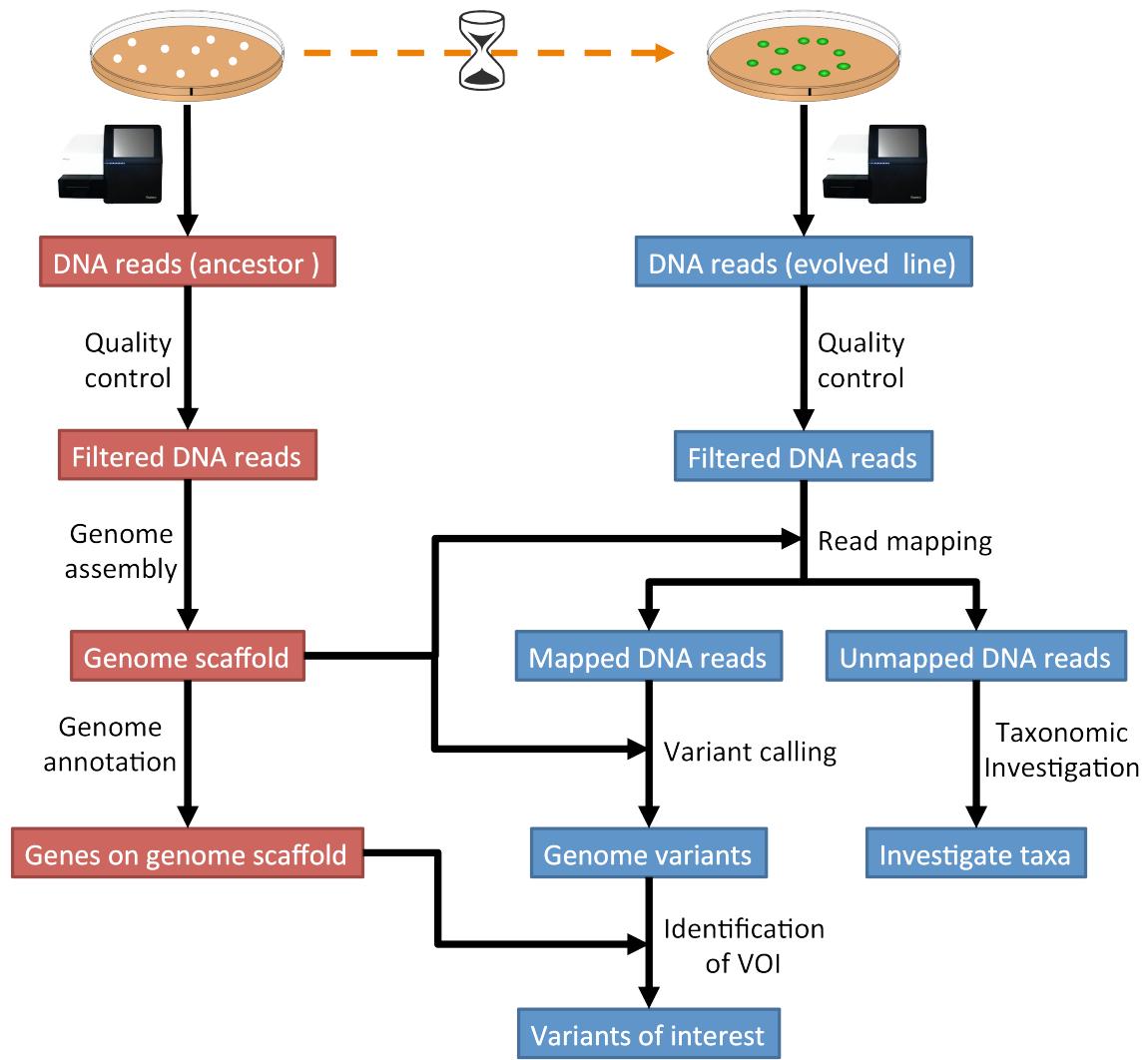


Fig. 1.1: The tutorial will follow this workflow.

TOOL INSTALLATION

2.1 Install the conda package manager

We will use the package/tool managing system `conda`³⁵ 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 -O 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 79) page.

2.1.1 Update .bashrc and .zshrc config-files

Before we are able to use `conda`³⁶ we need to tell our shell where it can find the program. We add the right path to the `conda`³⁷ 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 Biolinix 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 `conda`³⁸ command:

³⁵ <http://conda.pydata.org/miniconda.html>

³⁶ <http://conda.pydata.org/miniconda.html>

³⁷ <http://conda.pydata.org/miniconda.html>

³⁸ <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 `conda`³⁹ 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 `conda`⁴⁰ 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.

2.4 General conda commands

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

(continues on next page)

³⁹ <http://conda.pydata.org/miniconda.html>

⁴⁰ <http://conda.pydata.org/miniconda.html>

(continued from previous page)

```
# 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]

# deactivate env
conda deactivate
```


QUALITY CONTROL

3.1 Preface

There are many sources of errors that can influence the quality of your sequencing run [[ROBASKY2014](#)] (page 85). 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](#)] (page 85).

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

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

Note: Should the download fail, download manually from [Downloads](#) (page 79).

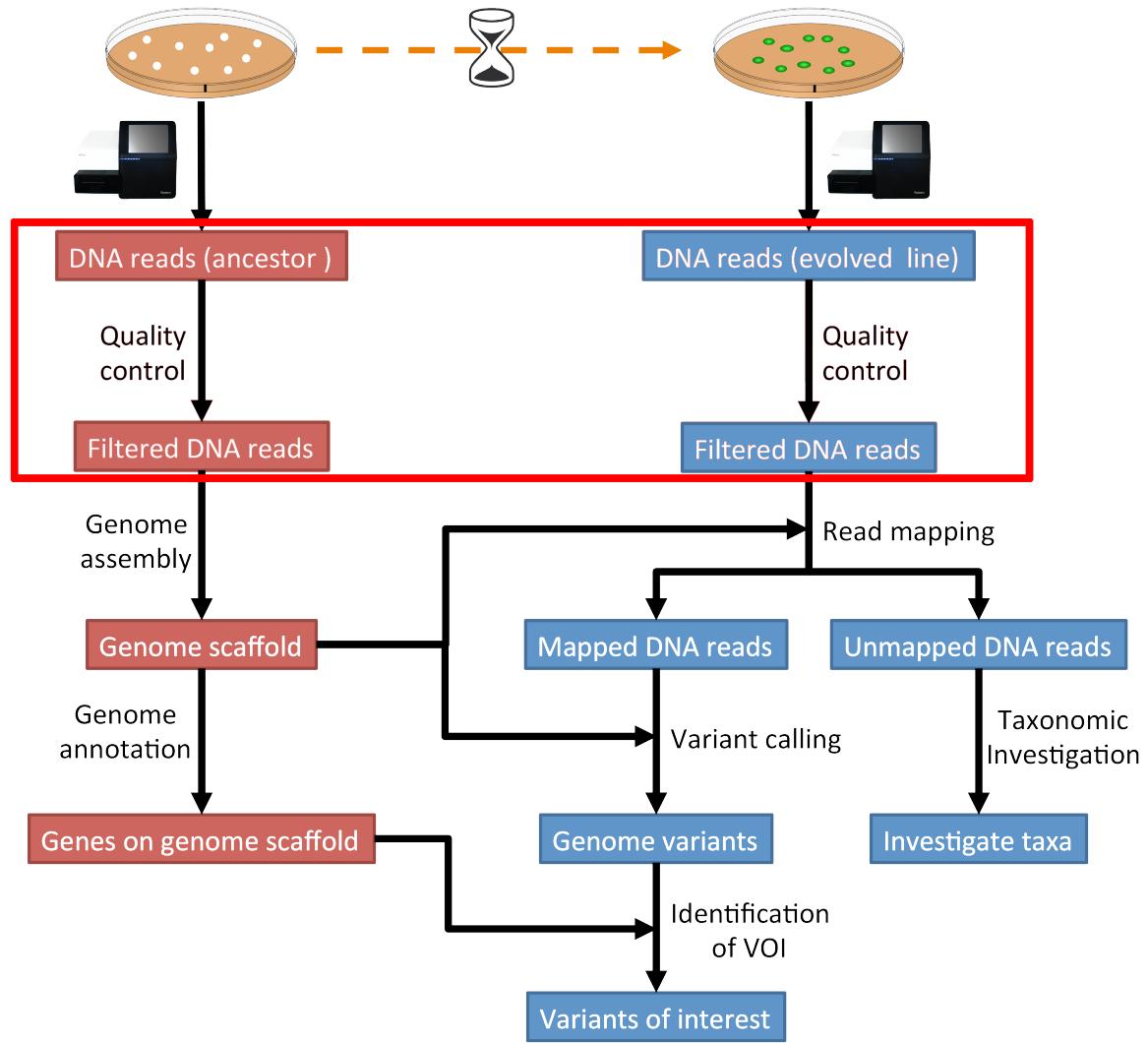


Fig. 3.1: The part of the workflow we will work on in this section marked in red.

The data is from a paired-end sequencing run data (see Fig. 3.2) from an Illumina⁶⁹ MiSeq [GLENN2011] (page 85). Thus, we have two files, one for each end of the read.

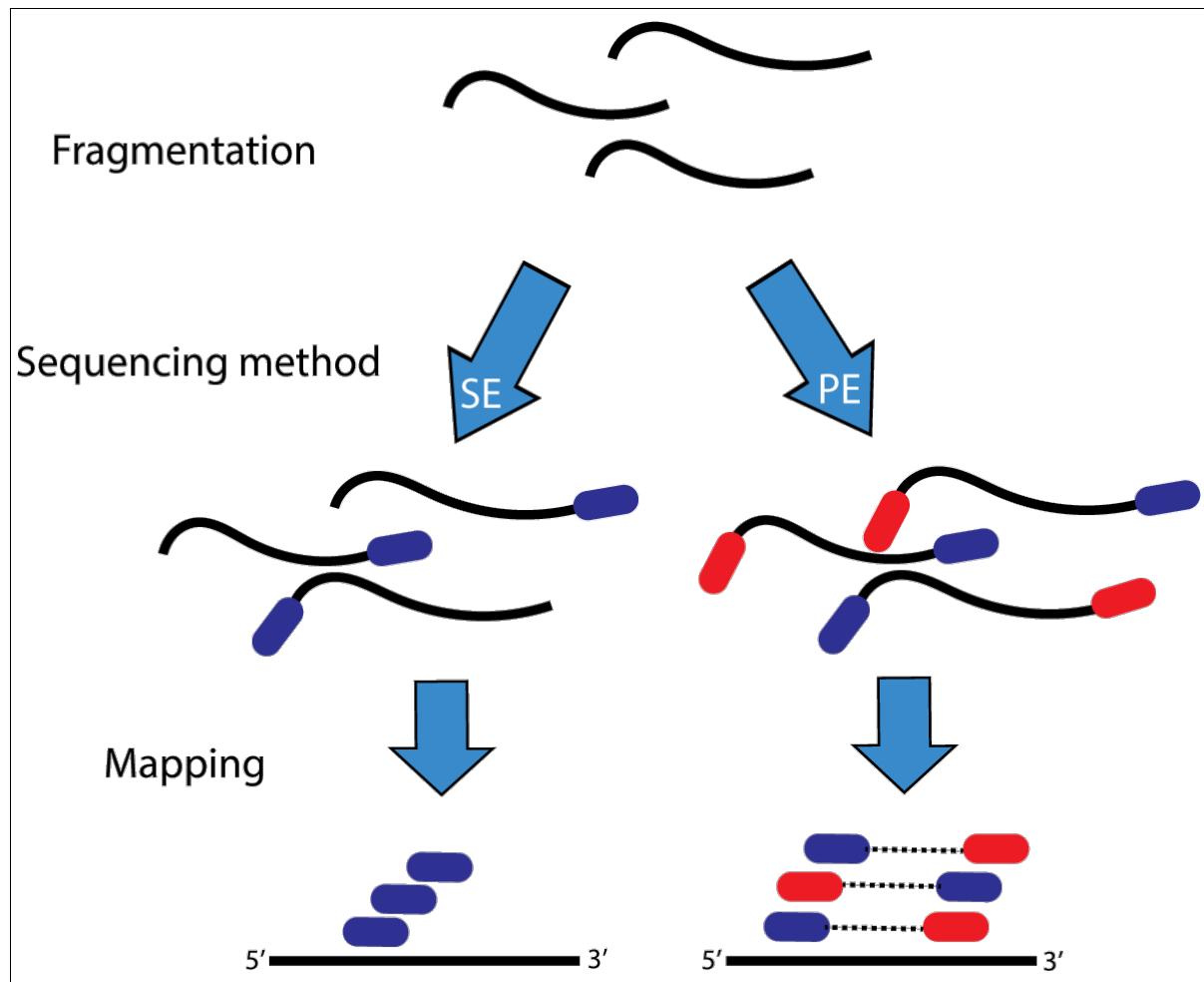


Fig. 3.2: Illustration of single-end (SE) versus paired-end (PE) sequencing.

If you need to refresh how Illumina⁷⁰ paired-end sequencing works have a look at the Illumina technology webpage⁷¹ and this video⁷².

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. Illumina⁷³ adapters have been removed as well already! The process is explained below but we are **not** going to do it.

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.

⁶⁹ <http://illumina.com>

⁷⁰ <http://illumina.com>

⁷¹ http://www.illumina.com/technology/next-generation-sequencing/paired-end-sequencing_assay.html

⁷² <https://youtu.be/HMyCqWhwB8E>

⁷³ <http://illumina.com>

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-page⁷⁴](#)!

A useful tool to decode base qualities can be found [here⁷⁵](#).

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

PhiX⁷⁶ 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 runs⁷⁷](#). 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](#)] (page 85). The steps involve mapping all reads to the “known” PhiX genome, and removing all of those sequence reads from the data.

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 29) section on how to map reads against a reference genome.

⁷⁴ https://en.wikipedia.org/wiki/FASTQ_format

⁷⁵ <http://broadinstitute.github.io/picard/explain-qualities.html>

⁷⁶ https://en.wikipedia.org/wiki/Phi_X_174

⁷⁷ <http://www.illumina.com/products/by-type/sequencing-kits/cluster-gen-sequencing-reagents/phix-control-v3.html>

3.8 Adapter trimming

The process of sequencing DNA via Illumina⁷⁸ 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.

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. Illumina⁷⁹ itself provides a document⁸⁰ that describes the adapters used for their different technologies. Also the FastQC⁸¹ tool, we will be using later on, provides a collection of contaminants and adapters⁸².

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-mcf⁸³ from the ea-utils suite⁸⁴ 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 adapters.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 Quality assessment of sequencing reads (SolexaQA++)

To assess the sequence read quality of the Illumina⁸⁵ run we make use of a program called SolexaQA++⁸⁶ [COX2010] (page 85). SolexaQA++⁸⁷ was originally developed to work with Solexa data (since bought by Illumina⁸⁸), but long since working with Illumina⁸⁹ data. It produces nice graphics that intuitively show the quality of the sequences. it is also able to dynamically trim the bad quality ends off the reads.

From the webpage:

“SolexaQA calculates sequence quality statistics and creates visual representations of data quality for second-generation sequencing data. Originally developed for the Illumina system (historically known as “Solexa”), SolexaQA now also supports Ion Torrent and 454 data.”

3.9.1 Install SolexaQA++

Unfortunately, currently we cannot install SolexaQA++⁹⁰ with conda⁹¹.

⁷⁸ <http://illumina.com>

⁷⁹ <http://illumina.com>

⁸⁰ <https://support.illumina.com/downloads/illumina-customer-sequence-letter.html>

⁸¹ <http://www.bioinformatics.babraham.ac.uk/projects/fastqc/>

⁸² https://github.com/csf-ngs/fastqc/blob/master/Contaminants/contaminant_list.txt

⁸³ <https://github.com/ExpressionAnalysis/ea-utils/blob/wiki/FastqMcf.md>

⁸⁴ <https://expressionanalysis.github.io/ea-utils/>

⁸⁵ <http://illumina.com>

⁸⁶ <http://solexaqa.sourceforge.net>

⁸⁷ <http://solexaqa.sourceforge.net>

⁸⁸ <http://illumina.com>

⁸⁹ <http://illumina.com>

⁹⁰ <http://solexaqa.sourceforge.net>

⁹¹ <http://conda.pydata.org/miniconda.html>

```
curl -O http://compbio.massey.ac.nz/data/203341/SolexaQA.tar.gz

# uncompress the archive
tar -xvzf SolexaQA.tar.gz

# make the file executable
chmod a+x SolexaQA/Linux_x64/SolexaQA++

# copy program to root folder
cp ./SolexaQA/Linux_x64/SolexaQA++ .

# run the program
./SolexaQA++
```

Note: Should the download fail, download manually from *Downloads* (page 79).

3.9.2 SolexaQA++ manual

SolexaQA++⁹² has three modes that can be run. Type:

```
./SolexaQA++
```

```
SolexaQA++ v3.1.3
Released under GNU General Public License version 3
C++ version developed by Mauro Truglio (M.Truglio@massey.ac.nz)

Usage: SolexaQA++ <command> [options]

Command: analysis      quality analysis and graphs generation
          dynamictrim   trim reads using a chosen threshold
          lengthsort    sort reads by a chosen length
```

The three modes are: `analysis`, `dynamictrim`, and `lengthsort`:

`analysis` - the primary quality analysis and visualization tool. Designed to run on unmodified FASTQ files obtained directly from Illumina⁹³, Ion Torrent or 454 sequencers.

`dynamictrim` - a read trimmer that individually crops each read to its longest contiguous segment for which quality scores are greater than a user-supplied quality cutoff.

`lengthsort` - a program to separate high quality reads from low quality reads. LengthSort assigns trimmed reads to paired-end, singleton and discard files based on a user-defined length cutoff.

3.9.3 SolexaQA++ dynamic trimming

We will use SolexaQA++⁹⁴ dynamic trim the reads, to chop off nucleotides with a bad quality score.

Todo:

1. Create a directory for the result-files -> `trimmed/`.
2. Run SolexaQA++⁹⁵ `dynamictrim` with the untrimmed data and a probability cutoff of 0.05., and submit result-directory `trimmed/`.
3. Investigate the result-files in `trimmed/`, e.g. do the file-sizes change to the original files?

⁹² <http://solexaqa.sourceforge.net>

⁹³ <http://illumina.com>

⁹⁴ <http://solexaqa.sourceforge.net>

⁹⁵ <http://solexaqa.sourceforge.net>

-
4. SolexaQA++⁹⁶ dynamictrim produces a graphical output. Explain what the graph shows. Find help on the SolexaQA++⁹⁷ website.
-

Hint: Should you not get 1 and/or 2 right, try the commands in *Code: SolexaQA++ trimming* (page 77).

3.9.4 SolexaQA++ analysis on trimmed data

Todo:

1. Create a directory for the result-files -> **trimmed-solexaqa**.
 2. Use SolexaQA++⁹⁸ to do the quality assessment with the trimmed data-set.
 3. Compare your results to the examples of a particularly bad MiSeq run (Fig. 3.6 to Fig. 3.6, taken from SolexaQA++⁹⁹ website). Write down your observations.
 4. What elements in these example figures (Fig. 3.3 to Fig. 3.6) indicate that they show a bad run? Write down your explanations.
-

Hint: Should you not get 1 and/or 2 right, try the commands in *Code: SolexaQA++ qc* (page 77).

3.10 Sickle for dynamic trimming (alternative to SolexaQA++)

Should the dynamic trimming not work with SolexaQA++¹⁰⁰, you can alternatively use Sickle¹⁰¹.

```
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

# sickle parameters:
sickle --help

# as we are dealing with paired-end data you will be using "sickle pe"
sickle pe --help

# run sickle like so:
sickle pe -g 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
```

Hint: Should you be unable to run Sickle¹⁰² or SolexaQA++¹⁰³ at all to trim the data. You can download

⁹⁶ <http://solexaqa.sourceforge.net>

⁹⁷ <http://solexaqa.sourceforge.net>

⁹⁸ <http://solexaqa.sourceforge.net>

⁹⁹ <http://solexaqa.sourceforge.net>

¹⁰⁰ <http://solexaqa.sourceforge.net>

¹⁰¹ <https://github.com/najoshi/sickle>

¹⁰² <https://github.com/najoshi/sickle>

¹⁰³ <http://solexaqa.sourceforge.net>

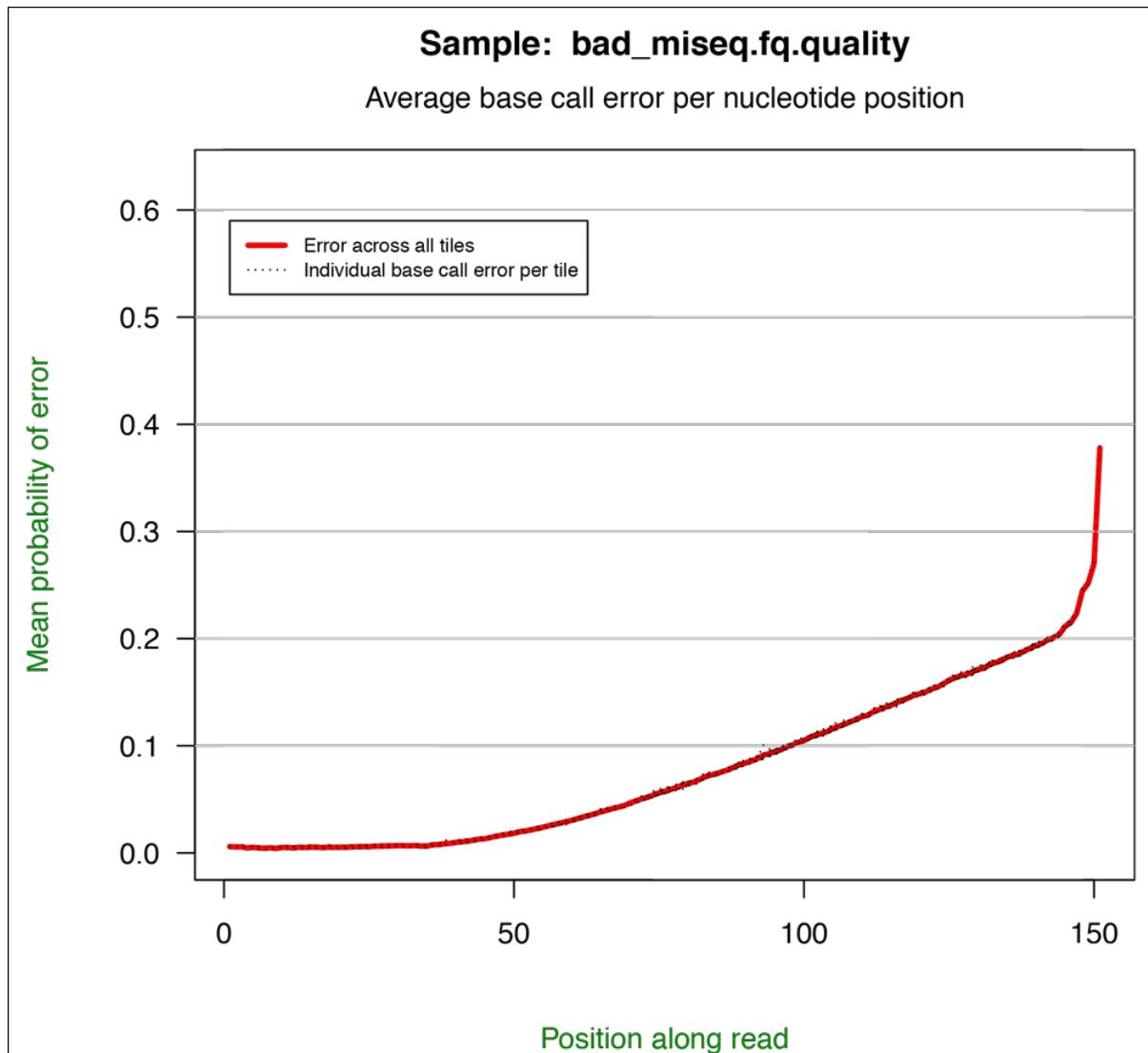


Fig. 3.3: SolexaQA++ example quality plot along reads of a bad MiSeq run

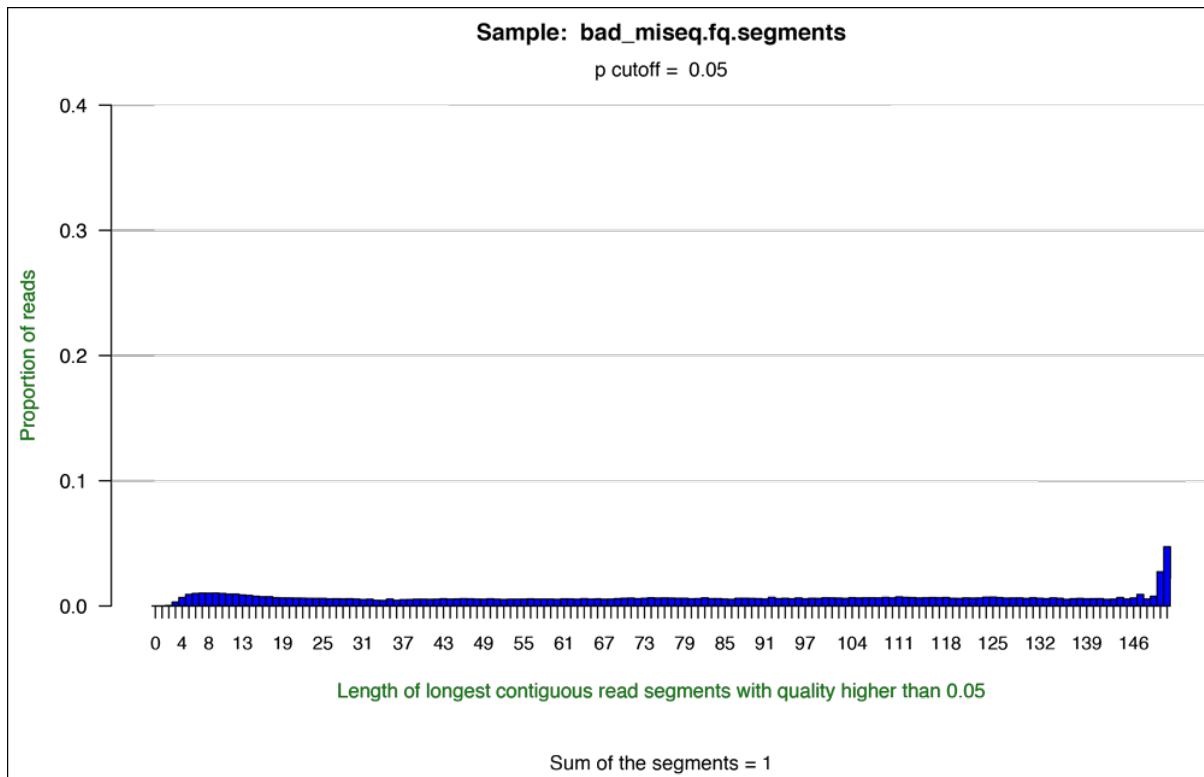


Fig. 3.4: SolexaQA++ example histogram plot of a bad MiSeq run.

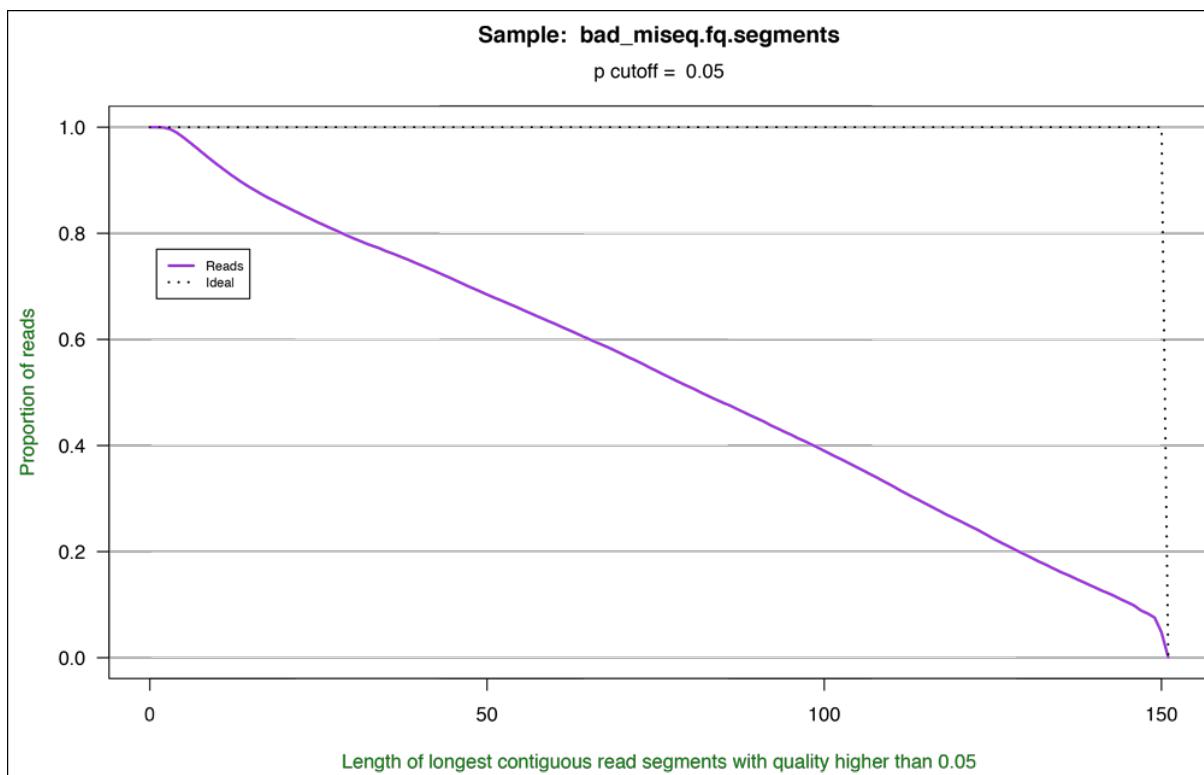


Fig. 3.5: SolexaQA++ example cumulative plot of a bad MiSeq run.

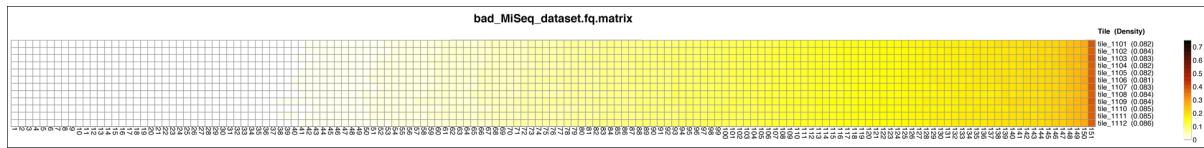


Fig. 3.6: SolexaQA++ example quality heatmap of a bad MiSeq run.

the trimmed dataset [here¹⁰⁴](#). Unarchive and uncompress the files with `tar -xvzf trimmed.tar.gz`.

3.11 Quality assessment of sequencing reads (FastQC)

3.11.1 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
    data.

    If no files to process are specified on the command line then the program
    will start as an interactive graphical application. If files are provided
    on the command line then the program will run with no user interaction
    required. In this mode it is suitable for inclusion into a standardised
    analysis pipeline.
```

3.11.2 FastQC manual

FastQC¹⁰⁵ is a very simple program to run that provides similar and additional information to SolexaQA++¹⁰⁶.

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:

¹⁰⁴ <http://compbio.massey.ac.nz/data/203341/trimmed.tar.gz>

¹⁰⁵ <http://www.bioinformatics.babraham.ac.uk/projects/fastqc/>

¹⁰⁶ <http://solexaqa.sourceforge.net>

```
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 FastQC¹⁰⁷ made some nice help pages explaining each of the plots and results you expect to see [here](#)¹⁰⁸.

3.11.3 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 [FastQC](#)¹⁰⁹ website and read about sequencing QC reports for good and bad Illumina¹¹⁰ sequencing runs.
4. Compare your results to these examples ([Fig. 3.7](#) to [Fig. 3.9](#)) of a particularly bad run (taken from the [FastQC](#)¹¹¹ website) and write down your observations with regards to your data.
5. What elements in these example figures ([Fig. 3.7](#) to [Fig. 3.9](#)) indicate that the example is from a bad run?

Hint: Should you not get it right, try the commands in [Code: FastQC](#) (page 77).

¹⁰⁷ <http://www.bioinformatics.babraham.ac.uk/projects/fastqc/>

¹⁰⁸ <http://www.bioinformatics.babraham.ac.uk/projects/fastqc/Help/3%20Analysis%20Modules/>

¹⁰⁹ <http://www.bioinformatics.babraham.ac.uk/projects/fastqc/>

¹¹⁰ <http://illumina.com>

¹¹¹ <http://www.bioinformatics.babraham.ac.uk/projects/fastqc/>

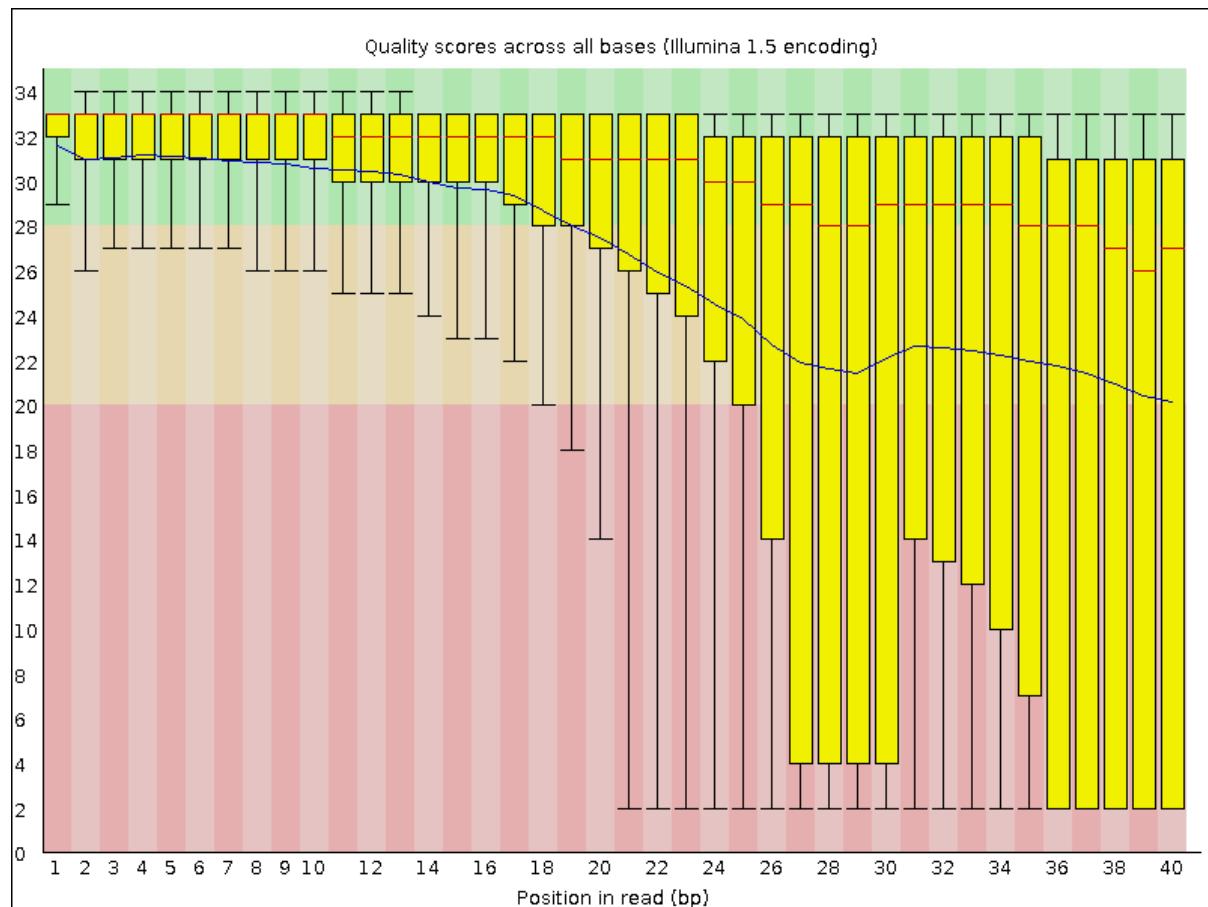


Fig. 3.7: Quality score across bases.

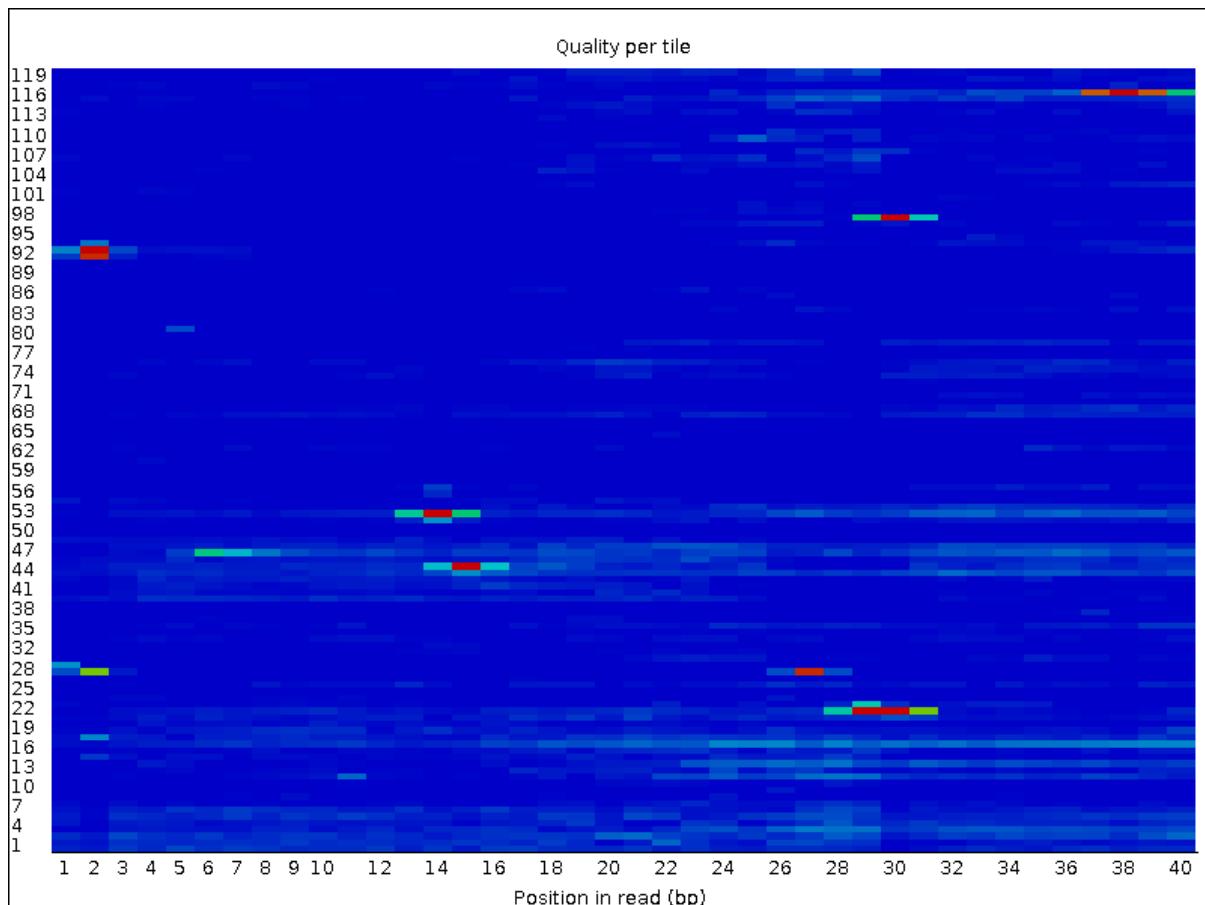


Fig. 3.8: Quality per tile.

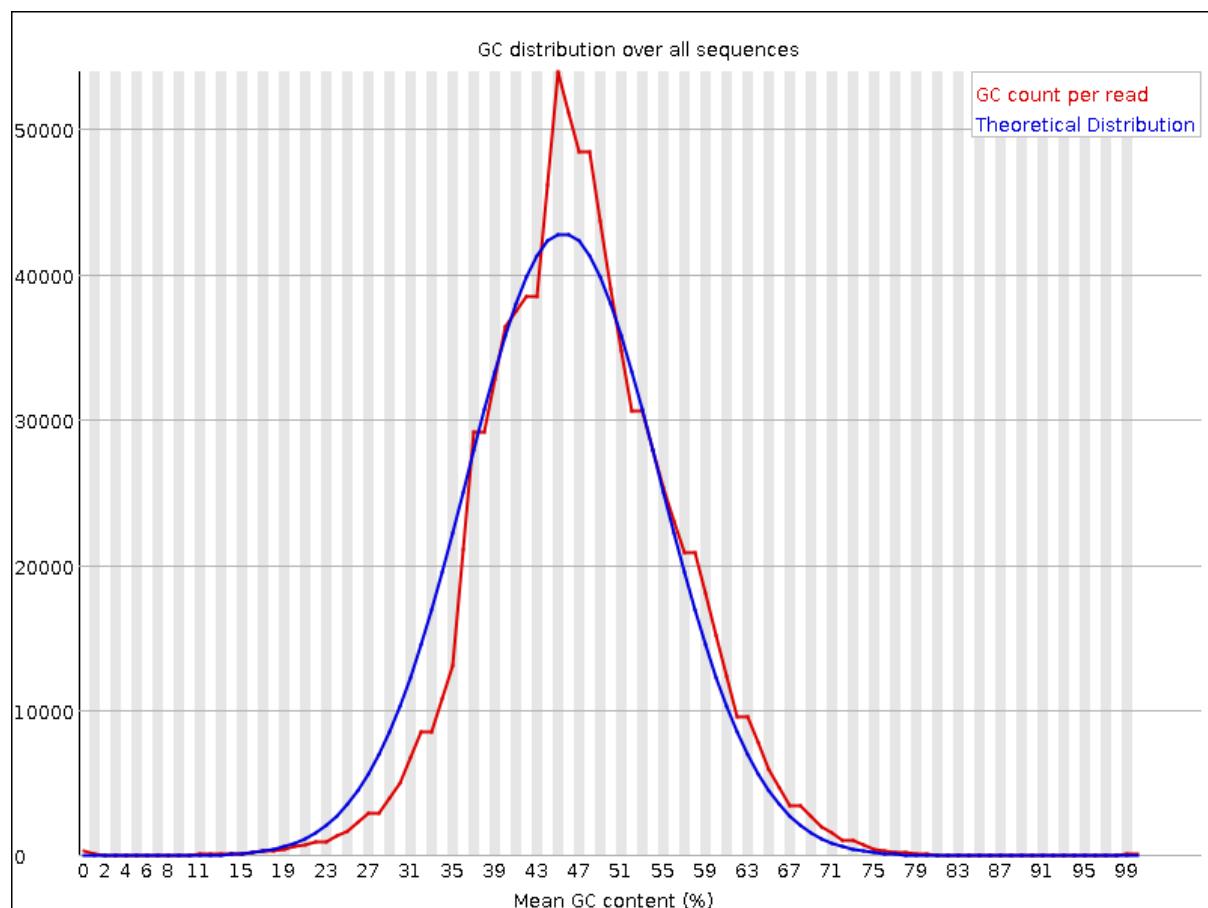


Fig. 3.9: GC distribution over all sequences.

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/
SolexaQA/
SolexaQA++
trimmed/
trimmed-fastqc/
trimmed-solexaqa/
```

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
```

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

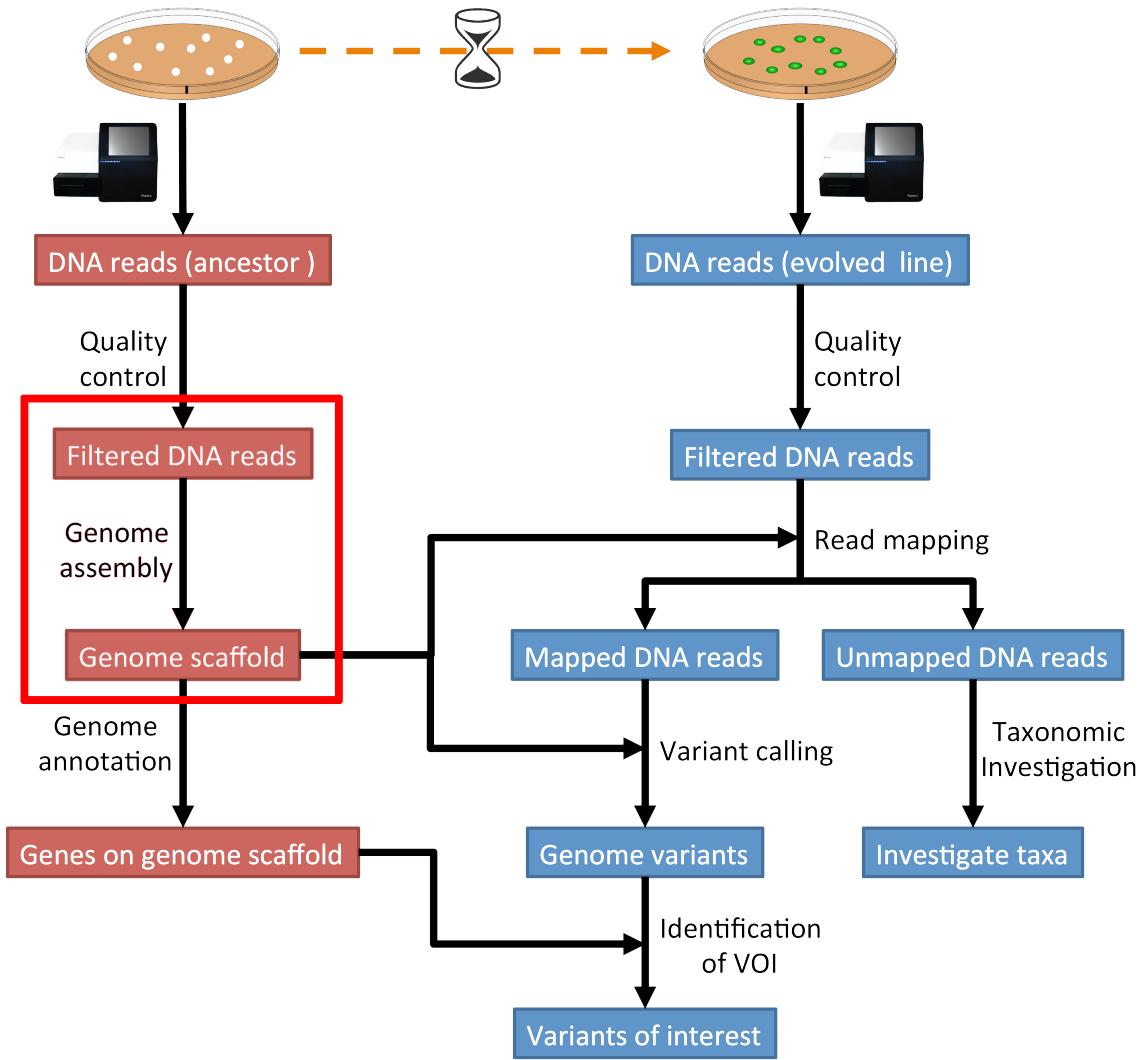


Fig. 4.1: The part of the workflow we will work on in this section marked in red.

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

# sub sample reads
seqtk sample -s11 trimmed/ancestor-R1.fastq.trimmed.gz 0.1 | gzip > sampled/ancestor-R1.fastq.
˓trimmed.gz
seqtk sample -s11 trimmed/ancestor-R1.fastq.trimmed.gz 0.1 | gzip > sampled/ancestor-R1.fastq.
˓trimmed.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 `SPAdes`¹⁴⁵ 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 `SPAdes`¹⁴⁶ fo assembling our genome. In a recent evaluation of assembly software, `SPAdes`¹⁴⁷ was found to be a good choice for fungal genomes [ABBAS2014] (page 85). 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
mkdir assembly

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

¹⁴⁵ <http://bioinf.spbau.ru/spades>

¹⁴⁶ <http://bioinf.spbau.ru/spades>

¹⁴⁷ <http://bioinf.spbau.ru/spades>

The two files we need to submit to [SPAdes](#)¹⁴⁸ are two paired-end read files.

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

Todo:

1. Run [SPAdes](#)¹⁴⁹ with default parameters on the ancestor
 2. Read in the [SPAdes](#)¹⁵⁰ manual about assembling with 2x150bp reads
 3. Run [SPAdes](#)¹⁵¹ a second time but use the options suggested at the [SPAdes](#)¹⁵² manual section 3.4¹⁵³ 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 78).

4.6 Assembly quality assessment

4.6.1 Assembly statistics

[Quast](#)¹⁵⁴ (QQuality ASsesment Tool) [[GUREVICH2013](#)] (page 85), 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 contig bases
- genes and operons covered

It is easy with [Quast](#)¹⁵⁵ to compare these measures among several assemblies. The program can be used on their [website](#)¹⁵⁶.

```
conda install quast
```

Run [Quast](#)¹⁵⁷ with both assembly scaffolds.fasta files to compare the results.

Note: Should you be unable to run [SPAdes](#)¹⁵⁸ on the data, you can manually download the assembly from [Downloads](#) (page 79). 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
```

¹⁴⁸ <http://bioinf.spbau.ru/spades>

¹⁴⁹ <http://bioinf.spbau.ru/spades>

¹⁵⁰ <http://bioinf.spbau.ru/spades>

¹⁵¹ <http://bioinf.spbau.ru/spades>

¹⁵² <http://bioinf.spbau.ru/spades>

¹⁵³ <http://spades.bioinf.spbau.ru/release3.9.1/manual.html#sec3.4>

¹⁵⁴ <http://quast.bioinf.spbau.ru/>

¹⁵⁵ <http://quast.bioinf.spbau.ru/>

¹⁵⁶ <http://quast.bioinf.spbau.ru/>

¹⁵⁷ <http://quast.bioinf.spbau.ru/>

¹⁵⁸ <http://bioinf.spbau.ru/spades>

Todo:

1. Compare the results of [Quast¹⁵⁹](#) 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 [Quast¹⁶⁰](#) 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 78).

4.8 Assemblathon

Todo: Now that you know the basics for assembling a genome and judging their quality, play with the [SPAdes¹⁶¹](#) 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 [Quast¹⁶²](#)
-

4.9 Further reading

4.9.1 Background on Genome Assemblies

- How to apply de Bruijn graphs to genome assembly. [\[COMPEAU2011\]](#) (page 85)
- Sequence assembly demystified. [\[NAGARAJAN2013\]](#) (page 85)

4.9.2 Evaluation of Genome Assembly Software

- GAGE: A critical evaluation of genome assemblies and assembly algorithms. [\[SALZBERG2012\]](#) (page 85)
- Assessment of de novo assemblers for draft genomes: a case study with fungal genomes. [\[ABBAS2014\]](#) (page 85)

¹⁵⁹ <http://quast.bioinf.spbau.ru/>

¹⁶⁰ <http://quast.bioinf.spbau.ru/>

¹⁶¹ <http://bioinf.spbau.ru/spades>

¹⁶² <http://quast.bioinf.spbau.ru/>

4.10 Web links

- Lectures for this topic: [Genome Assembly: An Introduction](#)¹⁶³
- [SPAdes](#)¹⁶⁴
- [Quast](#)¹⁶⁵
- [Bandage](#)¹⁶⁶ (Bioinformatics Application for Navigating De novo Assembly Graphs Easily) is a program that visualizes a genome assembly as a graph [\[WICK2015\]](#) (page 86).

¹⁶³ <https://dx.doi.org/10.6084/m9.figshare.2972323.v1>

¹⁶⁴ <http://bioinf.spbau.ru/spades>

¹⁶⁵ <http://quast.bioinf.spbau.ru/>

¹⁶⁶ <https://rrwick.github.io/Bandage/>

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/
SolexaQA/
SolexaQA++
trimmed/
trimmed-fastqc/
trimmed-solexaqa/
```

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 23). We are going to use the quality trimmed forward and backward DNA

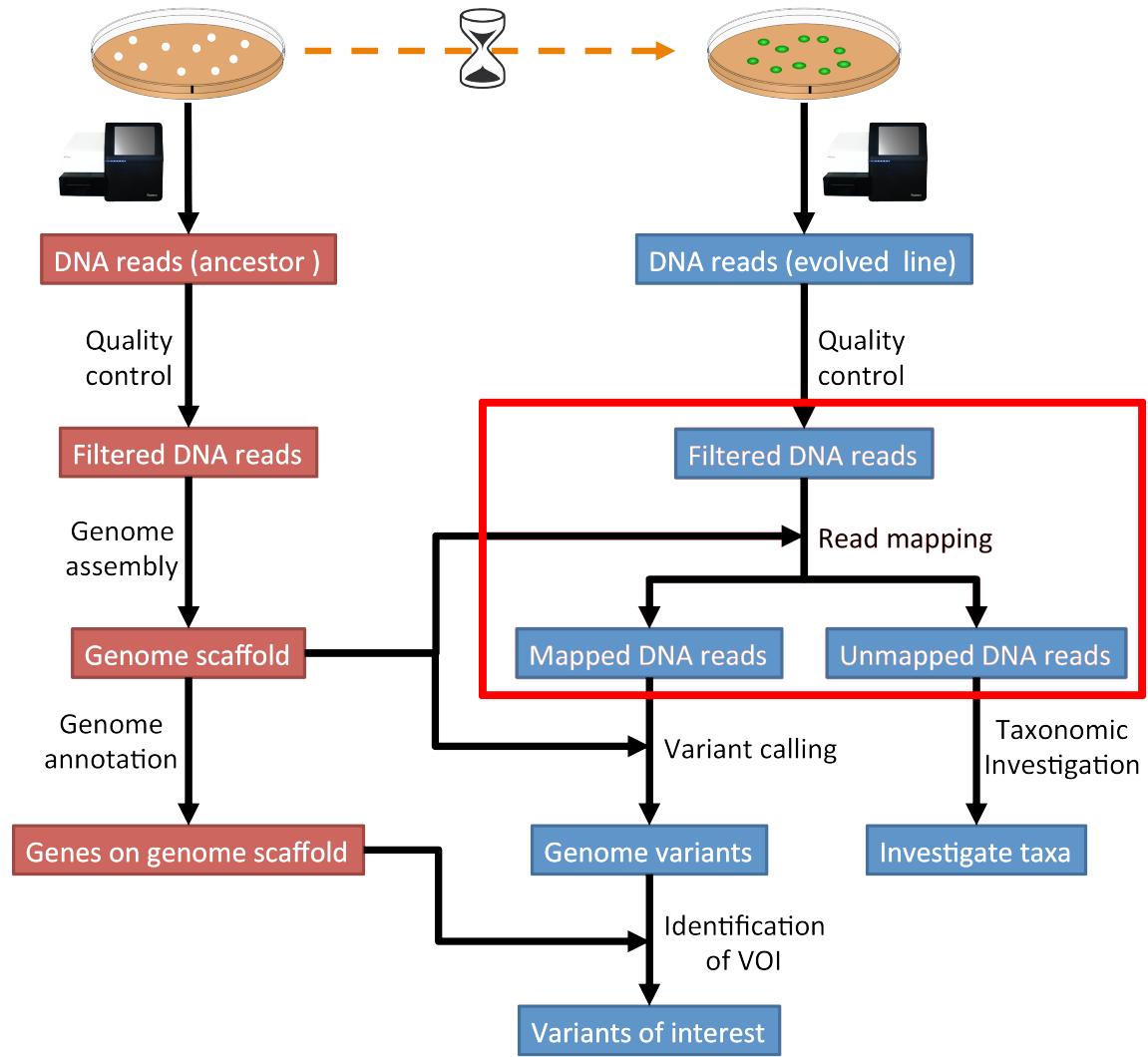


Fig. 5.1: The part of the workflow we will work on in this section marked in red.

sequences of the evolved line and use a program called [Bowtie2](#)²⁰¹ 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 [Bowtie2](#)²⁰² to map our reads to a genome.

It is simple to install and use.

```
conda activate ngs
conda install samtools
conda install bamtools
conda install bedtools
conda install bowtie2
conda install bwa
```

5.6 Bowtie2

5.6.1 Overview

[Bowtie2](#)²⁰³ is a short read aligner, that can take a reference genome and map single- or paired-end data to it [\[TRAPNELL2009\]](#) (page 86). It requires an indexing step in which one supplies the reference genome and [Bowtie2](#)²⁰⁴ 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 [Bowtie2](#)²⁰⁵ tools we are going to use are shown below:

```
# 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 [Bowtie2](#)²⁰⁶ default is set to 500 which is often considered too short²⁰⁷.

5.6.2 Creating a reference index for mapping

Todo: Create an [Bowtie2](#)²⁰⁸ index for our reference genome assembly. Attention! Remember which file you need to submit to [Bowtie2](#)²⁰⁹.

²⁰¹ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁰² <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁰³ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁰⁴ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁰⁵ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁰⁶ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁰⁷ <http://lab.loman.net/2013/05/02/use-x-with-bowtie2-to-set-minimum-and-maximum-insert-sizes-for-nexera-libraries/>

²⁰⁸ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁰⁹ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

Hint: Should you not get it right, try the commands in [Code: Bowtie2 indexing](#) (page 78).

Note: Should you be unable to run [Bowtie2²¹⁰](#) indexing on the data, you can download the index from [Downloads](#) (page 79). Unarchive and uncompress the files with `tar -xvzf bowtie2-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 `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 78).

Note: [Bowtie2²¹¹](#) 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!

5.7 BWA

Attention: If the mapping did not succeed with [Bowtie2²¹²](#). We can use the aligner [BWA²¹³](#) explained in this section. If the mapping with [Bowtie2²¹⁴](#) did work, you can jump this section. You can jump straight ahead to [Section 5.8](#).

5.7.1 Overview

[BWA²¹⁵](#) is a short read aligner, that can take a reference genome and map single- or paired-end data to it. It requires an indexing step in which one supplies the reference genome and [BWA²¹⁶](#) 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 [BWA²¹⁷](#) tools we are going to use are shown below:

```
# bwa index help  
bwa index  
  
# indexing  
bwa index reference-genome.fa  
  
# bwa mem help  
bwa mem
```

(continues on next page)

²¹⁰ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²¹¹ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²¹² <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²¹³ <http://bio-bwa.sourceforge.net/>

²¹⁴ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²¹⁵ <http://bio-bwa.sourceforge.net/>

²¹⁶ <http://bio-bwa.sourceforge.net/>

²¹⁷ <http://bio-bwa.sourceforge.net/>

(continued from previous page)

```
# single-end mapping
bwa mem reference-genome.fa reads.fq > aln-se.sam

# paired-end mapping
bwa mem reference-genome.fa read1.fq read2.fq > aln-pe.sam
```

5.7.2 Creating a reference index for mapping

Todo: Create an [BWA²¹⁸](#) index for our reference genome assembly. Attention! Remember which file you need to submit to [BWA²¹⁹](#).

Hint: Should you not get it right, try the commands in [Code: BWA indexing](#) (page 78).

Note: Should you be unable to run [BWA²²⁰](#) indexing on the data, you can download the index from [Downloads](#) (page 79). Unarchive and uncompress the files with tar -xvzf bwa-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 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 78).

5.8 The sam mapping file-format

[Bowtie2²²¹](#) and [BWA²²²](#) 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 [here²²³](#) and even more information can be found [here²²⁴](#). 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](#).

²¹⁸ <http://bio-bwa.sourceforge.net/>

²¹⁹ <http://bio-bwa.sourceforge.net/>

²²⁰ <http://bio-bwa.sourceforge.net/>

²²¹ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²²² <http://bio-bwa.sourceforge.net/>

²²³ <http://bio-bwa.sourceforge.net/bwa.shtml#4>

²²⁴ <http://samtools.github.io/hts-specs/SAMv1.pdf>

Table 5.1: The sam-file format fields.

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

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 flag²²⁵](#) information on SAM records, it is helpful when working with many tools to first clean up read pairing information and flags with [SAMtools²²⁶](#). 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.
 - **-O bam**: specifies that we want compressed bam output from fixmate

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` flag²²⁷ information later below to extract specific alignments.

Hint: A very useful tools to explain flags can be found [here](#)²²⁸.

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

²²⁵ <http://bio-bwa.sourceforge.net/bwa.shtml#4>

226 http://samtools.sourceforge.net/

²²⁷ <http://bio-bwa.sourceforge.net/bwa.shtml#4>

228 <http://broadinstitute.github.io/picard/explain-flags.html>

```
rm mappings/evolved-6.sam
```

5.9.2 Sorting

We are going to use SAMtools²²⁹ again to sort the bam-file into **coordinate order**:

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

- -o: specifies the name of the output file.
- -O 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 subsequently 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 79).

5.10 Mapping statistics

5.10.1 Stats with SAMtools

Lets get an mapping overview:

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

Todo: Look at the mapping statistics and understand their meaning²³⁰. 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 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.

²²⁹ <http://samtools.sourceforge.net/>

²³⁰ <https://www.biostars.org/p/12475/>

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

Now we quickly use some R²³¹ to make a coverage plot for contig NODE20. Open a R²³² 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='position', 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='position', ylab='coverage')
dev.off()
```

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

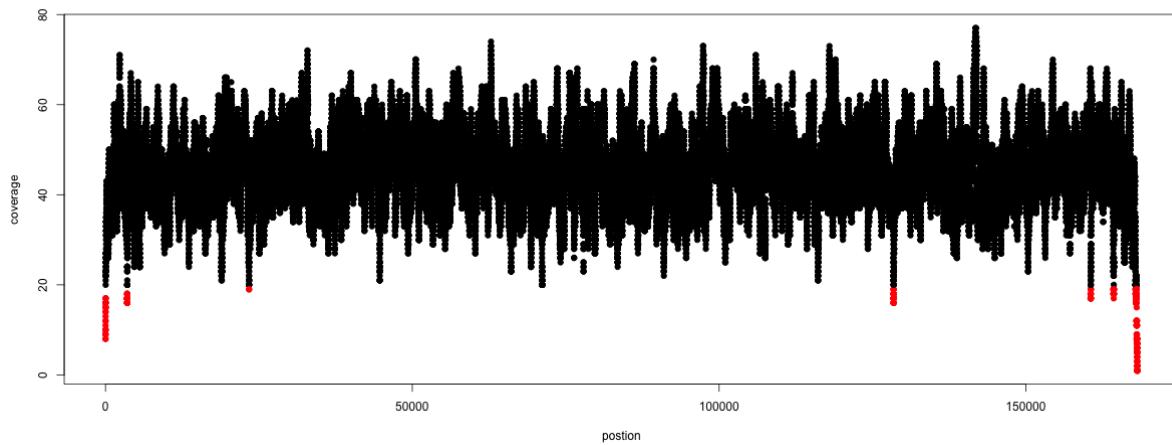


Fig. 5.2: A example coverage plot for a contig with highlighted in red regions with a coverage below 20 reads.

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 QualiMap²³³.

QualiMap²³⁴ 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.

²³¹ <https://www.r-project.org/>

²³² <https://www.r-project.org/>

²³³ <http://qualimap.bioinfo.cipf.es/>

²³⁴ <http://qualimap.bioinfo.cipf.es/>

Installation:

```
conda install qualimap
```

Run QualiMap²³⁵ with:

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

This will create a report in the mapping folder. See this [webpage²³⁶](#) to get help on the sections in the report.

Todo: Install QualiMap²³⁷ 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 Bowtie2²³⁸ 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 SAMtools²³⁹.

Todo: Explain what concordant and discordant read pairs are? Look at the [Bowtie2²⁴⁰](#) 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).

```
samtools view -h -b -f 3 mappings/evolved-6.sorted.dedup.bam > mappings/evolved-6.sorted.dedup.  
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 flag²⁴¹](#) 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: $MAPQ = -10 * \log_{10}(p)$, where p is the probability that the

²³⁵ <http://qualimap.bioinfo.cipf.es/>

²³⁶ http://qualimap.bioinfo.cipf.es/doc_html/analysis.html#output

²³⁷ <http://qualimap.bioinfo.cipf.es/>

²³⁸ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²³⁹ <http://samtools.sourceforge.net/>

²⁴⁰ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁴¹ <http://bio-bwa.sourceforge.net/bwa.shtml#4>

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 [here²⁴²](https://sequencing.qcfail.com/articles/mapq-values-are-really-useful-but-their-implementation-is-a-mess/). Bottom-line is that we need to be aware that different tools use this value in different ways and 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*²⁴³:

```
 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 [link²⁴⁴](#) 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 *QualiMap*²⁴⁵ 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 *Kraken2*²⁴⁶ like in section *Taxonomic investigation* (page 39) 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:

```
 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 flag²⁴⁷](#) 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.

```
 bamToFastq -i mappings/evolved-6.sorted.unmapped.bam -fq mappings/evolved-6.sorted.unmapped.R1.  
 ↵fastq -fq2 mappings/evolved-6.sorted.unmapped.R2.fastq
```

²⁴² <https://sequencing.qcfail.com/articles/mapq-values-are-really-useful-but-their-implementation-is-a-mess/>

²⁴³ <http://bowtie-bio.sourceforge.net/bowtie2/index.shtml>

²⁴⁴ <https://sequencing.qcfail.com/articles/mapq-values-are-really-useful-but-their-implementation-is-a-mess/>

²⁴⁵ <http://qualimap.bioinfo.cipf.es/>

²⁴⁶ <https://www.ccb.jhu.edu/software/kraken2/>

²⁴⁷ <http://bio-bwa.sourceforge.net/bwa.shtml#4>

TAXONOMIC INVESTIGATION

6.1 Preface

We want to investigate 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 [Kraken2²⁷⁷](#) 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/
SolexaQA/
SolexaQA++
trimmed/
trimmed-fastqc/
trimmed-solexaqa/
```

6.4 Kraken2

We will be using a tool called [Kraken2²⁷⁸](#) [[WOOD2014](#)] (page 86). This tool uses k-mers to assign a taxonomic labels in form of [NCBI Taxonomy²⁷⁹](#) 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

²⁷⁷ <https://wwwccb.jhu.edu/software/kraken2/>

²⁷⁸ <https://wwwccb.jhu.edu/software/kraken2/>

²⁷⁹ <https://www.ncbi.nlm.nih.gov/taxonomy>

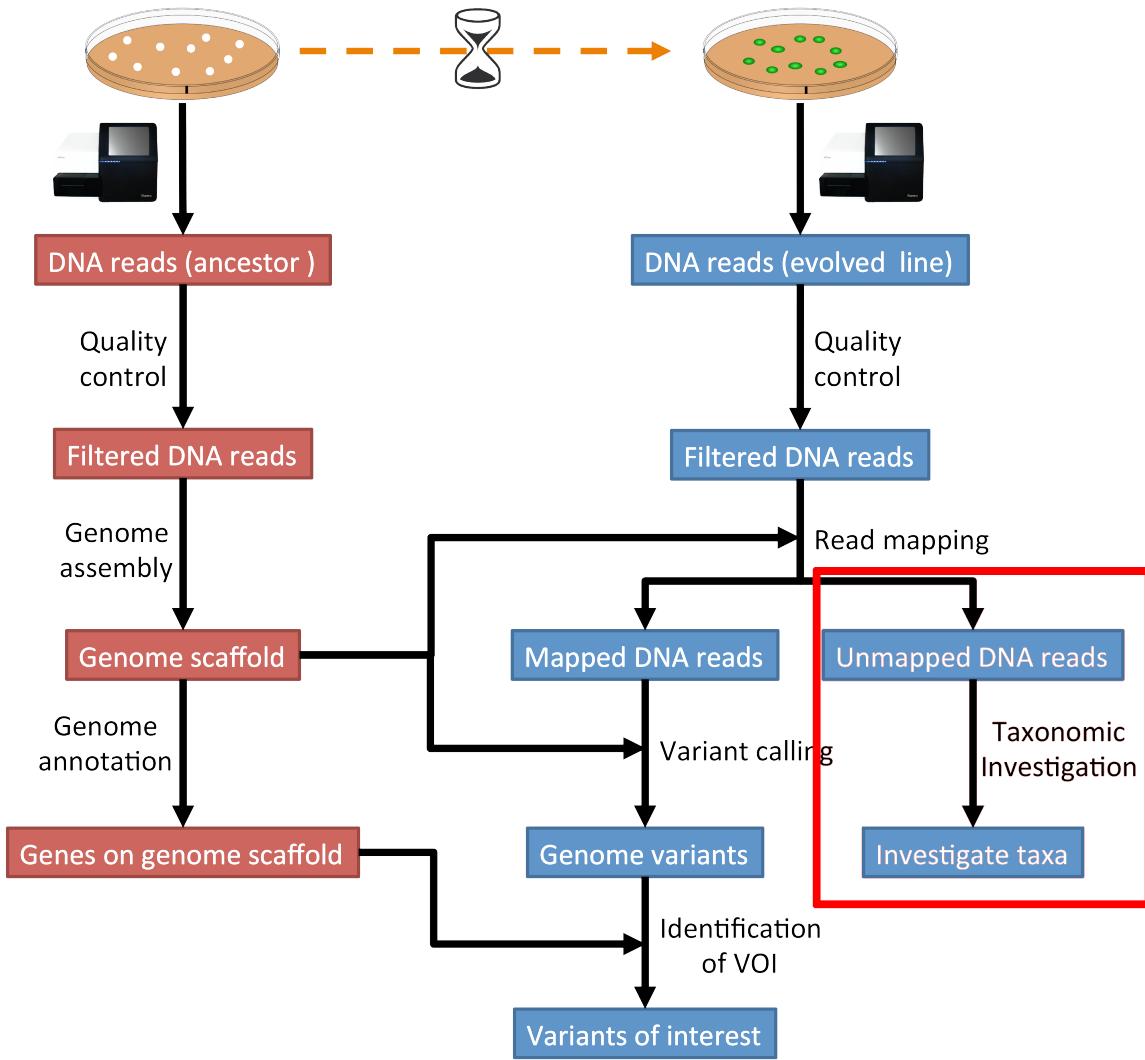


Fig. 6.1: The part of the workflow we will work on in this section marked in red.

label. If the k-mer 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 Kraken2²⁸⁰. 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 Kraken2²⁸¹ database that can be used to assign the taxonomic labels to sequences. We opt for downloading the pre-build “minikraken” database from the Kraken2²⁸² website:

```
curl -O 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 79) 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 Kraken2²⁸³ 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 Kraken2²⁸⁴ 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
```

However, we may have fastq-files, so we need to use --fastq-input which tells Kraken2²⁸⁵ that it is dealing with fastq-formated files. In addition, we are dealing with paired-end data, which we can tell

²⁸⁰ <https://www.ccb.jhu.edu/software/kraken2/>

²⁸¹ <https://www.ccb.jhu.edu/software/kraken2/>

²⁸² <https://www.ccb.jhu.edu/software/kraken2/>

²⁸³ <https://www.ccb.jhu.edu/software/kraken2/>

²⁸⁴ <https://www.ccb.jhu.edu/software/kraken2/>

²⁸⁵ <https://www.ccb.jhu.edu/software/kraken2/>

Kraken2²⁸⁶ with the switch --paired. Here, we are investigating one of the unmapped paired-end read files of the evolved line.

```
kraken2 --use-names --threads 4 --db minikraken2_v2_8GB --fastq-input --report evolved-6 --paired ..  
→ /mappings/evolved-6.sorted.unmapped.R1.fastq ./mappings/evolved-6.sorted.unmapped.R2.fastq >  
→ evolved-6.kraken
```

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:

C	7001326F:121:CBVVLANXX:1:1105:2240:12640	816	251	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
C	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
U	7001326F:121:CBVVLANXX:1:1105:11030:12689	0	251	0:91 A:35 0:91
U	7001326F:121:CBVVLANXX:1:1105:7157:12806	0	206	0:69 A:35 0:68

Each sequence classified by Kraken2²⁸⁷ results in a single line of output. Output lines contain five tab-delimited 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 Kraken2²⁸⁸ used to label the sequence; this is 0 if the sequence is unclassified and otherwise should be the NCBI Taxonomy²⁸⁹ 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 Kraken2²⁹⁰ manual can be accessed [here²⁹¹](#).

6.4.3 Investigate taxa

We can use the webpage NCBI TaxIdentifier²⁹² to quickly get the names to the taxonomy identifier. However, this is impractical as we are dealing potentially with many sequences. Kraken2²⁹³ has some scripts that help us understand our results better.

Because we used the Kraken2²⁹⁴ 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.

²⁸⁶ <https://wwwccb.jhu.edu/software/kraken2/>

²⁸⁷ <https://wwwccb.jhu.edu/software/kraken2/>

²⁸⁸ <https://wwwccb.jhu.edu/software/kraken2/>

²⁸⁹ <https://www.ncbi.nlm.nih.gov/taxonomy>

²⁹⁰ <https://wwwccb.jhu.edu/software/kraken2/>

²⁹¹ <https://wwwccb.jhu.edu/software/kraken2/index.shtml?t=manual>

²⁹² https://www.ncbi.nlm.nih.gov/Taxonomy/TaxIdentifier/tax_identifier.cgi

²⁹³ <https://wwwccb.jhu.edu/software/kraken2/>

²⁹⁴ <https://wwwccb.jhu.edu/software/kraken2/>

83.56	514312	514312	U	0	unclassified
16.44	101180	0	R	1	root
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	0	D2	68336	Bacteroidetes/Chlorobi group
13.99	86103	8	P	976	Bacteroidetes
13.94	85798	2	C	200643	Bacteroidia
13.94	85789	19	O	171549	Bacteroidales
13.87	85392	0	F	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 Taxonomy²⁹⁵ 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 create 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 Taxonomy²⁹⁶ id.

6.4.4 Bracken

Bracken²⁹⁷ stands for Bayesian Reestimation of Abundance with KrakenEN, and is a statistical method that computes the abundance of species in DNA sequences from a metagenomics sample [LU2017] (page 86). Bracken²⁹⁸ uses the taxonomy labels assigned by Kraken2²⁹⁹ (see above) to estimate the number of reads originating from each species present in a sample. Bracken³⁰⁰ classifies reads to the best matching location in the taxonomic tree, but does not estimate abundances of species. Combined with the Kraken classifier, Bracken³⁰¹ will produce more accurate species- and genus-level abundance estimates than Kraken2³⁰² alone.

The use of Bracken³⁰³ subsequent to Kraken2³⁰⁴ is optional but might improve on the Kraken2³⁰⁵ results.

²⁹⁵ <https://www.ncbi.nlm.nih.gov/taxonomy>

²⁹⁶ <https://www.ncbi.nlm.nih.gov/taxonomy>

²⁹⁷ <https://ccb.jhu.edu/software/bracken/index.shtml>

²⁹⁸ <https://ccb.jhu.edu/software/bracken/index.shtml>

²⁹⁹ <https://www.ccb.jhu.edu/software/kraken2/>

³⁰⁰ <https://ccb.jhu.edu/software/bracken/index.shtml>

³⁰¹ <https://ccb.jhu.edu/software/bracken/index.shtml>

³⁰² <https://www.ccb.jhu.edu/software/kraken2/>

³⁰³ <https://ccb.jhu.edu/software/bracken/index.shtml>

³⁰⁴ <https://www.ccb.jhu.edu/software/kraken2/>

³⁰⁵ <https://www.ccb.jhu.edu/software/kraken2/>

Installation

We installed Bracken³⁰⁶ already together with Kraken2³⁰⁷ above, so it should be ready to be used.

However, we need to download some additional files for the Kraken2³⁰⁸ 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 Bracken³⁰⁹ on the Kraken2³¹⁰ results to improve them.

The general structure of the Bracken³¹¹ command look like this:

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

- `-l S`: denotes the level we want to look at. S stands for species but other levels are available.
- `-d DB`: specifies the Kraken2³¹² database that should be used.

Let us apply Bracken³¹³ to the example above:

```
bracken -d minikraken2_v2_8GB -i evolved-6.kraken -l 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
Neorhizobium sp. NCHU2750	1825976	S	0	0	0	0.00000
Pseudomonas sp. MT-1	150396	S	0	0	0	0.00000
Ahniella affigens	2021234	S	1	0	1	0.00000
Sinorhizobium sp. CCBAU	05631	794846	S	0	0	0.00000
Cohnella sp. 18JY8-7	2480923	S	1	0	1	0.00000
Bacillus velezensis	492670	S	4	4	8	0.00002
Actinoplanes missouriensis	1866	S	2	8	10	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 Centrifuge³¹⁴ [KIM2017] (page 86). This tool uses a novel indexing scheme based on the Burrows-Wheeler transform (BWT) and the Ferragina-Manzini (FM) index, optimized specifically for the metagenomic classification problem to assign a taxonomic labels in form of NCBI Taxonomy³¹⁵ 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.

³⁰⁶ <https://ccb.jhu.edu/software/bracken/index.shtml>

³⁰⁷ <https://www.ccb.jhu.edu/software/kraken2/>

³⁰⁸ <https://www.ccb.jhu.edu/software/kraken2/>

³⁰⁹ <https://ccb.jhu.edu/software/bracken/index.shtml>

³¹⁰ <https://www.ccb.jhu.edu/software/kraken2/>

³¹¹ <https://ccb.jhu.edu/software/bracken/index.shtml>

³¹² <https://www.ccb.jhu.edu/software/kraken2/>

³¹³ <https://ccb.jhu.edu/software/bracken/index.shtml>

³¹⁴ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³¹⁵ <https://www.ncbi.nlm.nih.gov/taxonomy>

Note: I would normally use Kraken2³¹⁶ and only prefer Centrifuge³¹⁷ if memory and/or speed are an issue .

6.5.1 Installation

Use conda in the same fashion as before to install Centrifuge³¹⁸:

```
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 Centrifuge³¹⁹ database that can be used to assign the taxonomic labels to sequences. We opt for downloading the pre-build database from the Centrifuge³²⁰ website:

```
curl -O 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 79) 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 Centrifuge³²¹ 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 one it should contain. We call the Centrifuge³²² 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 Centrifuge³²³ that

³¹⁶ <https://www.ccb.jhu.edu/software/kraken2/>

³¹⁷ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³¹⁸ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³¹⁹ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³²⁰ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³²¹ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³²² <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³²³ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

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	queryLength	numMatches	
M02810:197:000000000-AV55U:1:1101:15316:8461					cid 1747	1747	1892	0
→103	135	1						
M02810:197:000000000-AV55U:1:1101:15563:3249					cid 161879	161879	18496	0
→151	151	1						
M02810:197:000000000-AV55U:1:1101:19743:5166					cid 564 564	10404	10404	117
→151	2							
M02810:197:000000000-AV55U:1:1101:19743:5166					cid 562 562	10404	10404	117
→151	2							

Each sequence classified by [Centrifuge³²⁴](#) results in a single line of output. Output lines contain eight tab-delimited fields; from left to right, they are according to the [Centrifuge³²⁵](#) 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 [Centrifuge³²⁶](#) 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	genomeSize	numReads	numUniqueReads	abundance
Pseudomonas aeruginosa	287	species	22457305	1	0	0.0
Pseudomonas fluorescens	294	species	14826544	1	1	0.0
Pseudomonas putida	303	species	6888188	1	0.0	
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 [Centrifuge³²⁷](#) 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.

³²⁴ <http://wwwccb.jhu.edu/software/centrifuge/index.shtml>

³²⁵ <http://wwwccb.jhu.edu/software/centrifuge/index.shtml>

³²⁶ <http://wwwccb.jhu.edu/software/centrifuge/index.shtml>

³²⁷ <http://wwwccb.jhu.edu/software/centrifuge/index.shtml>

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 Kraken2³²⁸, 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	P	1224	Proteobacteria
36.60	75	0	C	1236	Gammaproteobacteria
31.18	64	0	O	91347	Enterobacteriales
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 Kraken2³²⁹ 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)oain, (K)ingdom, (P)hylum, (C)lass, (O)rder, (F)amily, (G)enus, or (S)pecies. All other ranks are simply “-”.
5. NCBI Taxonomy ID
6. The indented scientific name

6.6 Visualisation (Krona)

We use the Krona³³⁰ tools to create a nice interactive visualisation of the taxa content of our sample [ONDOV2011] (page 86). Fig. 6.2 shows an example (albeit an artificial one) snapshot of the visualisa-

³²⁸ <https://wwwccb.jhu.edu/software/kraken2/>

³²⁹ <https://wwwccb.jhu.edu/software/kraken2/>

³³⁰ <https://github.com/marbl/Krona/wiki>

tion Krona³³¹ provides. Fig. 6.2 is a snapshot of the interactive web-page similar to the one we try to create.

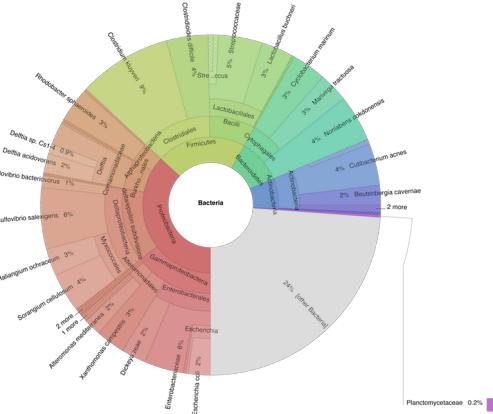


Fig. 6.2: Example of an Krona output webpage.

6.6.1 Installation

Install KRONA³³² with:

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

First some house-keeping to make the KRONA³³³ 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 KRONA³³⁴. However, if this fails we will skip this step and just download a pre-build one. Lets first try to build one.

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

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

```
# Download pre-build database
curl -O 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
```

³³¹ <https://github.com/marbl/Krona/wiki>

³³² <https://github.com/marbl/Krona/wiki>

³³³ <https://github.com/marbl/Krona/wiki>

³³⁴ <https://github.com/marbl/Krona/wiki>

Attention: Should this also fail we can download a pre-build database on the [Downloads](#) (page 79) page via a browser.

6.6.3 Visualise

Now, we use the tool `ktImportTaxonomy` from the `Krona`³³⁵ 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 `Kraken2`³³⁶ or `Centrifuge`³³⁷ 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 `Kraken2`³³⁸ results. Afterwards, we input these to the `Krona`³³⁹ script, and open the resulting web-page in a bowser. Done!

³³⁵ <https://github.com/marbl/Krona/wiki>

³³⁶ <https://www.ccb.jhu.edu/software/kraken2/>

³³⁷ <http://www.ccb.jhu.edu/software/centrifuge/index.shtml>

³³⁸ <https://www.ccb.jhu.edu/software/kraken2/>

³³⁹ <https://github.com/marbl/Krona/wiki>

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 [\[NIELSEN2011\]](#) (page 86).

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/
SolexaQA/
SolexaQA++
trimmed/
trimmed-fastqc/
trimmed-solexaqa/
```

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
# if not already installed
```

(continues on next page)

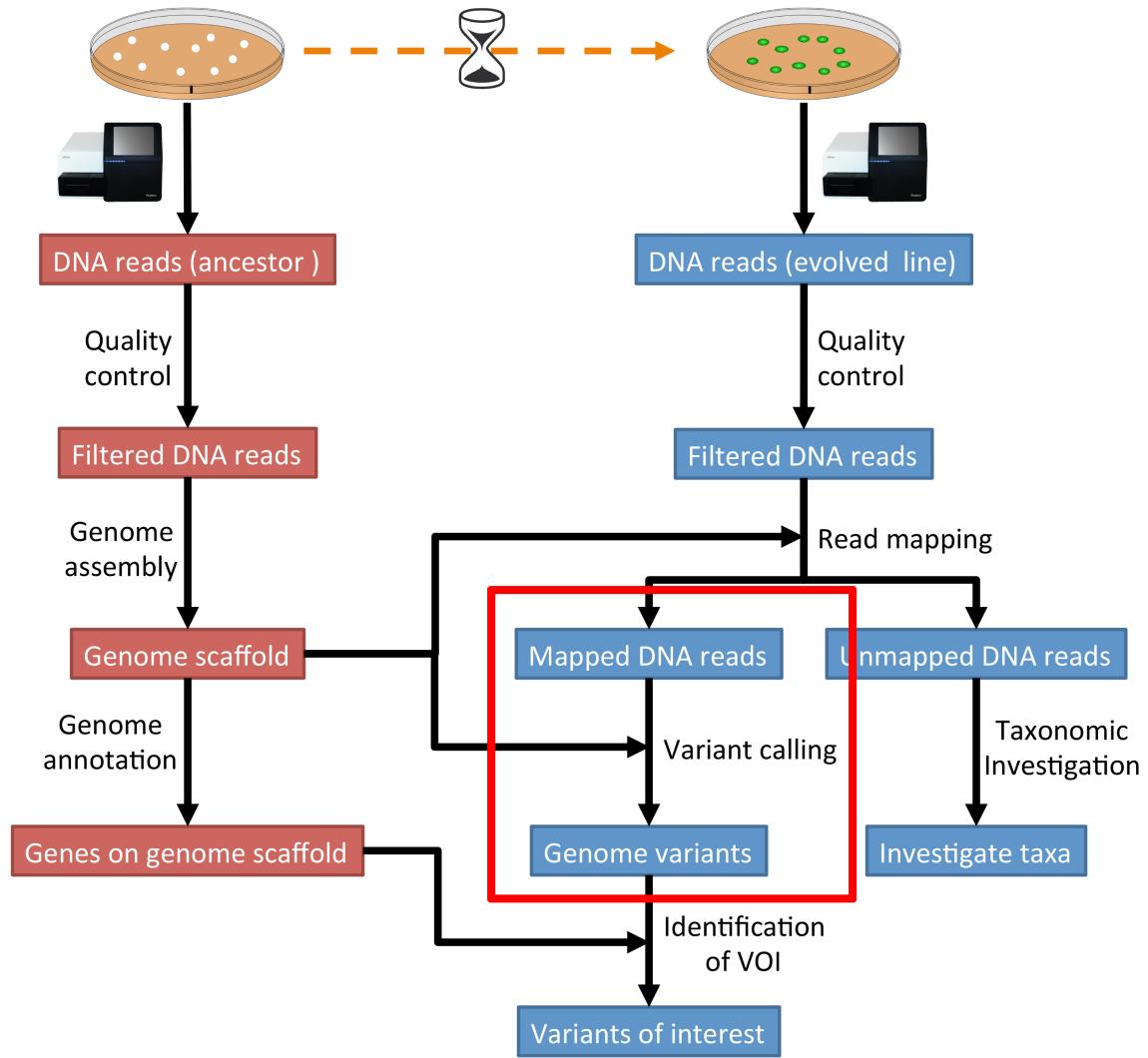


Fig. 7.1: The part of the workflow we will work on in this section marked in red.

(continued from previous page)

```
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 [SAMtools](#)³⁷² 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
```

SAMtools³⁷³ mpileup parameter:

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

BCFtools³⁷⁴ view parameter:

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

³⁷² <http://samtools.sourceforge.net/>

³⁷³ <http://samtools.sourceforge.net/>

³⁷⁴ <http://www.htslib.org/doc/bcftools.html>

7.7.2 Freebayes

As an alternative we can do some variant calling with another tool called `freebayes`³⁷⁵. 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 `freebayes`³⁷⁶ 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	ID	REF	ALT	QUAL	FILTER	INFO	FORMAT	mappings/evolved-6.sorted.
NODE_1_length_1419525_cov_15.3898	24721		.	T	164	.	.	GT:PL	DP=12;VDB=0.
	205941	SGB=-0.680642;MQ0F=0;AC=2;AN=2;DP4=0,0,12,0;MQ=40							
NODE_1_length_1419525_cov_15.3898	157033		.	AAGAGAGAGAGAGAGAGAGAGAGA					
	39.3328	AAGAGAGAGAGAGAGAGAGAGA	INDEL;IDV=6;IMF=0.146341;DP=41;VDB=0.0813946;SGB=-0.						
NODE_1_length_1419525_cov_15.3898	162469		T	C	19.609	.	.	GT:PL	DP=16;VDB=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=13,17,3,3;MQ=42							
	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:

³⁷⁵ <https://github.com/ekg/freebayes>

³⁷⁶ <https://github.com/ekg/freebayes>

Table 7.1: The vcf-file format fields.

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 sequence(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 our 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:

```
Location : variants/evolved-6.mpileup.vcf.gz
Failed Filters : 0
Passed Filters : 516
SNPs : 399
MNP's : 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)
MNP Het/Hom ratio : - (0/0)
Insertion Het/Hom ratio : 0.30 (24/80)
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 BCFtools³⁷⁷ 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.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:

³⁷⁷ <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.

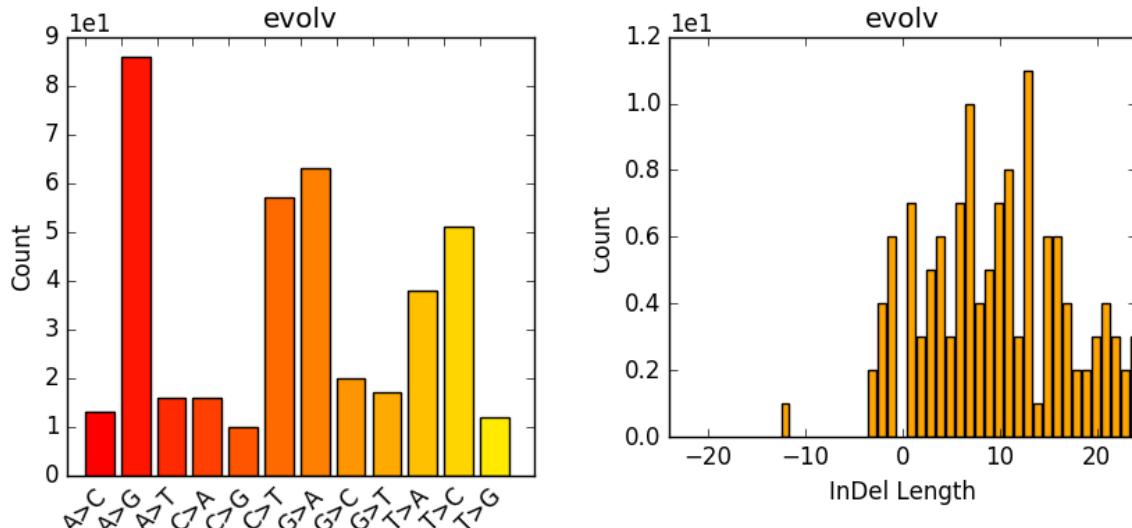


Fig. 7.2: Example of plot-vcfstats output.

7.8.3 Variant filtration

Variant filtration is a big topic in itself [OLSEN2015] (page 86). There is no consensus 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 vcffilter
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 `vcflib`³⁷⁸:

```
# 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 .

Quick stats for the filtered variants:

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

`freebayes`³⁷⁹ adds some extra information to the vcf-fields it creates. This allows for some more detailed filtering. This strategy will NOT work on the `SAMtools`³⁸⁰ mpileup called variants. Here we filter, based

³⁷⁸ <https://github.com/vcflib/vcflib#vcflib>

³⁷⁹ <https://github.com/ekg/freebayes>

³⁸⁰ <http://samtools.sourceforge.net/>

on some recommendation from the developer of freebayes³⁸¹:

```
zcat variants/evolved-6.freebayes.vcf.gz | vcffilter -f "QUAL > 1 & QUAL / AO > 10 & SAF > 0 & SAR_<br>_> 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. [here³⁸²](#).

³⁸¹ <https://github.com/ekg/freebayes>

³⁸² <https://github.com/ekg/freebayes#observation-filters-and-qualities>

GENOME ANNOTATION

8.1 Preface

In this section you will predict genes and assess your assembly using Augustus⁴¹³ and BUSCO⁴¹⁴.

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/
SolexaQA/
SolexaQA++
trimmed/
trimmed-fastqc/
trimmed-solexaqa/
variants/
```

⁴¹³ <http://augustus.gobics.de>

⁴¹⁴ <http://busco.ezlab.org>

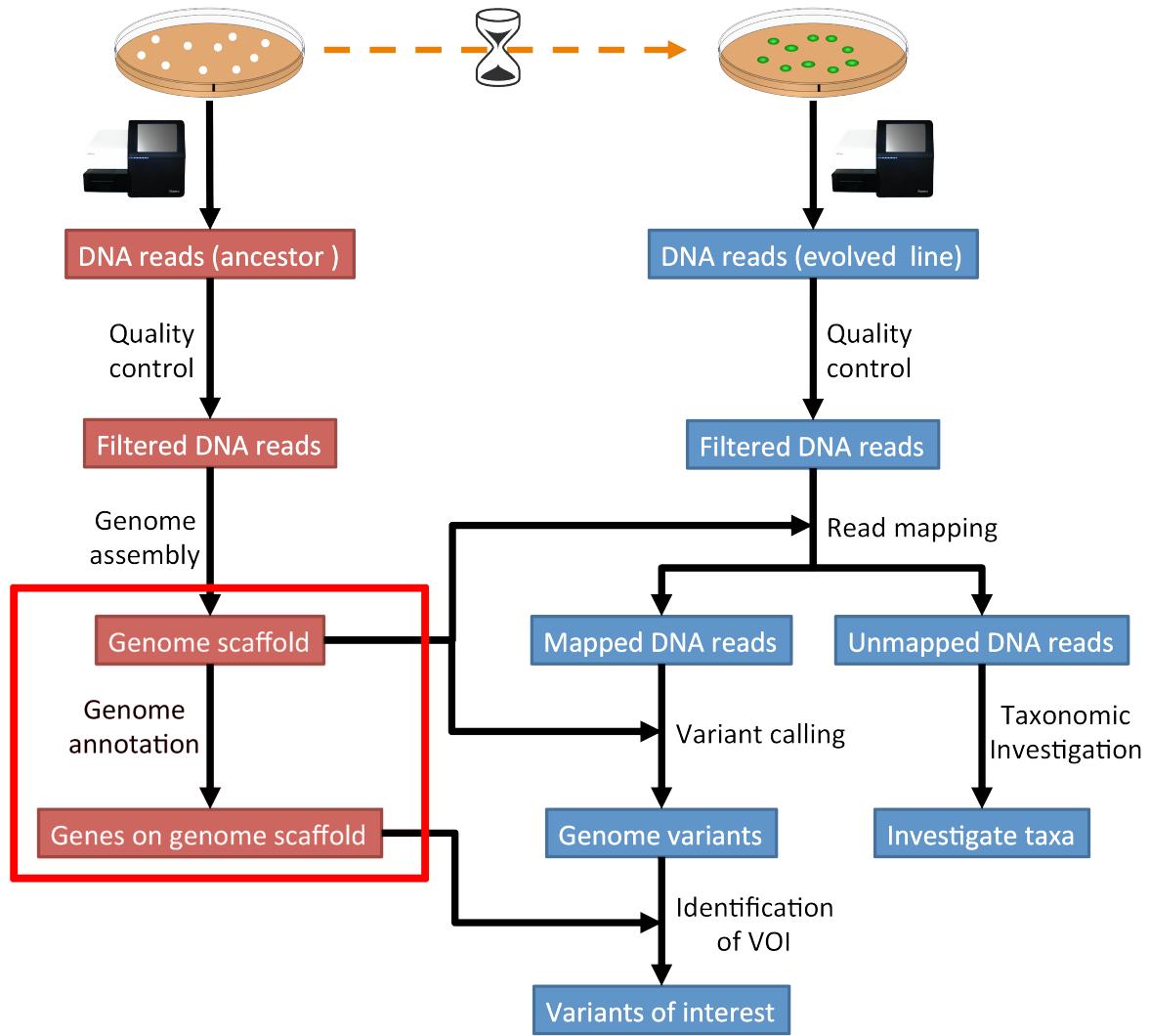


Fig. 8.1: The part of the workflow we will work on in this section marked in red.

8.5 Installing the software

```
# activate the env
conda activate ngs

conda install busco
```

This will install both the [Augustus⁴¹⁵](#) [[STANKE2005](#)] (page 86) and the [BUSCO⁴¹⁶](#) [[SIMAO2015](#)] (page 86) 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 [BUSCO⁴¹⁷](#) 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 79).

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/"
```

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

[BUSCO⁴¹⁸](#) will assess orthologue presence absence using [blastn⁴¹⁹](#), a rapid method of finding close matches in large databases (we will discuss this in lecture). It uses [blastn⁴²⁰](#) to make sure that it does 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)

⁴¹⁵ <http://augustus.gobics.de>

⁴¹⁶ <http://busco.ezlab.org>

⁴¹⁷ <http://busco.ezlab.org>

⁴¹⁸ <http://busco.ezlab.org>

⁴¹⁹ https://blast.ncbi.nlm.nih.gov/Blast.cgi?PAGE_TYPE=BlastSearch

⁴²⁰ https://blast.ncbi.nlm.nih.gov/Blast.cgi?PAGE_TYPE=BlastSearch

- 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 ./  
-saccharomyctales_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 [Augustus](#)⁴²¹ 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 its 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-  
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 79). Remember to unzip the file.

8.8 Interactive viewing

We will use the software [IGV](#)⁴²² to view the assembly, the gene predictions you have made, and the variants that you have called, all in one window.

8.9 Installing IGV⁴²³

We will not install this software using [conda](#)⁴²⁴. 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:

```
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
```

(continues on next page)

⁴²¹ <http://augustus.gobics.de>

⁴²² <http://software.broadinstitute.org/software/igv/>

⁴²³ <http://software.broadinstitute.org/software/igv/>

⁴²⁴ <http://conda.pydata.org/miniconda.html>

(continued from previous page)

```
# 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 79).

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 [BUSCO](#)⁴²⁵ 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?

⁴²⁵ <http://busco.ezlab.org>

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/
SolexaQA/
SolexaQA++
trimmed/
trimmed-fastqc/
trimmed-solexaqa/
variants/
```

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

```
mkdir phylogeny
```

Download the fasta file of the *S. cerevisiae* 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 79).

9.4 Installing the software

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

This will install a **BLAST**⁴⁵⁶ executable that you can use to remotely query the NCBI database.

```
conda install muscle
```

This will install **MUSCLE**⁴⁵⁷, alignment program that you can use to align nucleotide or protein sequences.

We will also install **RAXML-NG**⁴⁵⁸, 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**⁴⁵⁹ 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 **BLAST**⁴⁶⁰, 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 **BLAST**⁴⁶¹ hits similar to what you might see on the **BLAST**⁴⁶² web site.

Read through the output (e.g. using nano) to see what the results of your **BLAST**⁴⁶³ run was.

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

⁴⁵⁶ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁵⁷ <http://www.ebi.ac.uk/Tools/msa/muscle/>

⁴⁵⁸ <https://github.com/amkozlov/raxml-ng>

⁴⁵⁹ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁶⁰ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁶¹ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁶² <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁶³ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

```
blastn db ../assembly/spades-final/scaffolds.fasta query s_cerev_tef2.fas eval 1e-100 outfmt "6<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 BLAST⁴⁶⁴ result). This can easily be done in vi: Press the escape key, followed by: `:%s/\-//g`

Now we will BLAST⁴⁶⁵ 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 79).

9.6 Performing an alignment

We will use MUSCLE⁴⁶⁶ to perform our alignment on all the sequences in the BLAST⁴⁶⁷ 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-NG⁴⁶⁸ to build our phylogeny. This uses a maximum likelihood method to infer parameters of evolution and the topology of the tree. Again, the syntax of the command is fairly simple, except you must make sure that you are using the directory in which RAxML-NG⁴⁶⁹ 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
- `-p`: specify a random number seed for the parsimony inferences

⁴⁶⁴ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁶⁵ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁶⁶ <http://www.ebi.ac.uk/Tools/msa/muscle/>

⁴⁶⁷ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁴⁶⁸ <https://github.com/amkozlov/raxml-ng>

⁴⁶⁹ <https://github.com/amkozlov/raxml-ng>

```
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 \(iTOL\)](#)⁴⁷⁰ 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 page](#)⁴⁷¹.

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?

⁴⁷⁰ <http://itol.embl.de/upload.cgi>

⁴⁷¹ https://www.ncbi.nlm.nih.gov/Taxonomy/TaxIdentifier/tax_identifier.cgi

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/
```

10.5 General comments for identifying variants-of-interest

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

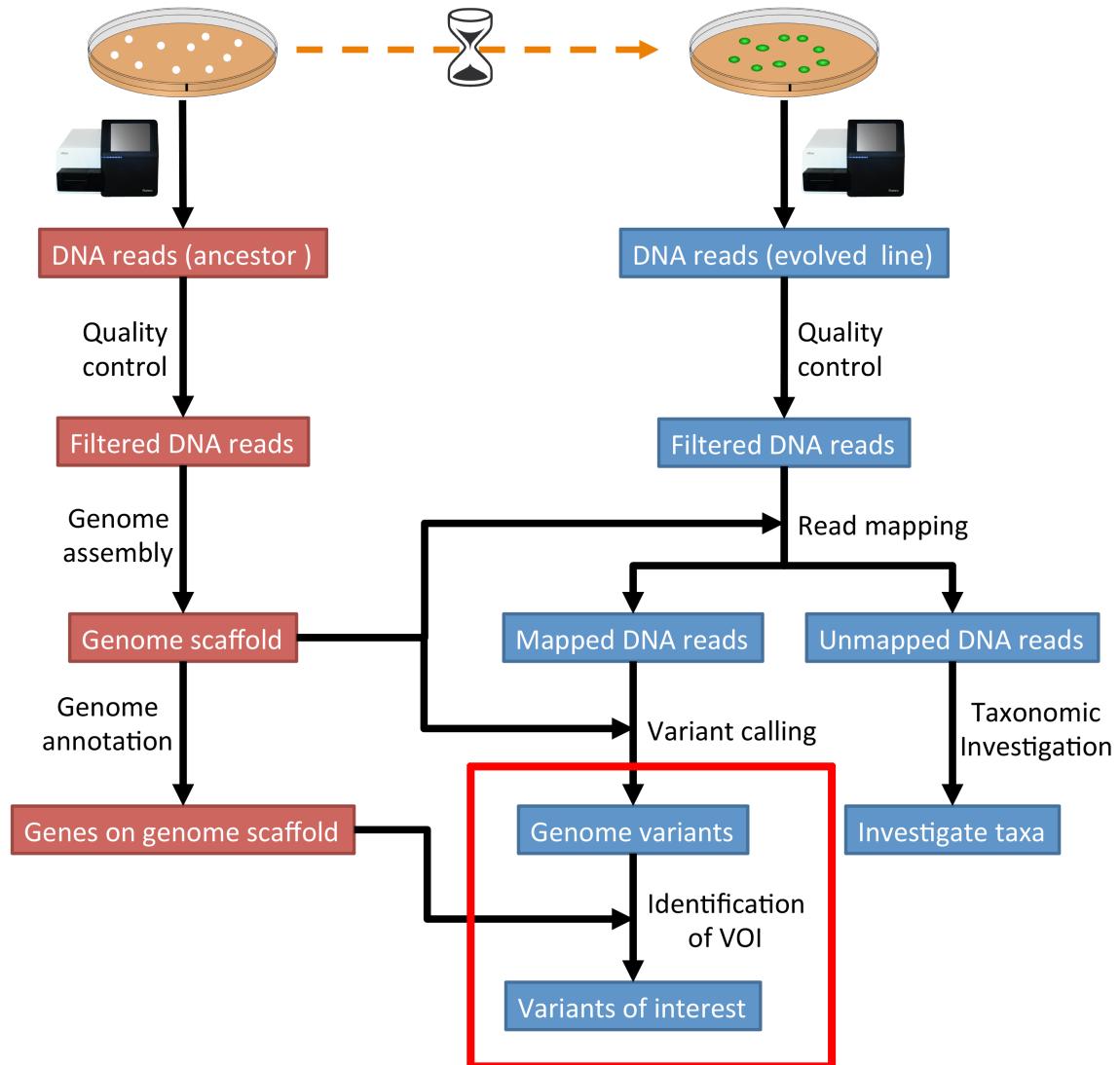


Fig. 10.1: The part of the workflow we will work on in this section marked in red.

- The quality score of the variant call.
 - Do we call the variant with a higher than 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 [SnpEff⁵⁰⁰](#) 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 have not 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 [SnpEff⁵⁰¹](#). 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 than the one shown here, depending on the version of [SnpEff⁵⁰²](#) that is installed.

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

```
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:

⁵⁰⁰ <http://snpeff.sourceforge.net/index.html>
⁵⁰¹ <http://snpeff.sourceforge.net/index.html>
⁵⁰² <http://snpeff.sourceforge.net/index.html>

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

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

```
#-----  
# Databases & Genomes  
#  
# One entry per genome version.  
#  
# For genome version 'ZZZ' the entries look like  
#   ZZZ.genome      : Real name for ZZZ (e.g. 'Human')  
#   ZZZ.reference    : [Optional] Comma separated list of URL to site/s Where information  
#   ↵for building ZZZ database was extracted.  
#   ZZZ.chrName.codonTable : [Optional] Define codon table used for chromosome 'chrName' (Default:  
#   ↵'codon.Standard')  
#  
#-----
```

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 SnpEff⁵⁰³ 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:

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

10.6.3 SNP annotation

Now we can use our new SnpEff⁵⁰⁴ database to annotate some variants, e.g.:

⁵⁰³ <http://snpeff.sourceforge.net/index.html>

⁵⁰⁴ <http://snpeff.sourceforge.net/index.html>

```
snpEff -c snpEff.config yeastanc ../variants/evolved-6.freebayes.filtered.vcf.gz > evolved-6.
→ freebayes.filtered.anno.vcf
```

SnpEff⁵⁰⁵ 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 79).

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. SnpEff⁵⁰⁶ will add fields here :

```
# evolved-6.freebayes.filtered.vcf (the original), column 8
AB=0.5;ABP=3.0103;AC=1;AF=0.5;AN=2;AO=56;CIGAR=1X;DP=112;DPB=112;DPRA=0;EPP=3.16541;EPPR=3.16541;
→ GTI=0;LEN=1;MEANALT=1;MQM=42;MQMR=42;NS=1;NUMALT=1;ODDS=331.872;PAIRED=1;PAIREDR=1;PAO=0;PQA=0;
→ PQR=0;PRO=0;QA=2128;QR=2154;RO=56;RPL=35;RPP=10.6105;RPPR=3.63072;RPR=21;RUN=1;SAF=30;SAP=3.63072;
→ SAR=26;SRF=31;SRP=4.40625;SRR=25;TYPE=snp

# evolved-6.freebayes.filtered.anno.vcf, column 8
AB=0.5;ABP=3.0103;AC=1;AF=0.5;AN=2;AO=56;CIGAR=1X;DP=112;DPB=112;DPRA=0;EPP=3.16541;EPPR=3.16541;
→ GTI=0;LEN=1;MEANALT=1;MQM=42;MQMR=42;NS=1;NUMALT=1;ODDS=331.872;PAIRED=1;PAIREDR=1;PAO=0;PQA=0;
→ PQR=0;PRO=0;QA=2128;QR=2154;RO=56;RPL=35;RPP=10.6105;RPPR=3.63072;RPR=21;RUN=1;SAF=30;SAP=3.63072;
→ SAR=26;SRF=31;SRP=4.40625;SRR=25;TYPE=snp;ANN=T|missense_variant|MODERATE|CDS_NODE_40_length_1292_
→ cov_29.5267_1_1292|GENE_CDS_NODE_40_length_1292_cov_29.5267_1_1292|transcript|TRANSCRIPT_CDS_NODE_
→ 40_length_1292_cov_29.5267_1_1292|protein_coding|1/1|c.664T>A|p.Ser222Thr|664/1292|664/1292|222/
→ 429||WARNING_TRANSCRIPT_INCOMPLETE,T|intragenic_variant|MODIFIER|GENE_NODE_40_length_1292_cov_29.
→ 5267_1_1292|GENE_NODE_40_length_1292_cov_29.5267_1_1292|gene_variant|GENE_NODE_40_length_1292_cov_
→ 29.5267_1_1292||n.629A>T|||||||
```

When expecting the second entry, we find that SnpEff⁵⁰⁷ 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 BLAST⁵⁰⁸ 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 dubliniensis*⁵⁰⁹ another fungi.

⁵⁰⁵ <http://snpEff.sourceforge.net/index.html>

⁵⁰⁶ <http://snpEff.sourceforge.net/index.html>

⁵⁰⁷ <http://snpEff.sourceforge.net/index.html>

⁵⁰⁸ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

⁵⁰⁹ https://en.wikipedia.org/wiki/Candida_dubliniensis

⁵¹⁰ <https://blast.ncbi.nlm.nih.gov/Blast.cgi>

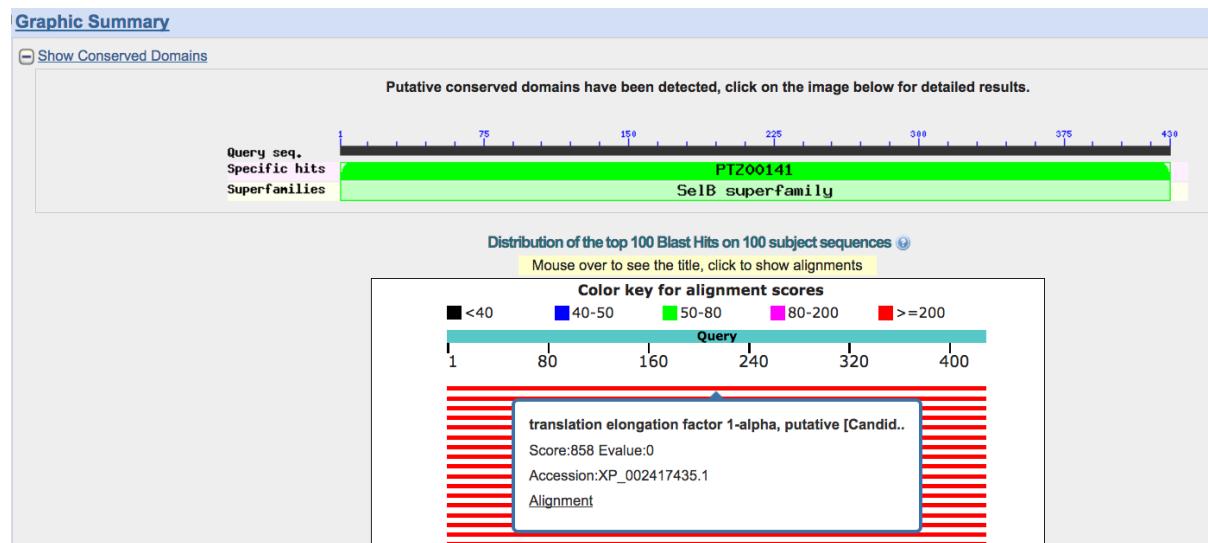


Fig. 10.2: Results of a BLAST⁵¹⁰ search of the CDS.

QUICK COMMAND REFERENCE

11.1 Shell commands

```
# Where in the directory tree am I?  
pwd  
  
# List the documents and sub-directories in the current directory  
ls  
  
# a bit nicer listing with more information  
ls -laF  
  
# Change into your home directory  
cd ~  
  
# Change back into the last directory  
cd -  
  
# Change one directory up in the tree  
cd ..  
  
# 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.txt  
  
# 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  
  
# create new env  
conda create -n [name] package [package] ...
```

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```
# activate env  
conda activate [name]  
  
# deactivate env  
conda deactivate
```

CODING SOLUTIONS

12.1 QC

12.1.1 Code: FastQC

Create directory:

```
mkdir trimmed-fastqc
```

Run FastQC:

```
fastqc -o trimmed-fastqc trimmed/ancestor-R1.fastq.trimmed.gz trimmed/ancestor-R2.fastq.trimmed.gz  
-trimmed/evolved-6-R1.fastq.trimmed.gz trimmed/evolved-6-R2.fastq.trimmed.gz
```

Open html webpages:

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

12.1.2 Code: SolexaQA++ trimming

Create directory for result-files:

```
mkdir trimmed
```

Run SolexaQA++:

```
./SolexaQA++ dynamictrim -p 0.05 -d trimmed/ data/ancestor-R1.fastq.gz  
./SolexaQA++ dynamictrim -p 0.05 -d trimmed/ data/ancestor-R2.fastq.gz  
./SolexaQA++ dynamictrim -p 0.05 -d trimmed/ data/evolved-6-R1.fastq.gz  
./SolexaQA++ dynamictrim -p 0.05 -d trimmed/ data/evolved-6-R2.fastq.gz
```

12.1.3 Code: SolexaQA++ qc

Create directory for result-files:

```
mkdir trimmed-solexaqa/
```

Run SolexaQA++:

```
./SolexaQA++ analysis -d trimmed-solexaqa trimmed/ancestor-R1.fastq.trimmed.gz  
./SolexaQA++ analysis -d trimmed-solexaqa trimmed/ancestor-R2.fastq.trimmed.gz  
./SolexaQA++ analysis -d trimmed-solexaqa trimmed/evolved-6-R1.fastq.trimmed.gz
```

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```
./SolexaQA++ analysis -d trimmed-solexaqa trimmed/evolved-6-R2.fastq.trimmed.gz
```

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.fastq.trimmed.gz -2 trimmed/ancestor-R2.fastq.trimmed.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 data/ancestor-R2.fastq.gz
```

12.3 Mapping

12.3.1 Code: Bowtie2 indexing

Build the index:

```
bowtie2-build assembly/spades-final/scaffolds.fasta assembly/spades-final/scaffolds
```

12.3.2 Code: Bowtie2 mapping

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

```
bowtie2 -X 1000 -x assembly/spades-final/scaffolds -1 trimmed/evolved-6-R1.fastq.trimmed.gz -2 trimmed/evolved-6-R2.fastq.trimmed.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:

```
bwa mem assembly/spades-final/scaffolds.fasta trimmed/evolved-6-R1.fastq.trimmed.gz trimmed/evolved-6-R2.fastq.trimmed.gz > mappings/evolved-6.sam
```

DOWNLOADS

13.1 Tools

- Miniconda installer [[EXTERNAL⁵⁹⁵](#) | [INTERNAL⁵⁹⁶](#) | [DROPBOX⁵⁹⁷](#)]
- Minikraken database [[EXTERNAL⁵⁹⁸](#)]
- Bracken file [[EXTERNAL⁵⁹⁹](#)]
- Centrifuge database [[EXTERNAL⁶⁰⁰](#) | [INTERNAL⁶⁰¹](#) | [DROPBOX⁶⁰²](#)]
- Krona taxonomy database [[INTERNAL⁶⁰³](#) | [DROPBOX⁶⁰⁴](#)]
- SolexaQA++ [[EXTERNAL⁶⁰⁵](#) | [INTERNAL⁶⁰⁶](#) | [DROPBOX⁶⁰⁷](#)]
- BUSCO Saccharomycetales_odb9 database [[EXTERNAL⁶⁰⁸](#) | [INTERNAL⁶⁰⁹](#) | [DROPBOX⁶¹⁰](#)]
- IGV [[EXTERNAL⁶¹¹](#) | [INTERNAL⁶¹²](#) | [DROPBOX⁶¹³](#)]
- RAxML-NG [[EXTERNAL⁶¹⁴](#) | [INTERNAL⁶¹⁵](#) | [DROPBOX⁶¹⁶](#)]

13.2 Data

- *Quality control* (page 9): Raw data-set [[INTERNAL⁶¹⁷](#) | [DROPBOX⁶¹⁸](#)]
- *Quality control* (page 9): Trimmed data-set [[INTERNAL⁶¹⁹](#) | [DROPBOX⁶²⁰](#)]

⁵⁹⁵ https://repo.continuum.io/miniconda/Miniconda3-latest-Linux-x86_64.sh

⁵⁹⁶ http://compbio.massey.ac.nz/data/203341/Miniconda3-latest-Linux-x86_64.sh

⁵⁹⁷ https://www.dropbox.com/s/tz2wocdzjr4grdy/Miniconda3-latest-Linux-x86_64.sh?dl=0

⁵⁹⁸ https://www.ccb.jhu.edu/software/kraken2/dl/minikraken2_v2_8GB.tgz

⁵⁹⁹ https://ccb.jhu.edu/software/bracken/dl/minikraken2_v2/database100mers.kmer_distrib

⁶⁰⁰ ftp://ftp.ccb.jhu.edu/pub/infphilo/centrifuge/data/p_compressed.tar.gz

⁶⁰¹ http://compbio.massey.ac.nz/data/203341/p_compressed.tar.gz

⁶⁰² https://www.dropbox.com/s/a4u6u37ngk2nawx/p_compressed.tar.gz?dl=0

⁶⁰³ <http://compbio.massey.ac.nz/data/203341/taxonomy.tab.gz>

⁶⁰⁴ <https://www.dropbox.com/s/cwf1qc5zyq65yvn/taxonomy.tab.gz?dl=0>

⁶⁰⁵ 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

⁶⁰⁶ <http://compbio.massey.ac.nz/data/203341/SolexaQA.tar.gz>

⁶⁰⁷ <https://www.dropbox.com/s/r9a7hg0tlwe6pk4/SolexaQA.tar.gz?dl=0>

⁶⁰⁸ http://busco.ezlab.org/datasets/saccharomycetales_odb9.tar.gz

⁶⁰⁹ http://compbio.massey.ac.nz/data/203341/saccharomycetales_odb9.tar.gz

⁶¹⁰ https://www.dropbox.com/s/7ow5yi6s5a0ente/saccharomycetales_odb9.tar.gz?dl=0

⁶¹¹ http://data.broadinstitute.org/igv/projects/downloads/IGV_2.3.92.zip

⁶¹² http://compbio.massey.ac.nz/data/203341/IGV_2.3.92.zip

⁶¹³ https://www.dropbox.com/s/bpucaolxhwf78le/IGV_2.3.92.zip?dl=0

⁶¹⁴ https://github.com/amkozlov/raxml-ng/releases/download/0.3.0/raxml-ng_v0.3.0b_linux_x86_64.zip

⁶¹⁵ http://compbio.massey.ac.nz/data/203341/raxml-ng_v0.3.0b_linux_x86_64.zip

⁶¹⁶ https://www.dropbox.com/s/iliws53ri5z4y69/raxml-ng_v0.3.0b_linux_x86_64.zip?dl=0

⁶¹⁷ <http://compbio.massey.ac.nz/data/203341/data.tar.gz>

⁶¹⁸ <https://www.dropbox.com/s/70gcfqzrqugwcn5/data.tar.gz?dl=0>

⁶¹⁹ <http://compbio.massey.ac.nz/data/203341/trimmed.tar.gz>

⁶²⁰ <https://www.dropbox.com/s/o6ioadoxfppbjrv/trimmed.tar.gz?dl=0>

- *Genome assembly* (page 23): Assembled data-set [INTERNAL⁶²¹ | DROPBOX⁶²²]
- *Read mapping* (page 29): Mapping index (bowtie2) [INTERNAL⁶²³ | DROPBOX⁶²⁴]
- *Read mapping* (page 29): Mapping index (bwa) [INTERNAL⁶²⁵ | DROPBOX⁶²⁶]
- *Read mapping* (page 29): Mapped data [INTERNAL⁶²⁷ | DROPBOX⁶²⁸]
- *Genome annotation* (page 59): GFF annotation file [INTERNAL⁶²⁹ | DROPBOX⁶³⁰]
- *Orthology and Phylogeny* (page 65): *S. cerevisiae* TEF2 gene file [INTERNAL⁶³¹ | DROPBOX⁶³²]
- *Orthology and Phylogeny* (page 65): BLAST file [INTERNAL⁶³³ | DROPBOX⁶³⁴]
- *Variants-of-interest* (page 69): SnpEff annotated vcf-file [INTERNAL⁶³⁵ | DROPBOX⁶³⁶]

⁶²¹ <http://compbio.massey.ac.nz/data/203341/assembly.tar.gz>

⁶²² <https://www.dropbox.com/s/vlyn2fxgkml5m8/assembly.tar.gz?dl=0>

⁶²³ <http://compbio.massey.ac.nz/data/203341/bowtie2-index.tar.gz>

⁶²⁴ <https://www.dropbox.com/s/dcbridsxl5bjhmif8/bowtie2-index.tar.gz?dl=0>

⁶²⁵ <http://compbio.massey.ac.nz/data/203341/bwa-index.tar.gz>

⁶²⁶ <https://www.dropbox.com/s/yidw27u56iast9z/bwa-index.tar.gz?dl=0>

⁶²⁷ <http://compbio.massey.ac.nz/data/203341/evolved-6.sorted.dedup.bam>

⁶²⁸ <https://www.dropbox.com/s/k1qn63rwnojhmrz/evolved-6.sorted.dedup.bam?dl=0>

⁶²⁹ http://compbio.massey.ac.nz/data/203341/your_new_fungus.gff.gz

⁶³⁰ https://www.dropbox.com/s/6bo9g8h3q6h1x8x/your_new_fungus.gff.gz?dl=0

⁶³¹ http://compbio.massey.ac.nz/data/203341/s_cerev_tef2.fas

⁶³² https://www.dropbox.com/s/ooxl2q0vp0bzmq3/s_cerev_tef2.fas?dl=0

⁶³³ http://compbio.massey.ac.nz/data/203341/blast_u.fas

⁶³⁴ https://www.dropbox.com/s/ul4dzx44lfoewzx/blast_u.fas?dl=0

⁶³⁵ <http://compbio.massey.ac.nz/data/203341/evolved-6.freebayes.filtered.anno.vcf>

⁶³⁶ <https://www.dropbox.com/s/67m45v5fghdh0d3/evolved-6.freebayes.filtered.anno.vcf?dl=0>

LIST OF FIGURES

1.1	The tutorial will follow this workflow.	4
3.1	The part of the workflow we will work on in this section marked in red.	10
3.2	Illustration of single-end (SE) versus paired-end (PE) sequencing.	11
3.3	SolexaQA++ example quality plot along reads of a bad MiSeq run	16
3.4	SolexaQA++ example histogram plot of a bad MiSeq run.	17
3.5	SolexaQA++ example cumulative plot of a bad MiSeq run.	17
3.6	SolexaQA++ example quality heatmap of a bad MiSeq run.	18
3.7	Quality score across bases.	20
3.8	Quality per tile.	21
3.9	GC distribution over all sequences.	22
4.1	The part of the workflow we will work on in this section marked in red.	24
5.1	The part of the workflow we will work on in this section marked in red.	30
5.2	A example coverage plot for a contig with highlighted in red regions with a coverage below 20 reads.	36
6.1	The part of the workflow we will work on in this section marked in red.	40
6.2	Example of an Krona output webpage.	48
7.1	The part of the workflow we will work on in this section marked in red.	52
7.2	Example of plot-vcfstats output.	56
8.1	The part of the workflow we will work on in this section marked in red.	60
10.1	The part of the workflow we will work on in this section marked in red.	70
10.2	Results of a BLAST search of the CDS.	74

LIST OF TABLES

5.1 The sam-file format fields.	34
7.1 The vcf-file format fields.	55

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¹⁶⁹ <http://bioinformatics.oxfordjournals.org/content/29/8/1072>

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