



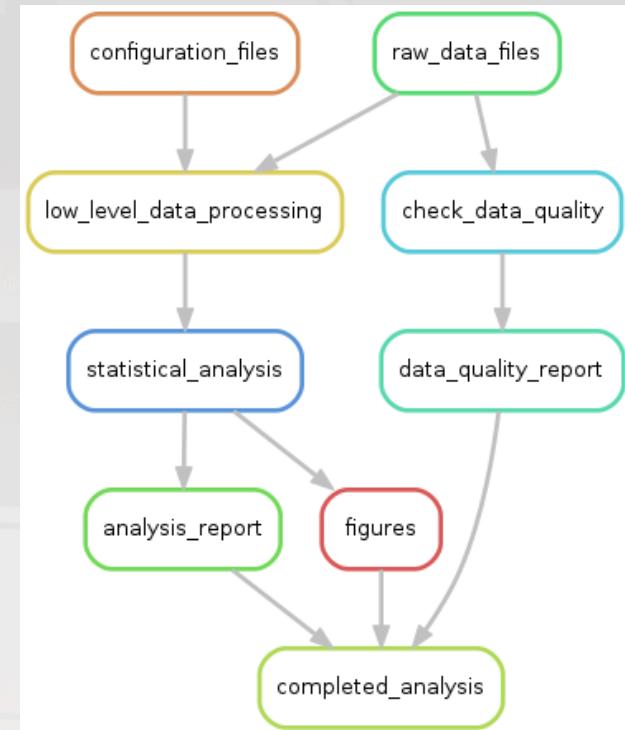
# The Snakemake Workflow Manager

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## A workflow manager is software that:

- Conducts a complex work flow or analysis
- Follows dependencies from results back to configuration and data files
- Executes statements step-by-step to carry out work flow





# Why use a workflow manager?

- Human efficiency and convenience
- Computational efficiency – only the required steps are executed
  - Great when your cluster job is preempted
- Reproducibility
- Portability between clusters, institutions
- Modularity – re-use and standardization



# Why choose snakemake?

Over 100 different workflow managers:

<https://github.com/pditommaso/awesome-pipeline>

Snakemake is:

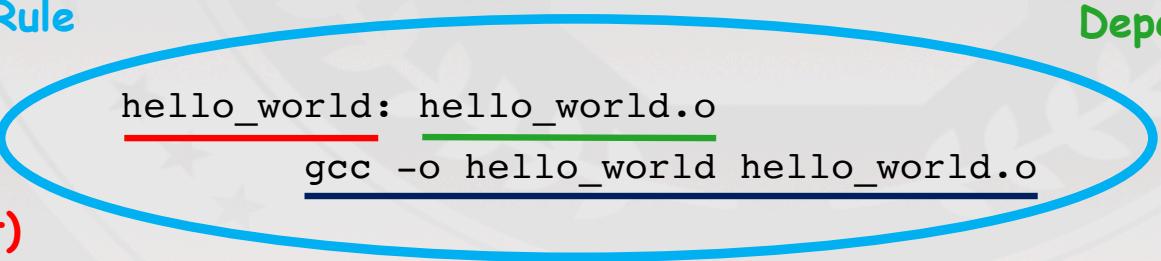
- Actively used and developed
- Can be configured for local and/or cluster execution
- Native SLURM support
- No significant system administration support required
- General purpose (not just for bioinformatics, for example)
- Significant functionality bang for your learning buck

# Installation options

- Use the CHPC module:
  - `module load snakemake/5.6.0`
- Install your own using pip:
  - `pip install --user snakemake`
  - `export PATH=$HOME/.local/bin:$PATH`
- Install your own using anaconda:
  - `module load anaconda3`
  - `conda install -c bioconda -c conda-forge snakemake`

# Snakemake is a better “make”

Classical Makefile example:



The diagram illustrates a Makefile rule with the following components:

- Target (output)**: The target file, `hello_world`, is highlighted in red.
- Rule**: The rule name, `hello_world`, is highlighted in blue.
- Dependencies (input)**: The dependencies, `hello_world.o`, are highlighted in green.
- Action**: The action command, `gcc -o hello_world hello_world.o`, is highlighted in dark blue.

```
hello_world: hello_world.o
    gcc -o hello_world hello_world.o

hello_world.o: hello_world.c hello_world.h
    gcc -c hello_world.c
```

# Snakemake workflows are built out of rules

```
rule link:  
    input: "hello_world.o"  
    output: "hello_world"  
    message: "Rule {rule} linking .o file {input}"  
    shell: "gcc -o {output} {input}"
```

Rules can have:

- names
- inputs
- outputs
- actions (shell or python)

Rules:

- are linked implicitly
- (or explicitly)
- can emit messages
- are executed in parallel if possible
- are executed locally or on a cluster

The first rule defines the default “target” for the workflow

# Snakefile syntax

- Snakemake work flows ("snakefiles") are python code
- All the python syntax rules apply:
  - Input and output file names in quotes
  - Shell commands in quotes
  - Whitespace / indentation is significant
  - Use either tabs or spaces (not both)
- Your snakefiles can include blocks of python code

## Rule inputs

- Inputs are one or more file names, in quotes, comma-separated
- Inputs are optional
- Inputs can have “symbolic” names

```
rule align:  
    input: index="hg19", data="sample1.fastq"  
    output: "sample1.sam"  
    shell: "bwa mem {input.index} {input.data} -o {output}"  
    message: "Rule {rule} aligning input file {input.data}"
```

## Rule outputs

- Same as inputs: one or more file names, in quotes, comma-separated
- Same as inputs: can have "symbolic names"
- Outputs are optional - common in top-level rule that simply checks if inputs are present.

```
rule align:  
    input: index="hg19", data="sample1.fastq"  
    output: "sample1.sam"  
    shell: "bwa mem {input.index} {input.data} -o {output}"  
    message: "Rule {rule} aligning input file {input.data}"
```



## Rule actions: the “shell:” section

- This is where you encode the actual work of the work flow
- By default: /bin/bash in strict mode (set –euo pipefail)
- Multi-line shell statements: use triple-quotes
- Can load modules, only affects the current rule.

```
rule link:  
    input: "hello_world.o"  
    output: "hello_world"  
    shell: """  
        module load gcc/6.1.0  
        gcc -o {output} {input}  
    """
```

<https://snakemake.readthedocs.io/en/stable/>

## Rule “run:” section: action as python code

- Instead of bash, the action can be written in python
- Put this in the “run:” section of the rule
- Note there are no quotes around the python code

```
rule usercount:  
    input: "userfile.txt"  
    output: "users.count"  
  
    run:  
        users=set()  
        with open(input[0]) as infile:  
            for line in infile:  
                uid=line.split()[0]  
                users.add(uid)  
        with open(output[0],'w') as outfile:  
            print(f"There are {len(users)} users.",file=outfile)
```

<https://snakemake.readthedocs.io/en/stable/>

## Rule messages

- Rules can emit messages with the “message:” section
- Messages are optional
- Really useful for monitoring your workflow
- Can access the inputs, outputs with {input}, {output}
- Can access the rule name as {rule}

```
rule align:  
    input: index="hg19", data="sample1.fastq"  
    output: "sample1.sam"  
    shell: "bwa mem {input.index} {input.data} -o {output}"  
    message: "Rule {rule} processing input file {input.data}"
```

# Snakemake command line arguments

First, need to load the module:

```
$ module load snakemake/5.6.0
```

Run snakemake on default "Snakefile", default (ie first) rule:

```
$ snakemake
```

Run snakemake on non-default snakefile:

```
$ snakemake -s my_snakefile
```

Run snakemake on non-default rule:

```
$ snakemake rule_name
```

Read the snakemake help:

```
$ snakemake --help
```



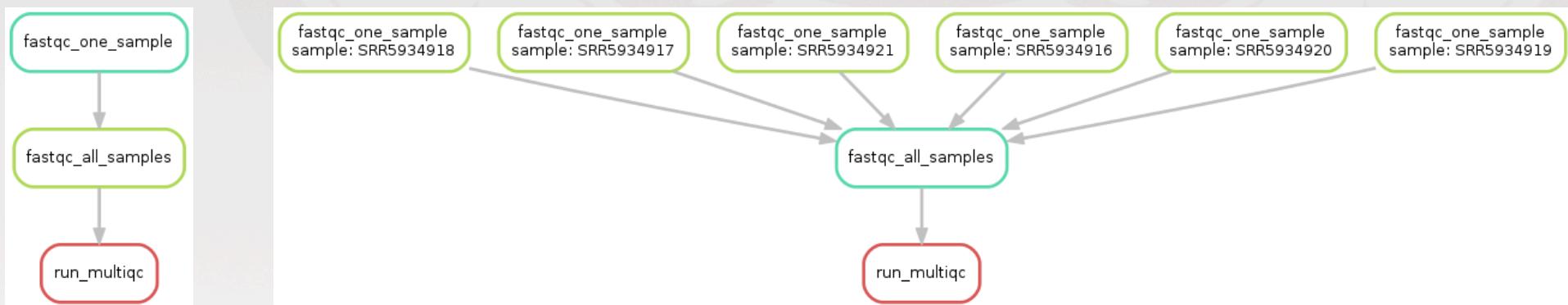
# Exercise1 - Simple workflow

See the exercise 1 instructions here:

<https://gitlab.chpc.utah.edu/bmilash/workflows-with-snakefile/-/tree/master/Exercises/Exercise1>

# Graphical output

- Rule graph
  - Shows in general how rules depend on one another, but not the actual inputs/outputs
  - `snakemake -s snakefile --rulegraph | dot -Tpng > rulegraph.png`
- Directed Acyclic Graph (DAG)
  - all targets represented
  - Completed rules have dashed outline
  - `snakemake -s snakefile --dag | dot -Tpng > dag.png`



# Directories as input or output

- In snakemake version 5.0 or later:
  - Directories as input or output must be specified with directory()
    - input: directory("data\_directory"), "data\_file"
- In older version of snakemake:
  - Directories as input or output are just named like regular files
    - input: "data\_directory", "data\_file"

# Wildcards: filename patterns

- These make rules reusable, not tied to specific files
- Rules with wildcards are ideal for parallel execution

How to do it:

- Create one rule that handles a single input -> output action using {variable} as a placeholder for the variable part of the input and output file name(s). This acts as a **template**.
- Create another rule whose **input** lists all the template rule's output files.
  - You can use the `expand()` function for this.
  - Python lists and list comprehension are useful here.



# Snakemake wildcard example

```
# Calculate the MD5 checksum for each sample's .txt file.  
# Here are the sample names embedded in the file names:  
samples=[ 'A', 'B', 'C', 'D', 'E', 'F' ]  
  
rule all_checksums:  
    input: expand("{sample}.md5", sample=samples)  
    # This produces the list ["A.md5", "B.md5", ... "F.md5"]  
  
rule one_checksum:  
    input: "{sample}.txt"  
    output: "{sample}.md5"  
    shell: "md5sum {input} > {output}"
```



## Exercise 2: Workflow with wildcards

See the exercise 2 instructions here:

<https://gitlab.chpc.utah.edu/bmilash/workflows-with-snakefile/-/tree/master/Exercises/Exercise2>



## Snakemake on a cluster

- Any snakemake workflow can run on a cluster:  
`snakemake --cluster-config cluster.yaml --jobs 20 ...`
- Cluster configuration file can be in JSON or YAML format
- The catch is that we must tell snakemake how to start a job:
  - `--cluster "sbatch -A {cluster.account} -p {cluster.partition}"`

# Cluster configuration

- Basic cluster configuration file:

```
# cluster.yaml - cluster configuration for my snakemake job.  
__default__:  
    partition: slurm_partition  
    account: slurm_account  
    time: 1:00:00  
    nodes: 1
```

- The `__default__` config applies to all rules
- Can override default with rule-specific configurations

```
image_processing:  
    partition: kingspeak_gpu  
    account: kingspeak_gpu
```

<https://snakemake.readthedocs.io/en/stable/>



## Local rules

- When running on a cluster, may want to specify some rules NOT run on the cluster
- localrules: rule1, rule2, rule3
- Snakemake knows to run rules without an action (e.g. “shell:”) locally.



# Watching your workflow run on the cluster

- Run the squeue command to see your SLURM jobs:

- `watch -n 3 squeue -M all -u $USER # Check jobs on all clusters every 3s.`

- You can get fancy with the output:

- `watch -n 3 squeue -M all 3 -u $USER -o "%.6i %.10P %.7a %.20j %.2t %.6M %R"`

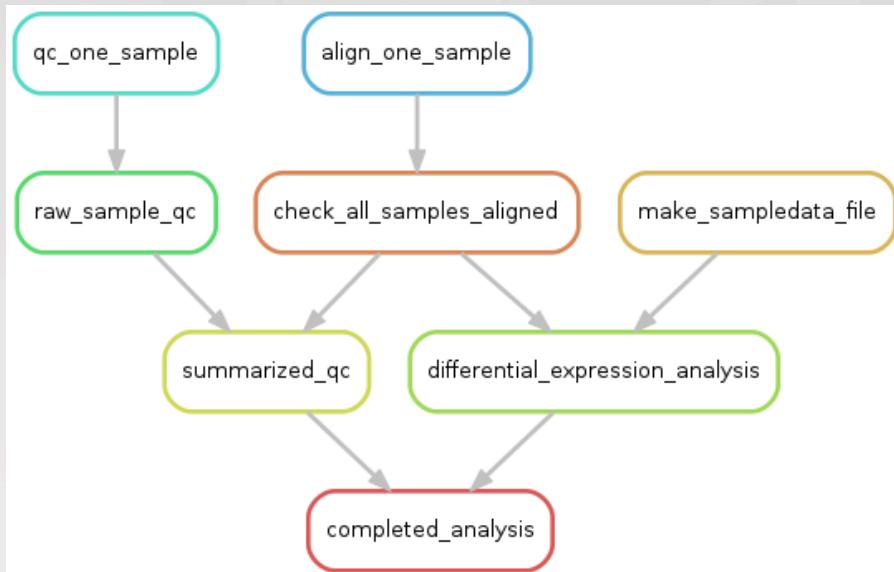


# Snakemake exercise 3

- See the exercise 3 instructions here:

<https://gitlab.chpc.utah.edu/bmilash/workflows-with-snakefile/tree/master/Exercises/Exercise3>

# Modular workflows



```
# In main Snakefile:
include: "Snakefile.qc"
```

```

# Snakefile.qc
rule summarized_qc:
    input: ...
    output: touch("qc.done")
    shell: ...

rule qc_one_sample:
    input: ...
    output: ...
    shell: ...

rule raw_sample_qc:
    input: ...
    output: ...
  
```

<https://snakemake.readthedocs.io/en/stable/>

# Developing complex workflows

1. Define “skeleton” of workflow, link rules together using `touch()`.
2. Start at beginning, implementing one rule at a time, testing as you go.
3. Use a small data set for testing, fast feedback
4. Implement the cluster configuration
5. Re-test
6. Run it with real data set

# Granularity

- Fine-grained
  - Many rules, simple shell statements
  - Efficient for local rules, easy debugging
  - Inefficient for cluster jobs, as each rule requires submitting a job
- Coarse-grained
  - Few rules, complex shell statements
  - More efficient on clusters

## Handling batches

- On a cluster, the snakemake paradigm maps the execution of one rule to one SLURM job – this may not fit your work flow well
  - Rule execution may be too small to fully occupy a node
  - Wait time in the SLURM queue on a busy cluster
- Solutions:
  - Write rules that process batches of samples or values
  - Use shared partitions in SLURM



## Snakemake is container-friendly

- Snakemake supports running code in containers using singularity
- See: <https://snakemake.readthedocs.io/...#running-jobs-in-containers>

# Snakemake may not be right for you

- What if your inputs and outputs aren't files?
- What if your cluster doesn't use SLURM or LSF?
  - HTCondor (Open Science Grid: > 1.2 billion core hours last year)
- What if your workflow changes?
- nextflow: <https://www.nextflow.io/>
  - non-file inputs and outputs
  - support for HTCondor (OSG) and many other schedulers
  - workflow file is part of the workflow – when a rule changes, it gets re-run