TRAINER'S MANUAL

Implementing Scalable Bioinformatic Workflows in Snakemake and Nextflow

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Across Australia Aug/Sep 2019

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Workshop Information

The Trainers



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Providing Feedback

While we endeavour to deliver a workshop with quality content and documentation in a venue conducive to an exciting, well run hands-on workshop with a bunch of knowledgeable and likable trainers, we know there are things we could do better.

Whilst we want to know what didn't quite hit the mark for you, what would be most helpful and least depressing, would be for you to provide ways to improve the workshop. i.e. constructive feedback. After all, if we knew something wasn't going to work, we wouldn't have done it or put it into the workshop in the first place!

Clearly, we also want to know what we did well! This gives us that "feel good" factor which will see us through those long days and nights in the lead up to such hands-on workshops!

With that in mind, we'll provide a some high tech mechanism through which you can provide anonymous feedback during the workshop:

1. Some empty ruled pages at the back of this handout. Use them for your own personal notes or for writing specific comments/feedback about the workshop as it progresses.

Document Structure

We have provided you with an electronic copy of the workshop's hands-on tutorial documents. We have done this for two reasons: 1) you will have something to take away with you at the end of the workshop, and 2) you can save time (mis)typing commands on the command line by using copy-and-paste.

We advise you to use Acrobat Reader to view the PDF. This is because it properly supports some features we have implemented to ensure that copy-and-paste of commands works as expected. This includes the appropriate copy-and-paste of special characters like tilde and hyphens as well as skipping line numbers for easy copy-and-past of whole code blocks.



While you could fly through the hands-on sessions doing copy-and-paste, you will learn more if you use the time saved from not having to type all those commands, to understand what each command is doing!

The commands to enter at a terminal look something like this:

```
tophat --solexa-quals -g 2 --library-type fr-unstranded -j \
annotation/Danio_rerio.Zv9.66.spliceSites -o tophat/ZV9_2cells \
genome/ZV9 data/2cells_1.fastq data/2cells_2.fastq
```

The following styled code is not to be entered at a terminal, it is simply to show you the syntax of the command. You must use your own judgement to substitute in the correct arguments, options, filenames etc

```
tophat [options] * <index_base> <reads_1> <reads_2>
```

The following is an example of how R commands are styled:

```
R --no-save
library(plotrix)
data <- read.table("run_25/stats.txt", header=TRUE)
weighted.hist(data$short1_cov+data$short2_cov, data$lgth, breaks=0:70)
q()
```

The following icons are used in the margin, throughout the documentation to help you navigate around the document more easily:

- [Important
- For reference
- Follow these steps
- Questions to answer
- Warning STOP and read
- Bonus exercise for fast learners
- Advanced exercise for super-fast learners



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Contributor(s):

Key Learning Outcomes

After completing this module the trainee should be able to:

- Install Snakemake in a conda environment
- Execute a Snakemake workflow
- Use the provided "profile" to execute jobs on a compute cluster
- Write simple Snakemake rules capable of generating some output(s) by executing some code which perates on some input(s)

Resources Required

For the purpose of this training you need access to:

- A compute cluster with the module command available to you for loading software
- Singularity (https://sylabs.io/singularity/) available as a module on the above cluster
- Conda(https://www.anaconda.com/distribution/) available as a module on the above cluster

Tools Used

Snakemake

```
https://snakemake.readthedocs.io
```

Graphviz

https://www.graphviz.org

Useful Links

Slurm Documentation

https://slurm.schedmd.com/documentation.html

Setting Up Your Environment

For the purpose of the workshop we will be working on the head node of an HPC cluster running slurm (https://slurm.schedmd.com/documentation.html). This is the most likely infrastructure that fellow bioinformaticians already find themselves using on a regular basis. We also assume that the cluster provides the module command for you to load software and the modules Anaconda3 and Singularity are available to use.

The execution of the Snakemake workflow will actually take place on the cluster head node with jobs being submitted to Slurm for queing and processing. From the head node, Snakemake will monitor the submitted jobs for their completion status and submit new jobs as dependent jobs complete successfully.

Connect to the Cluster Head Node



First up, lets connect to the head node of the HPC cluster using ssh.

See your local facilitator for connection details. You will have one user account per person.

Monitoring Slurm Jobs



You can monitor all jobs in the slurm squeue, or just your own job(s) using the slurm command squeue:

```
# All jobs in the queue
squeue

# Just your own jobs
squeue --user ${USER}
```

For convienience we have provided you with the sq function which produces nicer output than the default squeue and only shows your own jobs:

```
# Your own jobs
sq

# Someone elses jobs
sq --user ${SOMEONE_ELSE}
```

Install Snakemake

The recommended installation route for Snakemake is through a conda environment (https://snakemake.readthedocs.io/en/stable/getting_started/installation.html). As such, you need Anaconda3, usually avaiable to you on your cluster via the module system.



```
# We use a specific version for reproducibility reasons
    # Find the latest version: https://anaconda.org/search?q=snakemake
2
   SNAKEMAKE_VERSION="5.5.4"
3
    # Load miniconda
   module load \
6
     miniconda3-4.6.14-gcc-5.4.0-kkzv7zk
7
8
9
    #####
    # One-time commands
10
    #####
11
    # Integrate conda into bash
12
    conda init bash
13
    . \sim/.bashrc
14
15
    # Change the default location into which conda saves packages
16
    # and environments
17
    conda config --prepend pkgs_dirs /shared/${USER}/.conda/pkgs
18
    conda config --prepend envs_dirs /shared/${USER}/.conda/envs
19
20
    # Change the default channels used for finding software and
21
    # resolving dependencies
22
    conda config --add channels defaults
23
    conda config --add channels bioconda
24
    conda config --add channels conda-forge
25
   #####
26
```



Do NOT run the following command! This is provided for future reference so you know how to Install Snakemake on another system. Rather than creating the conda environment from scratch, we'll simply copy a pre-existing directory so we save time, and possible headaches.

```
# Install snakemake using conda
# This might take 5-10mins

conda create \
    --name snakemake \
    --yes \
    snakemake=${SNAKEMAKE_VERSION:-5.5.4}
```

Snakemake installation is now complete.



For the purposes of this workshop, simply copy the following .conda directory and you will have Snakemake setup and ready to go:

```
mkdir --parents /shared/${USER}
cp --recursive \
   /shared/ubuntu/.conda \
   /shared/${USER}/
```

All that is left to do is to activate the environment which will make **snakemake** available on the command line:

```
# Activate the newly created conda environment conda activate snakemake
```

Integrate Snakemake autocompletion into bash:

```
complete -o bashdefault -C snakemake-bash-completion snakemake
```

Test if Snakemake is actually working:

```
snakemake --version
```

If you experience problems with the installation, head to the Troubleshooting section for help.



While waiting for others to catch up, why not have a look into how you would go about updating Snakemake within this conda environment if there is a new version available.

```
conda update \
snakemake
```

Your First Minimal Snakefile

To get started with Snakemake, all you need to do is create a Snakefile (note the capitalisation) containing a rule which specifies how to create an output file.

Setup a working directory for this task:

```
mkdir --parents /shared/${USER}/snakemake/minimal
cd /shared/${USER}/snakemake/minimal
```

Create a file called **Snakefile** and add the following content:

```
rule hello_world:
output:
```

```
"Hello/World.txt",
shell:
"""
echo "Hello, World!" > {output}
"""
```

You can now run this workflow in one of 3 ways:

```
# Request Snakemake to generate the specific output file
# "Hello/World.txt"
snakemake Hello/World.txt

# Request Snakemake to execute the specific rule "hello_world"
snakemake hello_world
# Request Snakemake to execute the first rule in the Snakefile
snakemake
```



What happens if you run one of the above commands two or more times? Why? Snakemake found the requested output file was already there so decided not to regenerate it.

Generalising Rules with Wildcards

The original hello_world rule wasn't very flexible. We couldn't say "Hello, World!" in Spanish, Polish or French. However, we can generalise the rule using "wildcards":

```
rule hello_world:
output:
"{cheer}/{world}.txt",
shell:
"""
echo "{wildcards.cheer}, {wildcards.world}!" > {output}
"""
```

Now we can use whatever language we want:

```
# In English
snakemake Hello/World.txt

# In Polish
snakemake Czesc/Swiat.txt

# In French and Spanish at the same time
```

```
snakemake Monde/Monde.txt Ciao/Mondo.txt

Take a look at the files created:

tree ./
```

Submitting Jobs to Slurm

By default, Snakemake executes jobs on the same computer on which it is running. For Snakemake to be able to submit jobs to a cluster resource management/queuing system, such as Slurm, we can use a "profile" which convieniently contains scripts for job submission and monitoring as well as setting some additional Snakemake command line arguments so it can "talk" to a cluster backend.

To avoid having to delve into implementing our own "profile" for use with our Slurm cluster, we have created a Slurm profile ready for you to use. So lets grab it:

```
# Ensure a working directory exists and move into it
mkdir --parents /shared/${USER}/snakemake/tutorial
cd /shared/${USER}/snakemake/tutorial

# Clone the Snakemake template repository from GitHub
git clone https://github.com/UofABioinformaticsHub/snakemake_template ./

# Checkout the "simple" branch
git checkout simple
```

The Snakefile in this branch of the repository is the same "Hello, World!" example you created above, with wildcards. Lets see how we use the provided "profile" to get Snakemake to submit jobs to Slurm:

```
snakemake \
--profile profiles/slurm \
Hello/World.txt Czesc/Swiat.txt Monde/Monde.txt Ciao/Mondo.txt
```

If the STDOUT and STDERR of the command(s) in a rule are not explicitly sent to a file, then they will end up in Slurm's log file for a particular job which is normally something like slurm-<job_id>.out. This isn't that helpful for debugging purposes, so the provided profile changes this to logs/<rule_name>/<wildcards>.out e.g. logs/hello_world/cheer=Ciao,world=Mondo.out.

Cleanup after yourself!

```
snakemake \
--delete-all-output \
Hello/World.txt Czesc/Swiat.txt Monde/Monde.txt Ciao/Mondo.txt
```

A Bash "Pipeline"

A bioinformatic "pipeline" is commonly a single, monolithic bash script which performs all the tasks which need to be performed. For example, someone might have written a script for performing the following tasks:

- Run FastQC across all the raw read files
- Adapter, quality, and read length filtering using Trimmomatic
- Aggregating FastQC reports from the raw reads using MultiQC
- Index the reference FASTA file
- Perform a bwa-mem read alignment

We have this script available for you on the tutorial branch, switch to it and have a look:

```
git checkout tutorial
less analysis.sh
```

While the author of such a script should be commended for their efforts in documenting their analysis using a script, it has several significant limitations:

Not parallelised

loops over input files, executing independant commands in sequential order

Resources over-specified

the compute resources needed by the script are dictated by the command(s) with the largest requirement(s)

Not idempotent

significant programming logic is needed to wrap around commands to detect failures and only execute parts of the analysis which failed in earlier attempts



How might **you** modify the above script to:

- Add new samples
- Rerun the script if you find one of the files generated is corrupt
- Include readgroup information at the bwa-mem step (-b argumanet)

How would you avoid rerunning commands which take a long time and already completed successfully on a previous run e.g. the reference index, bwa-mem etc?

With difficulty! Enter - workflow management systems like Snakemake or Nextflow.

Reimplementing A Workflow in Snakemake

We will walk you through the steps of reimplementing the first few steps of the above script into a Snakemake workflow. Along the way, we will introduce the core concepts of Snakemake and then ask you to reimplement the bwa-mem step yourselves. For those working quickly, you will have the opportunity to reimplement the multiqc step. This will provide you with a foundation for you to be able to convert your own workflows into Snakemake rules and begin reaping the rewards of being able to run your analyses in Snakemake.

Getting the Data

We've provided you with some real data whole genome sequencing (WGS) data from wheat together with a small chunk of the wheat genome. The data set is small enough so each step in the analysis will take less than a couple of minutes to run. We have a copy of this data available locally to save on badnwidth, time and the possibility we are detected as a DDoS attack on some poor remote server!

```
# Get a copy of the data
cp --recursive \
   /shared/data/{raw_reads,references,misc} \
   ./

# Have a look at what files we'd provided
tree raw_reads references misc
```

Implementing BWA Indexing

We have provided an out-of-the-box Snakefile capable of indexing the provided reference sequence, together with comments. Lets have a bit of a play before we get around to actually running the workflow.

```
less Snakefile

# All these commands have the same effect

snakemake --dryrun bwa_index

snakemake --dryrun all

snakemake --dryrun
```

The effect --dryrun is to simply show you what "would" be run, without actually running it. It's useful to ensure you're going to get what you though, especially as your worflows get larger and more interconnected.

Another useful feature is to generate a directed acyclic graph (DAG) of the jobs which comprise the workflow and how they are linked together. Although for this workflow is not yet that impressive, but we'll have a look at how we generate the DAG:

```
snakemake \
--dag \
dot -Tpdf \
dag1.pdf
```

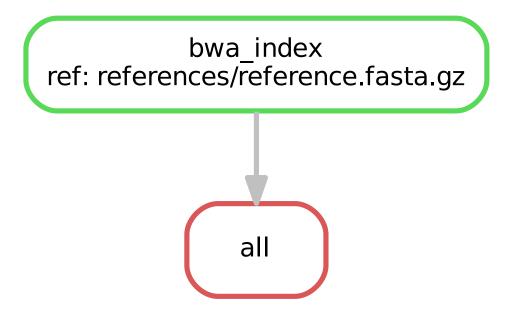


Figure 1: DAG of jobs showing bwa_index job dependant on the all pseudo-rule.

Implementing FastQC

Lets add a rule for performing FastQC on our input files. Looking at the analysis.sh file we see the following command is executed for each SAMPLE while interating over the SAMPLES list:

```
fastqc --threads 1 \
raw_reads/${SAMPLE}_R1.fastq.gz \
raw_reads/${SAMPLE}_R2.fastq.gz
```

This command can be converted into a Snakemake rule by adding the following rule to the Snakefile:

```
rule fastqc:
     input:
       r1 = "raw_reads/{SAMPLE}_R1.fastq.gz",
3
       r2 = "raw_reads/{SAMPLE}_R2.fastq.gz",
4
     output:
5
       zip = [ "raw_reads/{SAMPLE}_R1_fastqc.zip", \
6
           "raw_reads/{SAMPLE}_R2_fastqc.zip" ],
       html = [ "raw_reads/{SAMPLE}_R1_fastqc.html", \
7
           "raw_reads/{SAMPLE}_R2_fastqc.html" ],
     shell:
8
9
10
       fastqc --threads 1 {input.r1} {input.r2}
```

Now we can run Snakemake and request a "target" file which matches an output files defined by the above fastq rule:

```
snakemake --dryrun raw_reads/ACBarrie_R1_fastqc.html \
raw_reads/ACBarrie_R2_fastqc.html
```

There are a few improvements we can make to this rule:

- We don't need to process both the R1 and R2 read files with the same FastQC job. We can operate on one read file at a time. By doing this, Snakemake will be able to execute the FastQC job for each file in parallel.
- We want a convienient way of generating FastQC outputs for ALL samples without typing them all at the command line.

Improving FastQC Parallelisation

Add the following new rule to your Snakefile:

```
rule fastqc_single_input:
2
     input:
        "raw_reads/{prefix}.fastq.gz",
3
4
       zip = "raw_reads/{prefix}_fastqc.zip",
5
       html = "raw_reads/{prefix}_fastqc.html",
6
7
     shell:
        11 11 11
8
       fastqc --threads 1 {input}
9
10
```

Now run the same Snakemake dryrun command as before:

```
snakemake --dryrun raw_reads/ACBarrie_R1_fastqc.html \
raw_reads/ACBarrie_R2_fastqc.html
```



Why did Snakemake complain about an AmbiguousRuleException?

We now have two rules (fastqc and fastqc_single_input) capable of generating the two files we requested and it doesn't know which it should use.

How could we fix it? Hint: https://snakemake.readthedocs.io/en/stable/snakefiles/rules.html#handling-ambiguous-rules

Three options:

- Delete the old fastqc as the fastqc_single_input is superior
- Use ruleorder to define precedence
- Use --allow-ambiguity so Snakemake uses the first rule encountered

Go ahead and delete the fastqc rule in favour of the fastqc_single_input rule. Do the same dryrun and see how many jobs Snakemake would run in order to create those files:

```
snakemake --dryrun raw_reads/ACBarrie_R1_fastqc.html \
raw_reads/ACBarrie_R2_fastqc.html
```

Pseudo-Rules

We can use "pseudo-rules" to define a list of target filenames for creation when we use the rule name as a "target". Pseudo-rules consist of just an input directive:

```
rule all:
input:
"raw_reads/ACBarrie_R1_fastqc.html",
"raw_reads/ACBarrie_R2_fastqc.html",
```

By convention, the first pseudo-rule in the Snakefile is called all and specifies all the output filenames of the workflow. This now means we can execute a workflow in any of the following ways:

```
# Not specifying a target will result in Snakemake executing the
# first rule in the Snakefile ("all" in this case)
snakemake --dryrun

# Explicityly request the "all" rule
snakemake --dryrun all
```

When workflows get larger and the lists of filenames get bigger, specifying long lists of filenames in pseudo-rules can start to feel cumbersome. Since Snakemake syntax is an extension of Python, we can start to use some Python data structures and functions to help.

Add the following Python list of sample names (with most commented out for now) at the top of the file:

```
SAMPLES = [
1
      "ACBarrie",
2
      "Alsen",
3
    # "Baxter",
4
    # "Chara",
5
    # "Drysdale",
6
    # "Excalibur",
7
    # "Gladius",
    # "H45",
9
    # "Kukri",
10
    # "Pastor",
11
    # "RAC875",
12
    # "Volcanii",
    # "Westonia",
    # "Wyalkatchem",
15
    # "Xiaoyan",
16
    # "Yitpi",
17
18
```

Add FastQC output files for all samples in the SAMPLES list, as well as both read files, as

new targets to the existing all rule. We'll make use of the expand() function to simplify things somewhat. The resulting all pseudo-rule should look like this:

Lets take a look at what jobs would be run if we run the whole workflow. Remember, the following commands are equivilent:

```
# Explicitly run the "all" pseudo-rule
snakemake --dryrun all

# Run the first rule in the Snakefile. This should be the "all" rule by \
convention
snakemake --dryrun
```

Lets look at the DAG for the workflow:

```
snakemake \
--dag \
dot -Tpdf \
bwa_index
ref: references/reference.fasta.gz

fastqc_single_input prefix: ACBarrie_R1

fastqc_single_input prefix: ACBarrie_R2

fastqc_single_input prefix: ACBarrie_R2

fastqc_single_input prefix: Alsen_R1

fastqc_single_input prefix: Alsen_R1

fastqc_single_input prefix: Alsen_R1
```

Figure 2: DAG of jobs showing bwa_index and several fastqc_single_input job dependant on the all pseudo-rule.

Executing the Workflow on Slurm

Up until now, we've just been playing around with --dryrun, so lets move on and start executing the workflow on the Slurm cluster!

Remember, we need to use the Slurm profile we've provided you with so Snakemake knows how to communicate with Slurm. In addition, we're also going to execute the jobs within a singularity container which has the tools we need already installed inside it.

```
# Make sure Singularity is available
module load \
singularity-3.2.1-gcc-5.4.0-tn5ndnb
```

```
# Execute the workflow
snakemake \
--profile profiles/slurm \
--use-singularity
```

Depending on how quickly everyone else is in executing their workflows, you might get to see your jobs in the Slurm queue by executing this in another window:

```
1 sq
```

Implementing Trimmomatic

We've gone through implementing the FastQC command as a Snakemake rule and demonstrated the core concepts of Snakemake along the way. We'll go through implementing one more command as a Snakemake rule before you go off and try one on your own!

If you compare the trimmomatic command in analysis.sh to the rule provided below, you will see that we have simply pulled out all references to input or output files into the input or output directives. Where we had used the bash variable \${SAMPLE} in the filenames we are now using Snakemake "wildcards". It is almost the same syntax - just notice the absence of the \$ but the curly braces are retained. The biggest changes seen are in the shell directive, where we now have to refer to the input and output files via {input.r1}, {output.r1_unpaired} etc.

```
rule trimmomatic:
1
      input:
2
       r1 = "raw_reads/{SAMPLE}_R1.fastq.gz",
3
       r2 = "raw_reads/{SAMPLE}_R2.fastq.gz",
4
        adapters = "misc/trimmomatic_adapters/TruSeq3-PE.fa"
5
6
      output:
       r1 = "qc_reads/{SAMPLE}_R1.fastq.gz",
7
       r2 = "qc_reads/{SAMPLE}_R2.fastq.gz",
8
       r1_unpaired = "qc_reads/{SAMPLE}_R1.unpaired.fastq.gz",
9
       r2_unpaired = "qc_reads/{SAMPLE}_R2.unpaired.fastq.gz",
10
      shell:
11
        11 11 11
12
        trimmomatic PE \
13
          -threads 1 \
14
          {input.r1} {input.r2} \
15
          {output.r1} {output.r1_unpaired} \
16
          {output.r2} {output.r2_unpaired} \
17
          ILLUMINACLIP:{input.adapters}:2:30:10:3:true \
18
         LEADING:2 \
19
         TRAILING:2 \
20
          SLIDINGWINDOW:4:15 \
21
         MINLEN:36
22
        11 11 11
23
```

Next, we need to add the trimmomatic output files to our all pseudo-rule to make it convenient to create them. Your all pseudo-rule should look like this:

Implementing BWA-MEM

Now is your opportunity to put into practice what you have learnt from the above walk-thoughs of implementing FastQC and Trimmomatic commands. Your task is to implement the bwa mem command into a Snakemake rule.

Here are some questions to get you thinking as you try to implement this rule:



What input read files are required for the command/rule?

QC'd R1 reads for a sample: qc_reads/{SAMPLE}_R1.fastq.gz QC'd R2 reads for a sample: qc_reads/{SAMPLE}_R2.fastq.gz

Does the command/rule need the FASTA reference file or the index files as input?

The rule doesn't need the FASTA, it needs the index files:

- references/reference.fasta.gz.amb
- references/reference.fasta.gz.ann
- references/reference.fasta.gz.bwt
- references/reference.fasta.gz.pac
- references/reference.fasta.gz.sa

The BWA-MEM command uses a "prefix" to the FASTA index files, not the index filenames themselves. How will you specify this in the shell directive?

If you hard-coded it, what would the rule look like?

```
shell:
"""

bwa mem -t 1 \
references/reference.fasta.gz \
finput.r1} {input.r2} \
samtools view -b \
> {output}
"""
```



Hard-coding the path is simple, but not ideal. What if you changed the name of the reference file or wanted to use the rule with a different project? You would have to modify the paths in multiple places, once in the input directive and once in the shell directive.

Moving the hard-coded path out of the shell directive into the params directive (see: https://snakemake.readthedocs.io/en/stable/snakefiles/rules.html#non-file-parameters-for-rules).

```
params:
       prefix = "references/reference.fasta.gz",
2
3
4
       bwa mem -t 1 \
5
         {params.prefix} \
6
         {input.r1} {input.r2} \
7
        | samtools view -b \
8
       > {output}
9
10
```



Now you are making use of the params directive, this opens up the possibility of using some Python to do some string manipulations on the paths defined in the input directive. In particular, we can use a Python lambda function in the params directive (see: https://snakemake.readthedocs.io/en/stable/snakefiles/rules.html#non-file-parameters-for-rules).

How might you change a hard-coded path in the params directive to use a lambda function which manipulates the index file path(s) set in the input directive to define the prefix? Hint: take the path of one of the index files and remove the last few characters corresponding to the last file extension. You will probably need to do some reading of the Snakemake and/or Python documentation.

```
input:
1
       reference = expand("references/reference.fasta.gz.{ext}", \
2
           ext=["amb", "ann", "bwt", "pac", "sa"]),
3
4
     params:
       prefix = lambda wildcards, input: input["reference"][0][:-4],
5
      shell:
6
7
       bwa mem -t 1 \
8
         {params.prefix} \
9
         {input.r1} {input.r2} \
10
        | samtools view -b \
11
       > {output}
12
13
```

Now add the BAM files corresponding to the all the samples in the SAMPLES list, that you want the BWA-MEM rule to produce, to the all pseudo-rule.

```
input:
1
2
       reference = expand("references/reference.fasta.gz.{ext}", \
           ext=["amb","ann","bwt","pac","sa"]),
3
4
     params:
       prefix = lambda wildcards, input: input["reference"][0][:-4],
5
      shell:
6
7
       bwa mem -t 1 \
8
         {params.prefix} \
9
         {input.r1} {input.r2} \
10
        | samtools view -b \
11
       > {output}
12
        0.00
13
```

Don't worry if you didn't complete the above implementation of the BWA-MEM command, we have a git repository branch with the rules we developed. Get it, execute the workflow

and generate the DAG:

```
# Checkout the branch with our implementation of these rules
   git checkout final
3
    # Execute the workflow
4
    snakemake \
5
     --profile profiles/slurm \
6
7
     --use-singularity
8
    # Generate a DAG
9
    snakemake \
10
      --dag \
11
    | dot -Tpdf \
12
   > dag3.pdf
13
```

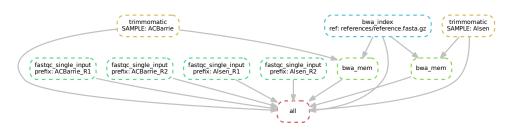


Figure 3: DAG of jobs showing the dependencies which exist in our final implementation of the analysis.sh workflow.

Adding New Samples

Our SAMPLES list contains a lot of samples which are currently commented out. Lets uncomment them and have a look at some other features of Snakemake:

```
# Manually uncomment the samples or use this sed command sed -i 's/^# "/ "/' Snakefile
```

With so many more samples, the DAG becomes next to useless:

```
# Generate a DAG
snakemake \
   --dag \
   | dot -Tpdf \
   > dag4.pdf
```

Figure 4: DAG of jobs for the whole workflow consisting of 16 samples.

Instead, the "rulegraph" might provide a better view of the workflow. Unlike the DAG, that shows the individual jobs and their dependencies, the rulegraph shows only the rules and and their dependencies so provides a simplified view of the workflow:

```
# Generate a rulegraph
snakemake \
    --rulegraph \
    | dot -Tpdf \
    > rulegraph.pdf
```

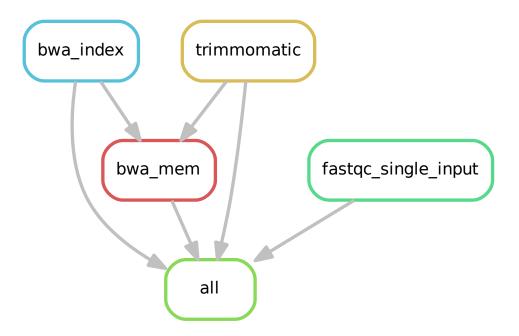


Figure 5: Rulegraph for the whole workflow.

Execute the rest of the workflow:

```
# Execute the workflow
snakemake \
   --profile profiles/slurm \
   --use-singularity
```



How many fastqc_single_input and total number of jobs are run as part of the whole workflow? Hint: try using --forceall in combination with --dryrun.

```
fastqc_single_input: 32
Total: 66
```

Using the Snakemake help, which command line argument can be used to get Snakemake to print the shell commands associated with each job during a dryrun?

```
snakemake \
--dryrun \
--printshellcmds \
--forceall
```

Using the Snakemake help, which command line argument can be used to delete all the outputs associated with a given "target"?

```
snakemake \
--delete-all-outputs
```

Snakemake Troubleshooting

Snakemake Install

If you have a broken or incomplete snakemake installation, try the following steps to fix things:

```
# deactivate the snakemake conda environment if it is already active conda deactivate

# Delete the snakemake conda environment conda env remove --name snakemake
```

Now try reinstalling snakemake.

Conda Software Environment Setup

If your job failed or timed out, you will need to re-run conda software environment setup job again. However, you may first need to release the Snakemake lock which protects you from running multiple instances of the same workflow at the same time:

```
snakemake \
--unlock
```

To ensure Snakemake starts with a clean slate, delete the "hidden" .snakemake directory:

```
1 rm -rf .snakemake
```

Getting Going After a Disconnect

If you find that your connection to the server has been dropped, you can get yourself going again using this convienient block of commands:

```
# Load the required software modules
module load \
miniconda3-4.6.14-gcc-5.4.0-kkzv7zk \
singularity-3.2.1-gcc-5.4.0-tn5ndnb

# Activate the snakemake conda environment and integrate shell \
autocompletion into bash
conda activate snakemake
complete -o bashdefault -C snakemake-bash-completion snakemake

# Move to the correct directory location
cd /shared/${USER}/snakemake-tutorial
```

Introduction to Nextflow

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Key Learning Outcomes

After completing this module the trainee should be able to:

- Install Nextflow and execute an existing Nextflow workflow locally
- Modify the workflow to allow its execution on a compute cluster
- Write simple Nextflow process definitions and connect them with channels
- Apply operators to transform items emitted by a channel
- Leverage Nextflow's implicit parallelisation to process multiple data chunks independently

Resources Required

For the purpose of this training you need access to:

- A compute cluster with the module command available to you for loading software
- https://sylabs.io/singularity/Singularity available as a module on the above cluster
- https://www.anaconda.com/distribution/conda available as a module on the above cluster

Tools Used

Nextflow

https://nextflow.io

Graphviz

https://www.graphviz.org

Useful Links

Nextflow Documentation

https://www.nextflow.io/docs/latest/index.html

Nextflow Patterns

http://nextflow-io.github.io/patterns/

Slurm Documentation

https://slurm.schedmd.com/documentation.html

Introduction

Setting Up Your Environment

For the purpose of the workshop we will be working on the head node of an HPC cluster running Slurm. This is the most likely infrastructure that fellow bioinformaticians already find themselves using on a regular basis. We also assume that the cluster provides the module command for you to load software and the modules Java and Singularity are available to use.

The execution of the Nextflow workflow will take place on the cluster head node with jobs being submitted to Slurm for queuing and processing. From the head node, Nextflow will monitor the submitted jobs for their completion status and submit new jobs as dependent jobs complete successfully.

Connect to the Cluster Head Node



First up, lets connect to the head node of the HPC cluster using ssh.

See your local facilitator for connection details. You should have one user account per person.

Install nextflow



```
# Load the Java module on your cluster
# If it's unavailable contact the cluster sysadmin
module load openjdk-1.8.0_202-b08-gcc-5.4.0-sypwasp

# Download and install nextflow executable
curl -s https://get.nextflow.io | bash

# You should now be able to run it
//nextflow help
```

The installation should have placed the executable in your working directory. It is preferable to move the executable to a directory accessible via \$PATH, to be able to run nextflow rather than having to remember to type the full /path/to/nextflow each time you want to run it.



Depending on the system this may suffice:

```
mkdir -p $HOME/bin
mv ./nextflow $HOME/bin
```

You should now be able to run nextflow without specifying the location of the binary. Let's see if it works by running a script which is nextflow's take on 'hello world'.

Hello (nextflow) world!



nextflow run rsuchecki/hello

Nextflow will pull the rsuchecki/hello GitHub repository and run its main script.



We are relying on nextflow's integration with git and git registries. The **alternative** would be to

```
git clone https://github.com/rsuchecki/hello.git nextflow run hello/main.nf
```

In which case the location of the cloned repository will be different to the one used by nextflow. You will also not have access to nextflow-git integration functionality.



Where do we find the local copy of hello? Hint: try nextflow commands related to pipeline sharing, such as list and info.

```
# List local clones of remote repositories
nextflow list
# Get detailed info about a repository
nextflow info hello #or nextflow rsuchecki/hello

project name: rsuchecki/hello

repository : https://github.com/rsuchecki/hello
local path : /home/rad/.nextflow/assets/rsuchecki/hello
main script : main.nf
revisions :

* master (default)
mybranch
slurm
testing
v1.1 [t]
v1.2 [t]
```

For now, we are mostly interested in the local path to the repository, the file name of the main script and its contents, which we will discuss next.



While waiting for others to catch up, why not have a look into how you would go about pulling and removing local clones of remote repositories using nextflow.

```
# remove local copy of rsuchecki/hello
nextflow drop hello
# pull rsuchecki/hello from remote without running the main script
nextflow pull rsuchecki/hello
```



What revisions (git branches or tags) are available for nextflow-io/hello? How would you run a specific revision?

```
# Available revisions
nextflow info hello
# Using -r/-revision, pointing to a listed tag or branch
nextflow run hello -revision v1.1
```

Nextflow basics

Processes and channels

- process a wrapper for a language-agnostic script which ensures isolation of the executed code.
- *channel* an asynchronous¹ FIFO queue which facilitates data flow to/from/between processes by linking their outputs/inputs.

 $^{^{1}\}mathrm{send}$ operation completes immediately, receiving stops the receiving process until the message has arrived

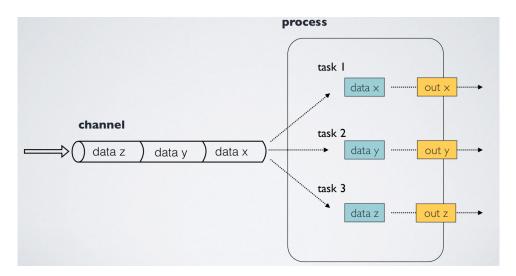


Figure 6: Nextflow building blocks: a *channel* "feeding" a processes. A *task* is an instance of a process. An isolated task is created for each emission (data chunk) from the input channel. Credit: Evan Floden

The main script

A nextflow script file name can be anything but in most cases it is best to stick to the default main.nf. The main script for the 'hello' example is as follows:

```
#!/usr/bin/env nextflow
2
    echo true
3
    cheers = Channel.from 'Bonjour', 'Ciao', 'Hello', 'Hola'
4
5
    //setting default value, to be modified at runtime
6
7
    params.world = 'world'
8
    process sayHello {
9
10
     input:
11
       val x from cheers
      script:
12
13
        echo '$x $params.world!'
14
15
```

A channel called **cheers** is created and emits each of the listed strings separately. A separate instance of the process **sayHello** is executed for each emission.



The contenet of the above script can be broken down as follows:

• The shebang line (line 1) is optional.

- Setting echo true will output stdout of (every) process to the terminal not advised for real world applications.
- Channel.from(some_list) creates a channel emitting the list elements one by one.
- Process definition (lines 6-13)
 - Input block (lines 7-8)
 - Script block (lines 9-12)
- The \$x in the script block is a nextflow variable local to the process, not a bash variable.
- Indentation is inconsequential.

In addition process directives could be inserted above the input block.

Hello HPC!

The nextflow hello example shown us how the **sayHello** process was executed separately for each input string as a separate *task*, but all the tasks were executed locally on our cluster's head node. We would now like each task to be submitted as a batch job for execution on one of the compute nodes.

```
V
```

```
nextflow run rsuchecki/hello -revision slurm
```

This is the modified version of the main.nf script. Submission to Slurm was achieved by adding executor 'slurm' directive to the process definition.

```
#!/usr/bin/env nextflow
    echo true
2
3
    cheers = Channel.from 'Bonjour', 'Ciao', 'Hello', 'Hola'
4
5
    //setting default value, to be modified at runtime
6
    params.world = 'world'
    process sayHello {
9
      executor 'slurm'
10
11
12
      input:
       val x from cheers
13
      script:
14
        11 11 11
15
        echo "$x $params.world from \$HOSTNAME on Slurm!"
16
17
```

You might also have noticed that we have modified the script block so that the messages printed to the terminal include the name of the compute node on which a given task is executed.



Note the difference between how nextflow variables (\$x,\$params.world) and bash variables (\$HOSTNAME) are included in the script block. There are alternative ways of including variables in scripts for execution by nextflow processes which may be more convenient if your script contains multiple special characters.

Hello task caching!

When the pipeline is launched with the **-resume** option, any attempt to execute already executed process with the same inputs, will cause the process execution to be skipped, producing the stored data as the output.

In this toy example we do not specify any outputs but the 'hello' messages printed to the terminal reflect this behaviour.



nextflow run rsuchecki/hello -revision slurm -resume

To avoid unintentionally re-computing long running tasks you may consider always running your pipelines with **-resume** and only omitting it on rare occasions when you want to re-compute the results even though inputs have not changed.

https://www.nextflow.io/docs/latest/process.html#cache

Hello command line options

Single-dashed options are reserved for nextflow engine (-resume, -revision, -ansi-log false etc). The double-dashed options are all yours and you are free to use them for your workflow. When you nextflow run some_script.nf --foo bar, the value of the parameter ('bar') will be accessible in main.nf as params.foo and within a script block as \$params.foo.



In the 'hello' example we use params.world which by default is set to 'word', so lets try to use an alternative string.

nextflow run rsuchecki/hello -revision slurm --world Mundo

Goodbye Hello

Nextflow facilitates but does not enforce separation of workflow logic from the configuration of compute and software environments as well as from other properties of the workflow. As such, you *could* get by developing nextflow workflows without worrying about that aspect – but you would be missing a lot in terms of flexibility, extensibility, portability and more

Nextflow looks for workflow configuration primarily in nextflow.config file, and additional config files can be included. Unsurprisingly the 'hello' example does not require much configuration, we would also like to crunch some real, albeit small, data.



This is mostly symbolic

nextflow drop rsuchecki/hello

Let's have a play with a slightly more practical workflow.

Example workflow

We are going to work with an example Nextflow workflow to demonstrate how they are run, improve your understanding of *processes* and *channels* and finally introduce *operators*, which are applied to channels to shape and direct flowing data.

This example workflow consists of the following steps:

- Running FastQC across the raw reads
- Aggregating the raw read FastQC reports using MultiQC
- Performing adapter, quality, and read length filtering using Trimmomatic
- Running FastQC across the QC'd reads
- Aggregating the QC read FastQC reports using MultiQC
- Indexing the reference FASTA file
- Performing a bwa-mem read alignment



Although not necessary for simply running the pipeline, in the training context it makes sense to start by cloning the workflow repository and moving to the directory.

```
mkdir -p /shared/${USER}/nextflow-tutorial
cd /shared/${USER}/nextflow-tutorial
git clone \
    https://github.com/csiro-crop-informatics/nextflow-embl-abr-webinar.git \
    example_workflow
cd example_workflow
git checkout noslurm
git branch
```

This time, in addition to main.nf we have a separate script which downloads the required data sets, which include a small reference FASTA file and 16 pairs of FASTQ files, each for a different bread wheat accession.

```
nextflow run setup_data.nf
```

If successful, we could now try to run the workflow...

```
nextflow run main.nf
```



This is expected to fail.

Unless all the software required by the pipeline is available on the \$PATH, which we don't expect, the pipeline should terminate with an error. The output information may help you identify the cause. Try to relate the error message to the relevant section of the main script (main.nf).



Which process has failed? What was the underlying cause? The cause was likely "command not found" and it may have been any of the processes for which the software tool is not available. Example:

```
Error executing process > 'fastqc_raw (Xiaoyan)'
2
3
   Caused by:
     Process `fastqc_raw (Xiaoyan)` terminated with an error exit \
4
         status (127)
5
    Command executed:
6
7
     fastqc --quiet --threads 1 *
8
9
    Command exit status:
10
     127
11
12
    Command output:
13
      (empty)
14
15
    Command error:
16
      .command.sh: line 2: fastqc: command not found
17
18
   Work dir: \
19
       /tmp/nextflow-embl-abr-webinar/work/15/505ea816d2411e68ea253ee126c181
```

There are two main issues with executing this workflow as is,

- 1. Third-party software tools have not been made available to the workflow.
- 2. We are trying to run the entire workflow on the cluster's head node.

There are different ways in which these issues could be addressed, for example using process directives at the top of each process definition. Depending on your cluster configuration this could be for example:

```
process foo {
  executor 'slurm'
  module 'samtools/1.9'
  //further code omitted
```

This is a perfectly valid syntax, which can be convenient, particularly during pipeline development, but for more portable workflows it is preferable to keep compute and software environment configuration separate from pipeline logic – in simple terms not in the workflow script (main.nf).

The config file(s) and profiles

Workflow configuration belongs in nextflow.config file. Transferring the above mention directives from process definitions in main.nf to nextflow.config would make things slightly better, e.g.

```
#nextflow.config
process.executor = 'slurm'
process.module = 'samtools/1.9'
```

or using the preferred syntax

```
process {
   executor = slurm
   module = 'samtools/1.9'
}
```

This is however still a bit rigid.

- You may be developing your pipeline on a local machine or a server where software modules are not available.
- If developing directly in the cluster environment, you may prefer your quick test runs to happen either on the head node or in an interactive session you are using, rather than always having jobs submitted to sit in the always-busy cluster queue.

Nextflow enables the definition of *profiles* which make it easy to run a workflow with different configuration settings, including, but not limited to executors and software environment.

For our pipeline we have defined several *profiles*, which allow us to execute the logic in main.nf while providing the required software either by creating a conda environment or by using Docker of Singularity containers where the conda environment has already been captured.

Relevant profiles

Identify the profile definitions in nextflow config. The ones most immediately relevant are:

```
profiles {
1
      //SOFTWARE
2
      conda {
3
       process {
4
          conda = "$baseDir/conf/conda.yaml"
5
6
7
      }
      singularity {
8
       process {
9
          container = \
10
              'shub://csiro-crop-informatics/nextflow-embl-abr-webinar'
11
        singularity {
12
          enabled = true
13
          autoMounts = true
14
          cacheDir = "singularity-images"
15
        }
16
      }
17
    }
18
```

As you can see, Nextflow makes it really easy to define software environment via Singularity or Conda².

Given that Singlularity is available on our cluster, let's start by using that profile, as the most robust way of setting up the software environment.

We will need Singularity for nextflow to be able to pull the container image from Singularity Hub and run the containerised software. By default the pipeline will process reads for a single accession – our head node should be able to handle this.

```
# Load the Singularity module
# If it is unavailable contact the cluster sysadmin

module load singularity-3.2.1-gcc-5.4.0-tn5ndnb

# Run the workflow
```

nextflow run main.nf -profile singularity

This is sufficient when running a workflow locally, in an interactive session or on a standalone server. The next step is to get nextflow to make use of the HPC batch submission

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 $^{^2}$ We also have a docker profile which you may find useful if you decide to run the workflow on your machine

system, to be able to run the full workflow without unleashing your sysadmins wrath.



Edit nextflow.config. Your task is to add a slurm profile which will set the appropriate executor.



There are of course many setting that can and in some cases must be set – refer to executors section of Nextflow documentation³. For running real-life pipelines in a cluster environment you will also use directives ⁴ controlling the resources (cpus, memory, time) requested for each job. Other possibly relevant directives include queue and scratch.

Cluster run

To avoid running the workflow on our head node or in an interactive session, we will use the slurm profile you have defined⁵. As before, the software environment will be handled via the singularity profile. For that, we will need Singularity on the head node for nextflow to be able to pull the container image from Singularity Hub (we could also use a locally stored image). Singularity will also be required on the compute nodes which will run the individual tasks, but this should happen seamlessly if an appropriate module is loaded on the head node, otherwise the required module would also have to be specified in the workflow configuration files.



By default a single accession will be processed. You may use the **-resume** flag to avoid re-computing already existing results.

```
# Load the Singularity module on your cluster
# If it is unavailable contact the cluster sysadmin

module load singularity-3.2.1-gcc-5.4.0-tn5ndnb

# Run the workflow

nextflow run main.nf -profile slurm, singularity -resume
```

³https://www.nextflow.io/docs/latest/executor.html

⁴https://www.nextflow.io/docs/latest/process.html#directives

⁵If you are struggling and can't get help, try: git stash && git checkout workshop

Under the hood

If you think you are ready to look under the hood and try to work out how nextflow stages process inputs, wraps process script blocks and submits them to the cluster, here is a start.



```
# Remove the work directory to limit the number of task directories \
    to look at

rm -r work

# Re-run for a single sample

nextflow run main.nf -profile slurm, singularity

# Take a peak

ls -la work/ | less

# or

tree -ah work/ | less
```

Each task is executed in a separate directory and every abbreviated hash displayed in the terminal can be related to a specific sub-directory of ./work, such as work/d2/c4517b0a81f61ceca29ec355ddeaa6/ in which you may find

```
# NF generated files
   .command.begin
2
3
   .command.err
   .command.log
4
   .command.out
5
   .command.run
6
7
   .command.sh
8
   .command.trace
   .exitcode
10
   # Output file
11
   H45.bam
12
13
   # Symlinks to input files
14
   H45_R1.paired.fastq.gz
   H45_R2.paired.fastq.gz
16
   reference.fasta.gz.amb
17
18
   reference.fasta.gz.ann
   reference.fasta.gz.bwt
   reference.fasta.gz.pac
  reference.fasta.gz.sa
```

Identify and investigate hidden file (starting with dot) containing the executed script and the one containing cluster and container handling.

Cluster run - all accessions

We have successfully submitted workflow to the cluster.

To be sure, feel free to re-run it again (and again, and again...) with -resume to avoid wasting CPU cycles.



```
nextflow run main.nf -profile slurm, singularity -resume
```

If all went well, the workflow successfully processed a single accession, let's have a closer look at the script to better understand how it handles the inputs before we proceed to run it on all the accessions.



In main.nf we create a channel which reads pairs of FASTQ files from a sub-directory of the ./data. We then apply some operators.

```
Channel.fromFilePairs("data/${region}/*_R{1,2}.fastq.gz")
   .take ( params.take == "all" ? -1 : params.take )
   .into { readPairsChannelA; readPairsChannelB }
```

- 1. Identify the two operators, refer to nextflow documentation^a as required and explain the purpose of each of the two operators.
- .take(n) limits the number of emissions from the channel to the first n items.
- .into{ $ch_1; ch_2; ...; ch_n$ } creates channels $ch_1, ch_2, ..., ch_n$ and connects source channel to the newly created channels, so that every emission is sent through each new channel.
- 2. How can you run the workflow for more than one accession? How about all of them? Recall that workflow parameters use double-dash syntax. Run the relevant commands

```
nextflow run main.nf -profile slurm, singularity -resume --take 2 nextflow run main.nf -profile slurm, singularity -resume --take all
```

ahttps://www.nextflow.io/docs/latest/operator.html

Monitoring your jobs on our cluster



You can monitor your job(s) in the slurm queue using the slurm command squeue:

```
squeue --user ${USER}
```

For convenience you are also provided with the sq function which produces nicer output and by default only shows your own jobs:

```
sq

# Someone elses jobs

q --user ${SOMEONE_ELSE}
```

If you want to see all jobs in the queue:

```
1 squeue
```



For an optional exercise you may try to re-run the workflow with conda. For that, you'll need to find and load a conda module before re-running the workflow with appropriate profile. Don't forget to use the -resume flag.

```
# Find the appropriate module name
module av -l 2>&1 | grep conda

# Load the module
module load miniconda3-4.6.14-gcc-5.4.0-kkzv7zk

# Run with conda
nextflow run main.nf -profile conda,slurm --take all -resume
```

If you remembered to use -resume, why do you think it appeared to not make a difference?

We have switched from singularity to conda so the software environment has changed.

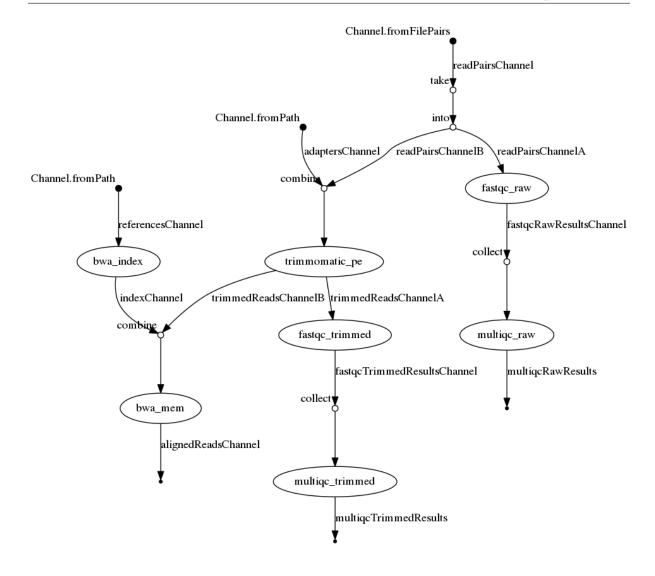


Figure 7: The example workflow



Investigate main.nf alongside Figure 7.

Which nextflow operators^a, in addition to the previously discussed, are used and for what purposes?

The .combine() operator outputs all combinations of items emitted by two channels. This results with a downstream process to be executed for each such combination. So e.g. bwa_mem will be executed for

 $(reference, accession_1), (reference, accession_2), ..., (reference, accession_n).$

The .collect() operator collects all the items emitted by a channel returns the resulting List as a single emission. This is required e.g. if a process needs to be executed once with all the samples as input.

ahttps://www.nextflow.io/docs/latest/operator.html

Workflow outputs

We now now each task is nicely isolated in a separate sub-directory under work, but how do I find my results? Was it work/a7:fc9339a827fb4b34d2408e1c3ee29c or maybe work/3c:8fdf958e96b448ecb83bd7806af382? This should be handled by applying the publishDir directive⁶ to selected processes. As with other directives, this can be included at the top of the process block or in a configuration file using process selectors⁷ to apply the directive to one or more relevant process. To keep things tidy-ish, we define the publishing of the outputs in a separate file which we includeConfig 'conf/publish.config' in nextflow.config.

In conf/publish.config we only really use the withName selectors. The alternative withLabel selectors are convenient e.g. when outputs of multiple processes are to be gather in one location, in which case we attach the same label to each of those processes.

Modify/extend the workflow

11

Edit main.nf. Your task is to add a process which will merge the bam files produced by the bwa_mem process.

Q

How do you ensure that **all** BAM files end up in the same instance of your process? Using the .collect() operator. Demonstrate your process definition to your facilitator.

```
process bam_merge {
  input:
  file('*.bam') from alignedReadsChannel.collect()

script:
  """
  samtools merge ${params.take}_accessions_megred.bam *.bam
  """
}
```

Where can we find the merged BAM file? Can you publish it to a human-readable location? Hint: only declared outputs can be published.

 $^{^6}$ https://www.nextflow.io/docs/latest/process.html#publishdir

⁷https://www.nextflow.io/docs/latest/config.html#process-selectors





Modify your merge process to allow samtools to use 2 cpus with --threads 2, don't forget to modify your process configuration to request 2 cpus per task.

Your own workflow (TODO: replace with variant calling?)

It is time to have a go at your own pipeline. Since we have some inputs and configuration files at hand, you can start a own.nf script file in the current directory and read the input files from ./data.



The simple pipeline should include the following:

- Code for reading FASTQ read files from ./data individually (i.e. not as pairs) into a channel.
- A process which will take a read file, count the reads and output the file name alongside the read count.
- A way of aggregating the individual count files into a single csv file. This could be done in another process or using an operator.

There are different ways of approaching the exercise, here is an example solution. For comparison, we demonstrate the aggregating step both as a process and using the collectFile() operator.

```
readsChannel = Channel.fromPath("data/**.fastq.gz")
2
   process countReads {
3
     input:
4
       file fastq from readsChannel
5
6
     output:
7
       file '*' into countsChannel1, countsChannel2
9
10
     echo -ne "${fastq}," > count
11
     zcat $fastq | paste - - - - | wc -l >> count
12
13
    }
14
15
   process aggregate {
16
     publishDir params.outdir
17
18
19
       file '*.count' from countsChannel1.collect()
20
21
22
      output:
       file '*.csv'
23
24
25
      cat *.count > counts_from_process.csv
26
27
28
29
    countsChannel2.collectFile(name: 'counts_from_operator.csv', \
30
       storeDir: params.outdir)
```

Troubleshooting

Disconnected from the cluster?

Missing modules - new shell session?

Make sure all the required modules are loaded.

```
V
```

```
# Java - essential for nextflow
module load openjdk-1.8.0_202-b08-gcc-5.4.0-sypwasp

# Singularity - our go to system for providing software for the example \
workflow
module load singularity-3.2.1-gcc-5.4.0-tn5ndnb

# If using conda
module load miniconda3-4.6.14-gcc-5.4.0-kkzv7zk
```

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