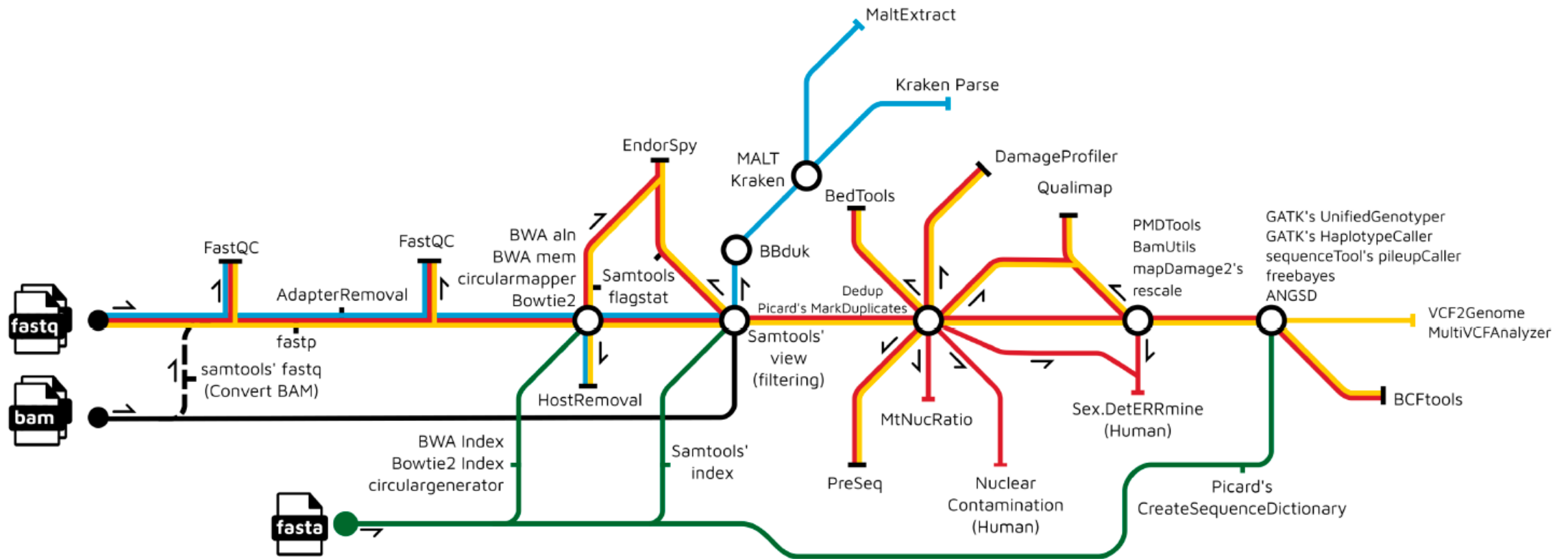


The logo for Nextflow, featuring the word "next" in a green, stylized font with a 3D effect, and "flow" in a black, sans-serif font. The text is set against a light gray rounded rectangular background.

nextflow

**Reproducible, Scalable, and Shareable
Data Analysis Workflows**

Writing modern workflows is complex

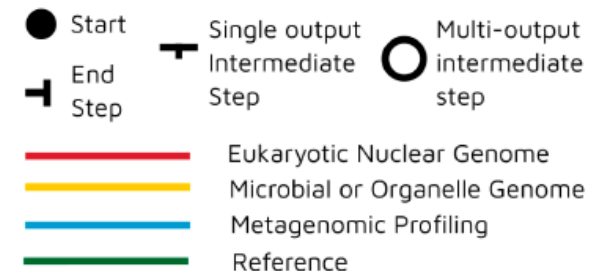


nf-core/eager v2.4

Example analysis pathways

state-of-the-art ancient DNA analysis pipeline

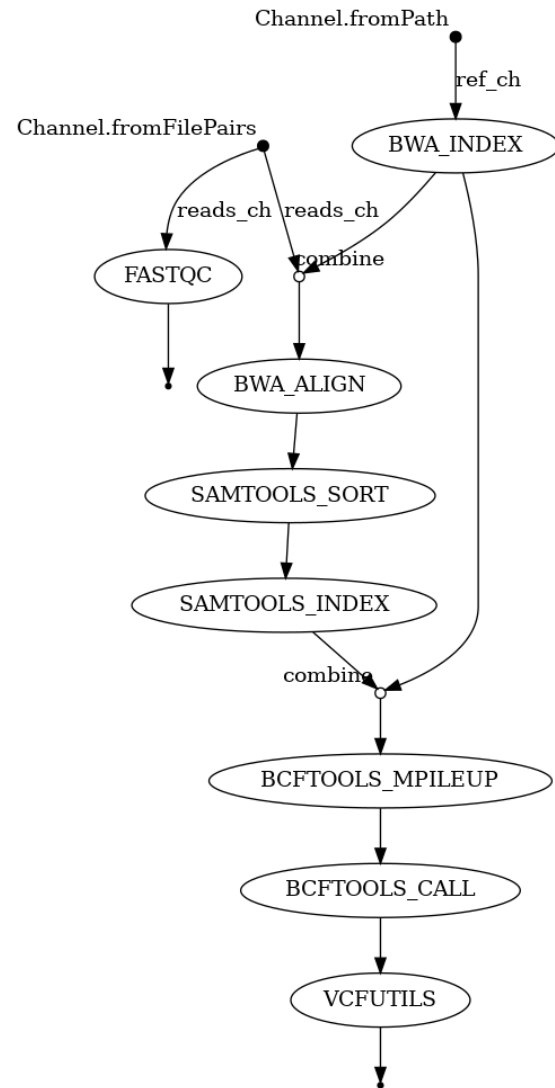
Legend



Data analysis workflows

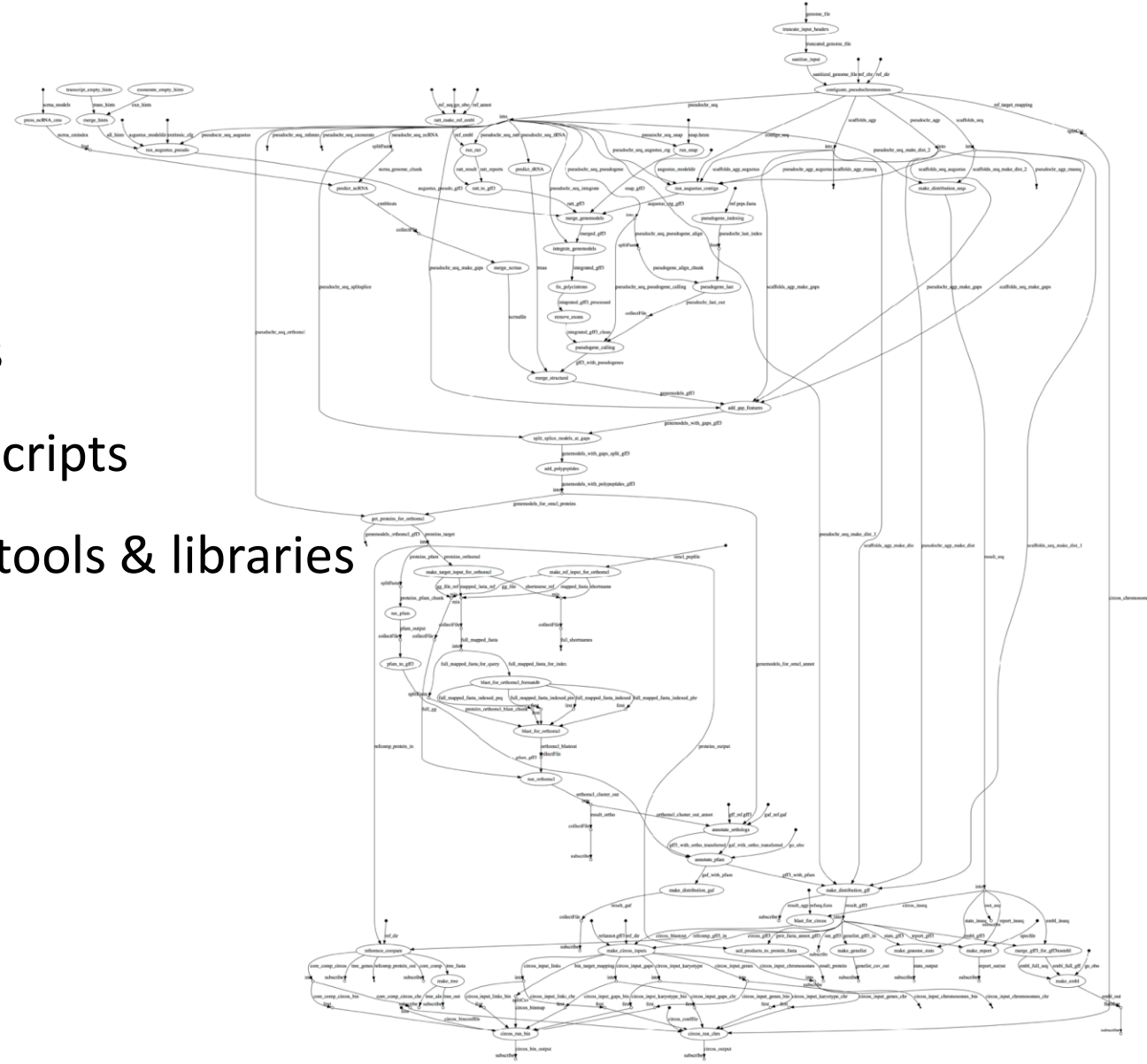
- **Data analysis applications perform computation to generate information from (large) datasets**
- **Embarrassingly parallel! → can spawn 100s-100k jobs over distributed cluster**
- **Mash-up of many different tools and scripts (dependencies!)**
- **Complex dependency trees and configuration → very fragile ecosystem!**

A lot of moving parts!



Variant-Calling pipeline

- 70 processes
- 55 external scripts
- 39 software tools & libraries



Steinbiss et al., Companion parasite genome annotation pipeline, DOI: 10.1093/nar/gkw292

To reproduce the result of a typical computational biology experiment requires 280 hours.



OPEN ACCESS Freely available online



Quantifying Reproducibility in Computational Biology: The Case of the Tuberculosis Drugome

Daniel Garijo¹, Sarah Kinnings², Li Xie³, Lei Xie⁴, Yinliang Zhang⁵, Philip E. Bourne^{3*}, Yolanda Gil^{6*}

1 Ontology Engineering Group, Facultad de Informática, Universidad Politécnica de Madrid, Madrid, Spain, **2** Department of Chemistry and Biochemistry, University of California San Diego, La Jolla, California, United States of America, **3** Skaggs School of Pharmacy and Pharmaceutical Sciences, University of California San Diego, La Jolla, California, United States of America, **4** Department of Computer Science, Hunter College, The City University of New York, New York, New York, United States of America, **5** School of Life Sciences, University of Science and Technology of China, Hefei, Anhui, China, **6** Information Sciences Institute and Department of Computer Science, University of Southern California, Los Angeles, California, United States of America

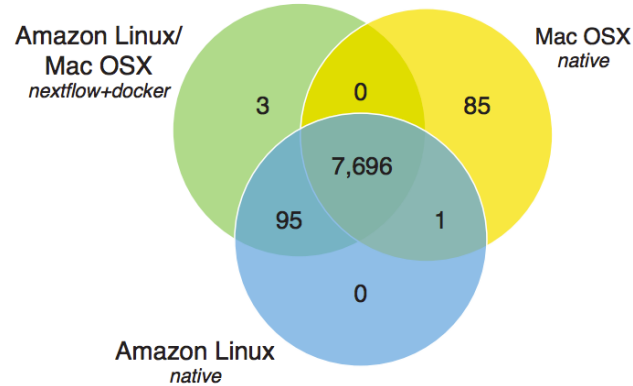
≈1.7 months!

The same application
deployed in
different environments
Produces
different results

Differences in genome annotations & expressed genes

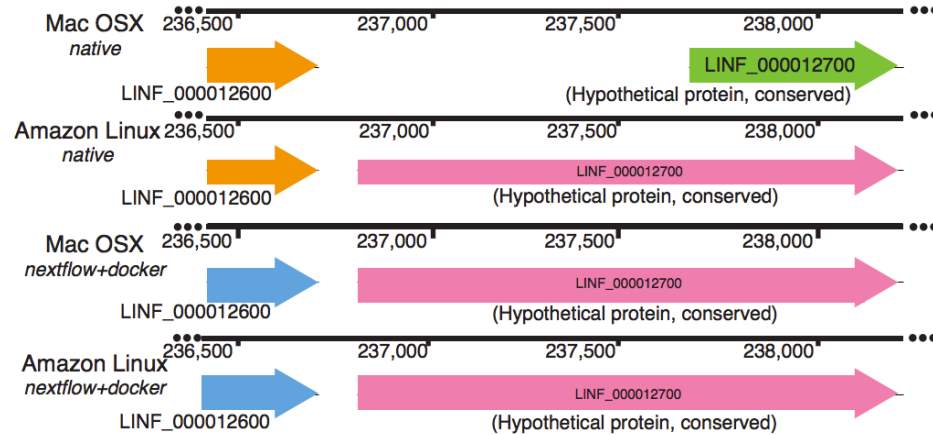
a

Gene annotation of *Leishmania infantum* with Companion



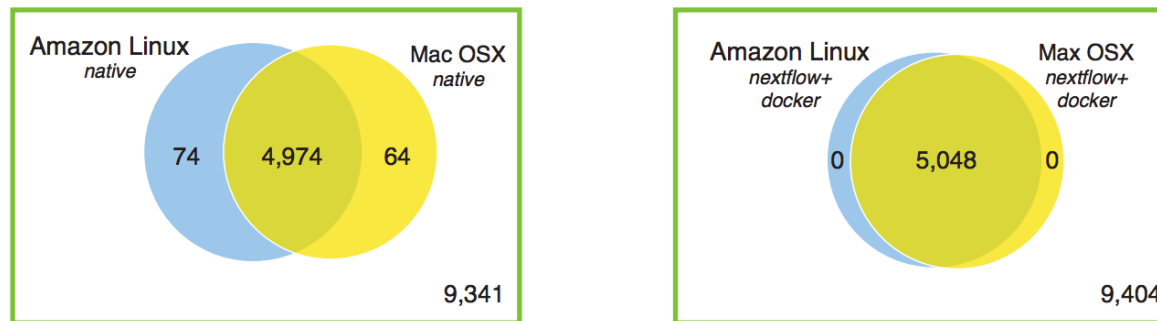
b

Leishmania infantum chromosome 1



c

Transcript quantification and differential expression with Kallisto and Sleuth



Why do Reproducible Research?

An article about computational results is advertising, not scholarship.

The actual scholarship is the full software environment, code and data, that produced the result

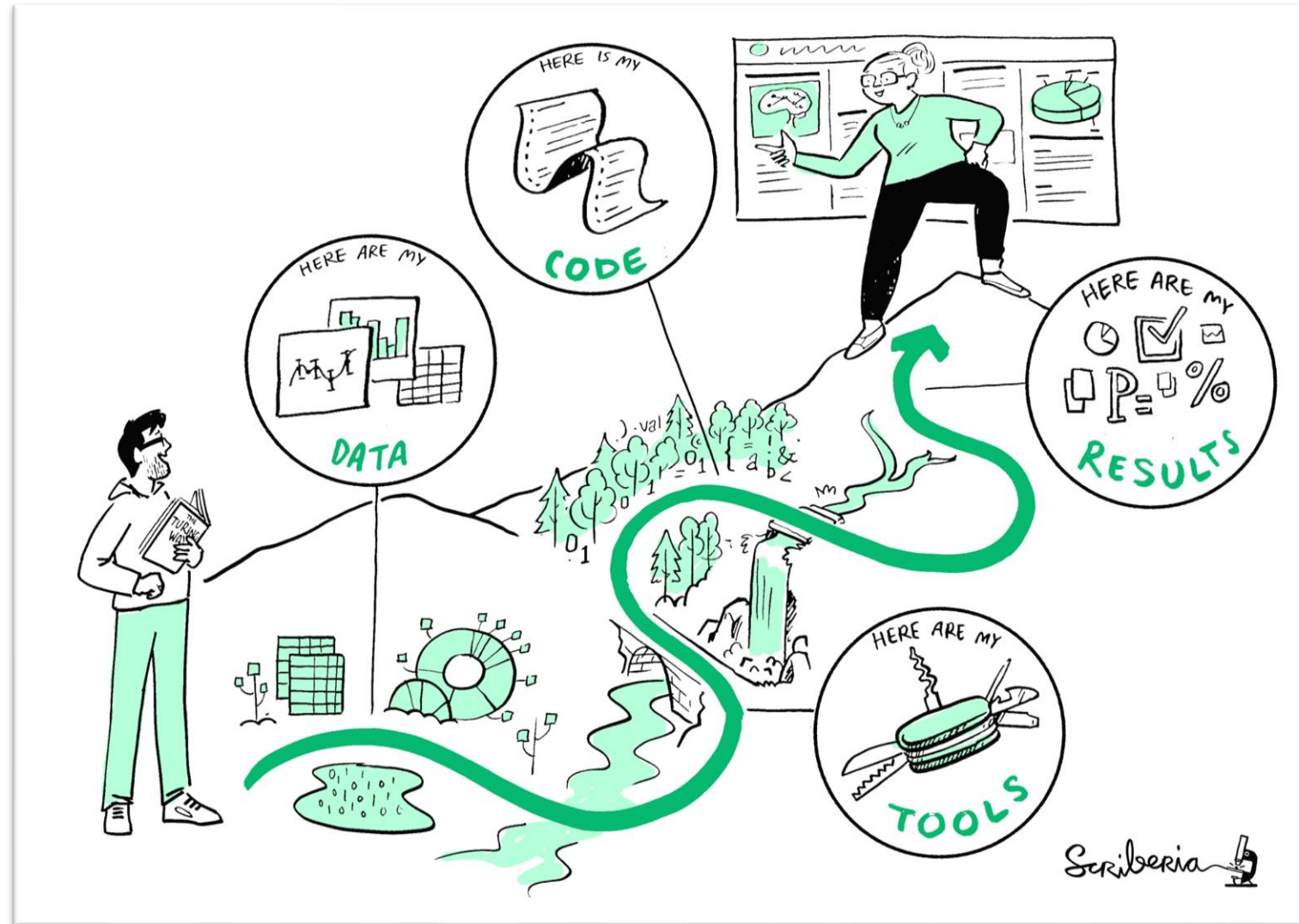
- **Claerbout & Karrenbach**

<https://library.seg.org/doi/abs/10.1190/1.1822162>, doi:<https://doi.org/10.1190/1.1822162>

Authors **need to** provide all the necessary data and the computer codes to run the analysis again, re-creating the results.

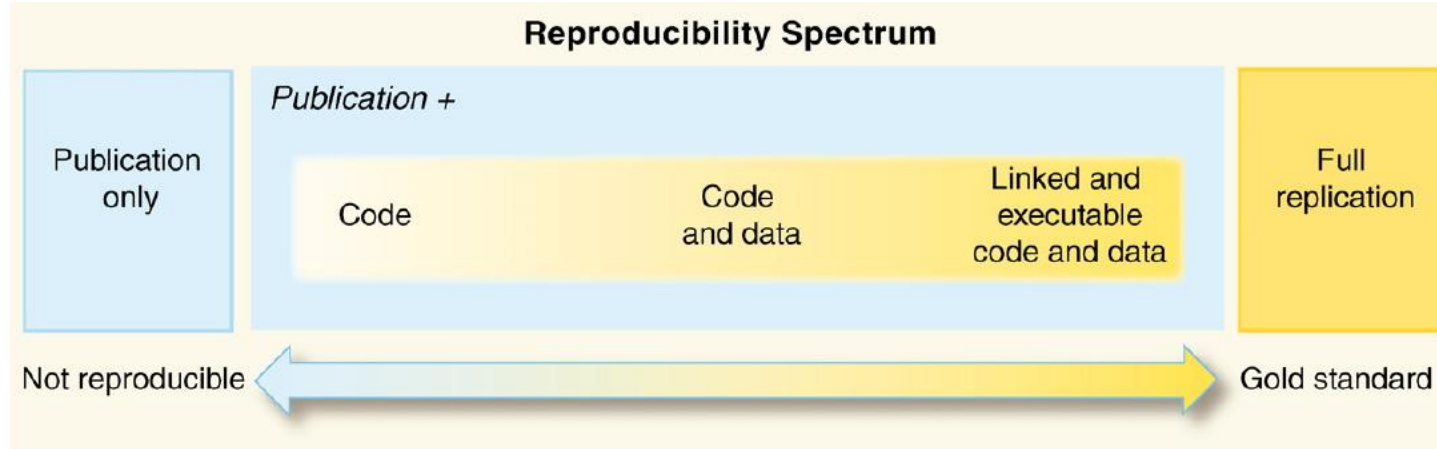


covers topics related to skills, tools and best practices for research reproducibility.



The Turing Way project illustration by Scriberia. Used under a CC-BY 4.0 licence. DOI: [10.5281/zenodo.3332807](https://doi.org/10.5281/zenodo.3332807).
<https://the-turing-way.netlify.app/reproducible-research/>

What does this actually look like?



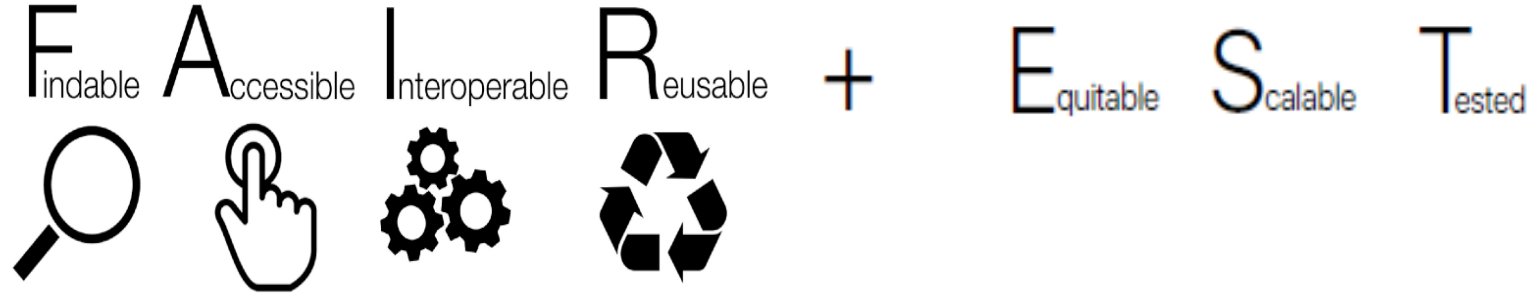
<https://science.sciencemag.org/content/334/6060/1226>



Reproducibility Checklist

- ☐ Uses Code – instead of pointing and clicking
- ☐ Data archived and available
- ☐ Open-Source – Code transparency is key to reproducibility
- ☐ Version tracking – using version-control software as Git and Github/Gitlab
- ☐ Replicate your environment – using containers or conda
- ☐ Analyses documented – Jupyter Notebooks or Rmarkdown
- ☐ Automated – workflow management tool (Snakemake, CWL, Nextflow)

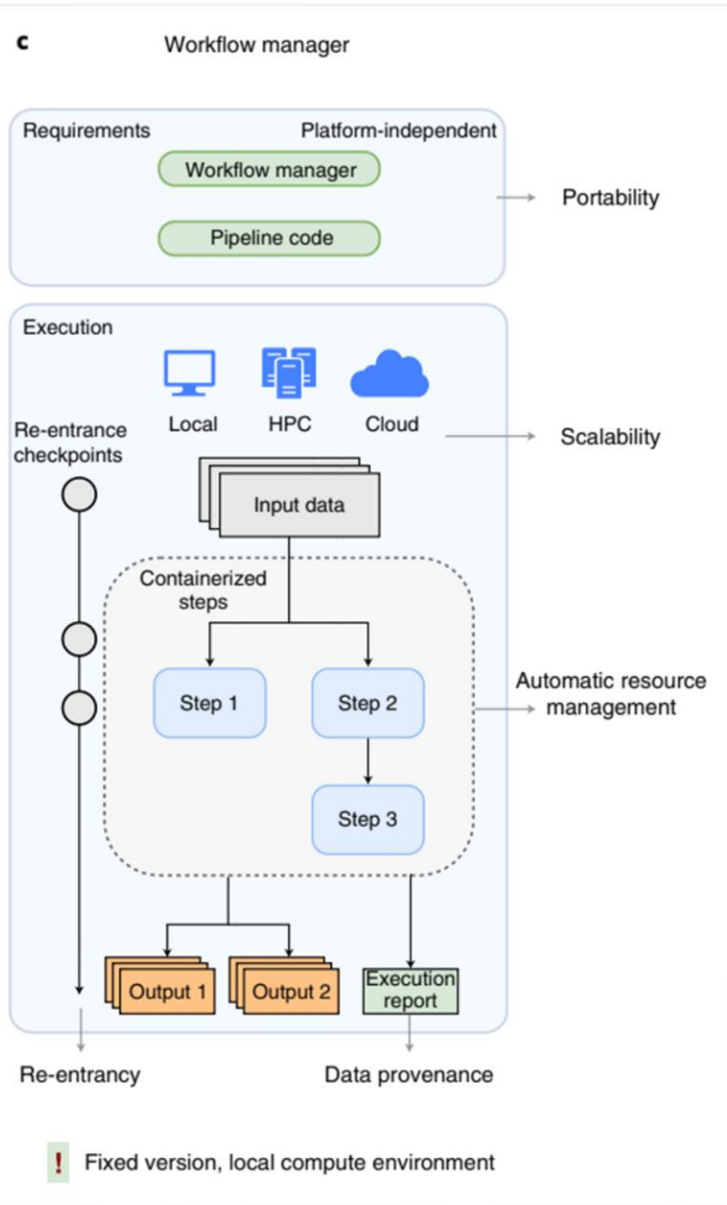
What should be our goal?



Streamlining data analysis to make it:

- Findable \Rightarrow openly available and searchable through repositories
- Accessible \Rightarrow can be used by everyone
- Interoperable \Rightarrow portable to any cluster & cloud
- Reusable \Rightarrow in a reproducible way
- Equitable \Rightarrow free of bias
- Scalable \Rightarrow from laptop to supercomputer
- Tested \Rightarrow validated using modern software practices

Scientific Workflow Management Systems



(WfMSs) automate computational analyses by stringing together individual data processing tasks into cohesive pipelines.

They abstract away the issues of orchestrating data movement and processing, managing task dependencies, and allocating resources within the compute infrastructure.

Aspects of WfMSs relevant to bioinformatics

- **Modularity of the pipeline to enable checkpointing**
- **Scalability with respect to the number of tasks in the pipeline**
- **Robustness against failures due to data issues, resource unavailability, or aborted execution**
- **Reproducibility via logs recording data provenance and task execution**
- **Portability across compute environments**
- **Ease of development by users with a range of experience and computational knowledge.**

What is Nextflow?

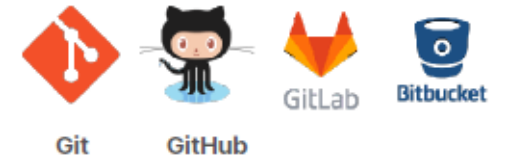
nextflow *script*

Write code
in any language



Define orchestration with
dataflow programming

Define software
dependencies
with containers



Version control

nextflow *runtime*

Orchestration of tasks to
deploy anywhere with ease

Supports all major platforms



**Nextflow is
a language,
a runtime and
a community**

<https://nextflow.io/>

How does it work?

- **Fast prototyping** \Rightarrow custom DSL that enables task composition, simplifies most use cases + general purpose programming language for corner cases
- **Easy parallelization** \Rightarrow declarative reactive programming model based on dataflow paradigm, implicit portable parallelism
- **Self-contained** \Rightarrow functional approach, a task execution is idempotent i.e., cannot modify the state of other tasks + isolate dependencies with containers
- **Portable deployments** \Rightarrow executor abstraction layer + deployment configuration from implementation logic

Task example

```
bwa mem reference.fa sample.fq \  
| samtools sort -o sample.bam
```


Task example

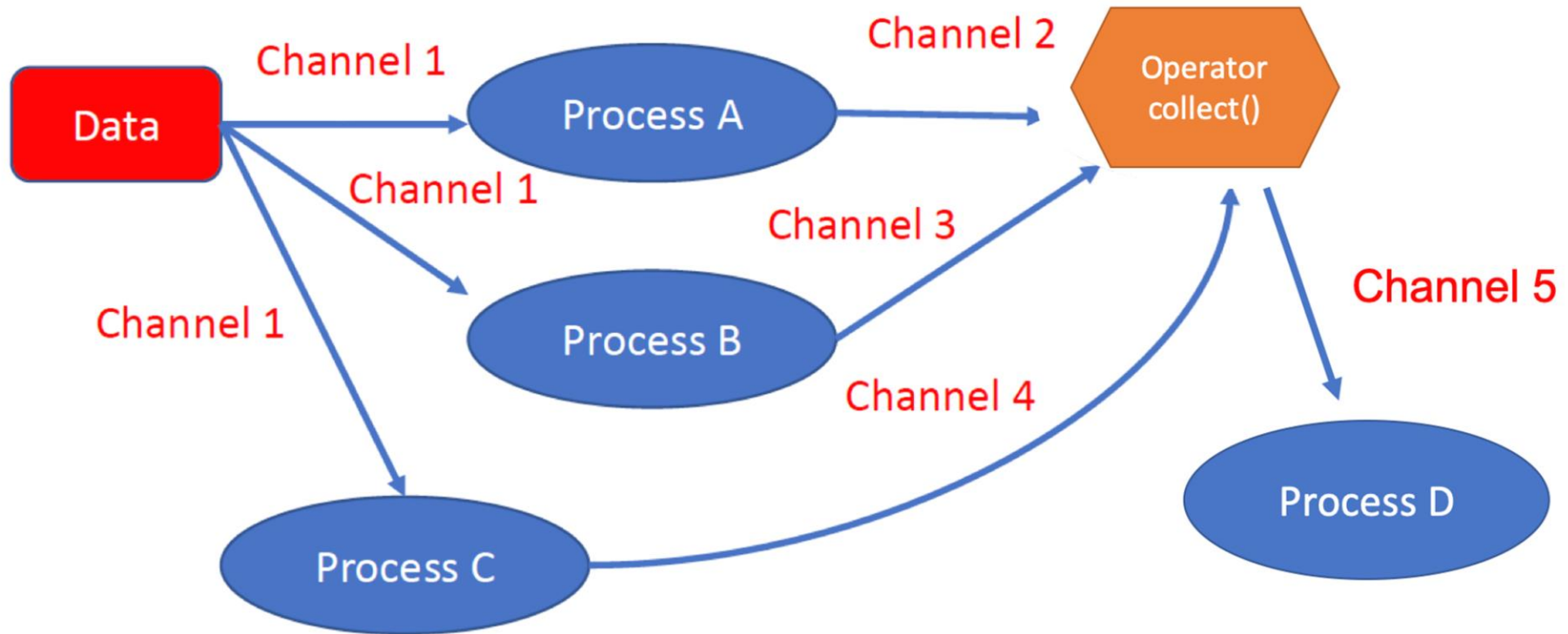
```
process align_sample {  
  -----> input:  
    file 'reference.fa' from genome_ch  
    file 'sample.fq' from reads_ch  
  -----> output:  
    file 'sample.bam' into bam_ch  
  -----> script:  
    ""  
    bwa mem reference.fa sample.fq \  
    | | | | | samtools sort -o sample.bam  
    ""  
}
```

Task Composition

```
process align_sample {  
  input:  
  file 'reference.fa' from genome_ch  
  file 'sample.fq' from reads_ch  
  output:  
  file 'sample.bam' into bam_ch  
  script:  
  """  
  bwa mem reference.fa sample.fq \  
  | | | | | samtools sort -o sample.bam  
  """  
}
```

```
process index_sample {  
  input:  
  file 'sample.bam' from bam_ch  
  output:  
  file 'sample.bai' into bai_ch  
  script:  
  """  
  samtools index sample.bam  
  """  
}
```

Workflow



Nextflow Syntax

task

```
process QUANT {  
  input:  
  path index  
  tuple val(pair_id), path(reads)  
  
  output:  
  path pair_id  
  
  script:  
  """  
  salmon quant -i $index \  
    -1 ${reads[0]} \  
    -2 ${reads[1]} \  
    -o $pair_id  
  """  
}
```

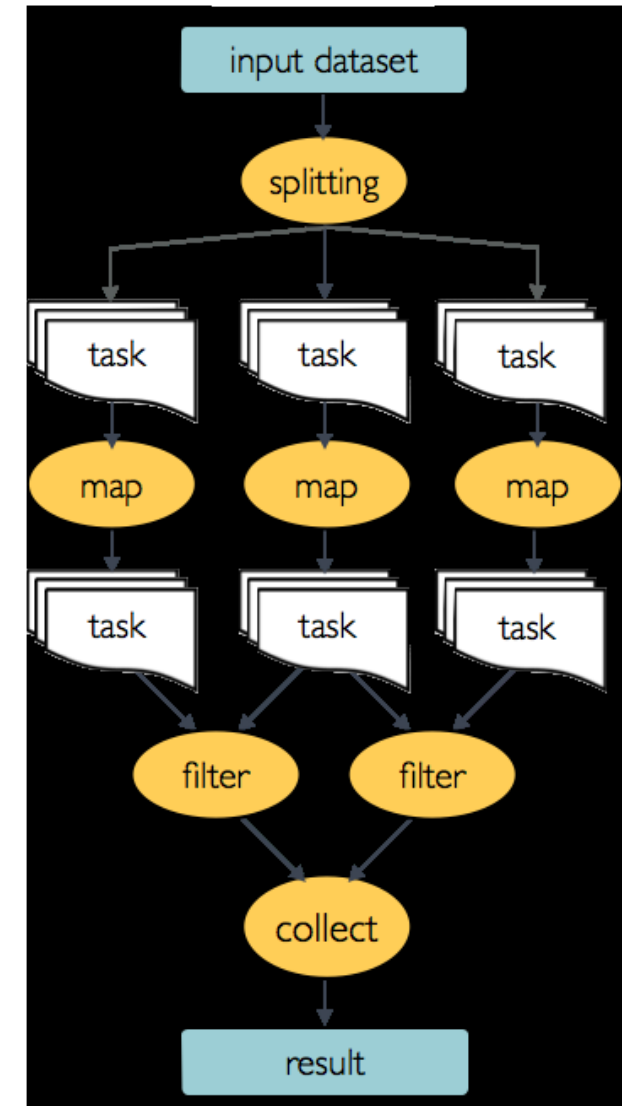
workflow

```
params.outdir = 'results'  
  
include { INDEX } from './index'  
include { QUANT } from './quant'  
include { FASTQC } from './fastqc'  
  
workflow RNASEQ {  
  take:  
    transcriptome  
    read_pairs_ch  
  
  main:  
    INDEX(transcriptome)  
    FASTQC(read_pairs_ch)  
    QUANT(INDEX.out, read_pairs_ch)  
  
  emit:  
    QUANT.out | concat(FASTQC.out) | collect  
}
```

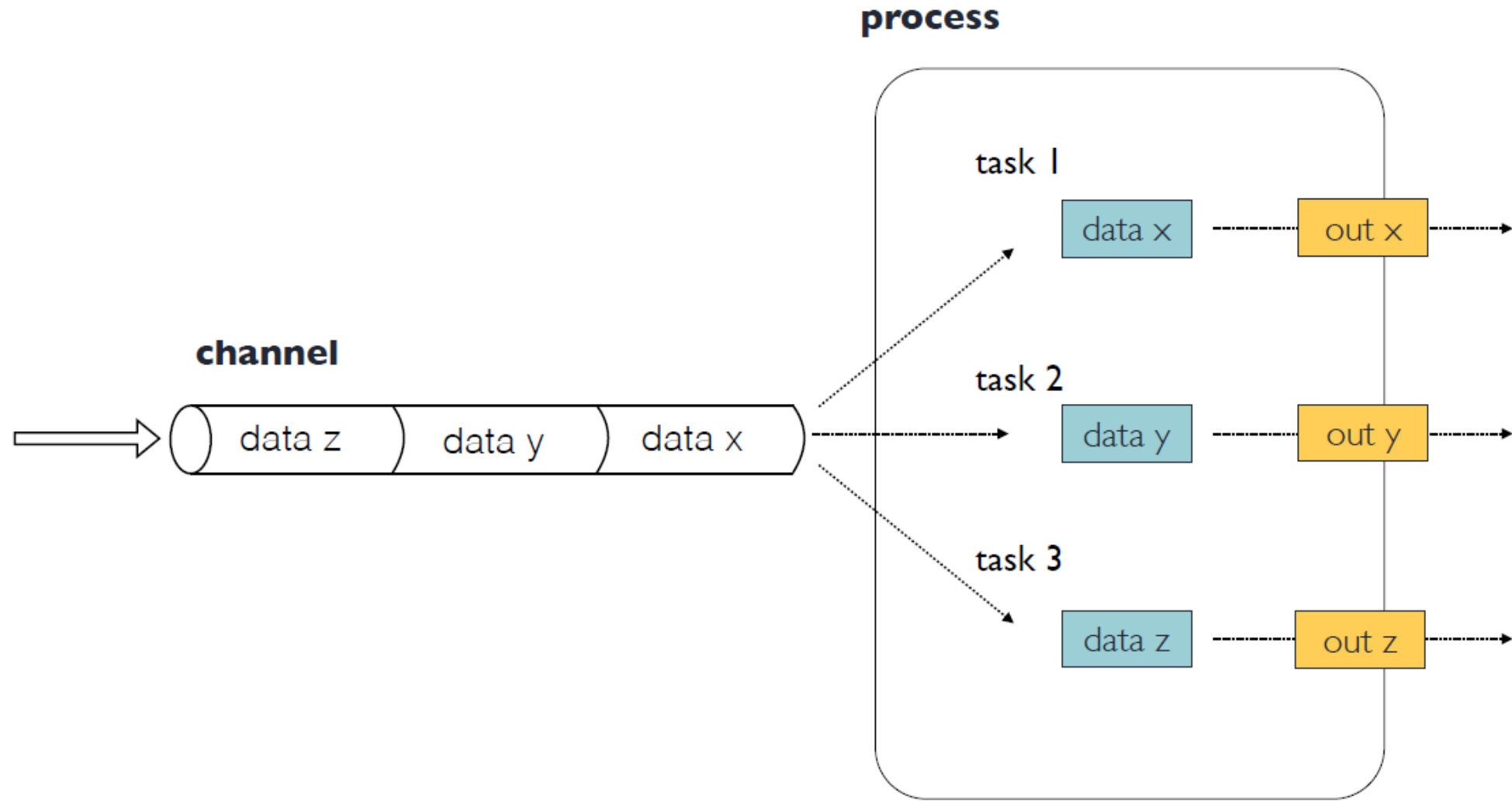
Scientists and engineers can now write complex, distributed and parallel data pipelines without requiring a degree in computer science.

Dataflow concepts

- **Declarative computational model for parallel process executions**
- **Processes wait for data, when an input set is ready the process is executed**
- **They communicate by using dataflow variables i.e., async FIFO queues called channels**
- **Parallelization and tasks dependencies are implicitly defined by process in/out declarations**



How does parallelization work?



How does parallelization work?

```
samples_ch = Channel.fromPath('data/sample.fastq')

process FASTQC {

  input:
  file reads from samples_ch
  output:
  file 'fastqc_logs' into fastqc_ch

  script:
  """
  mkdir fastqc_logs
  fastqc -o fastqc_logs -f fastq -q ${reads}
  """
}
```

```
samples_ch = Channel.fromPath('data/*.fastq')
```

```
process FASTQC {

  input:
  file reads from samples_ch
  output:
  file 'fastqc_logs' into fastqc_ch

  script:
  """
  mkdir fastqc_logs
  fastqc -o fastqc_logs -f fastq -q ${reads}
  """
}
```

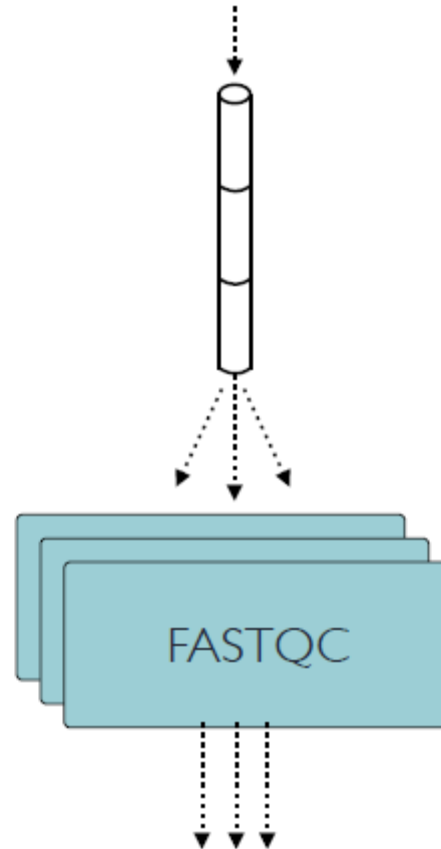
```
reads_ch = Channel.fromFilePairs( "data/*_{1,2}.fastq.gz", checkIfExists: true )
```



```
[SRR2584863, [/data/SRR2584863_1.fastq.gz, /data/SRR2584863_2.fastq.gz]]
[SRR2584866, [/data/SRR2584866_1.fastq.gz, /data/SRR2584866_2.fastq.gz]]
[SRR2589044, [/data/SRR2589044_1.fastq.gz, /data/SRR2589044_2.fastq.gz]]
```


Implicit parallelism

```
Channel.fromPath("data/*.fastq")
```

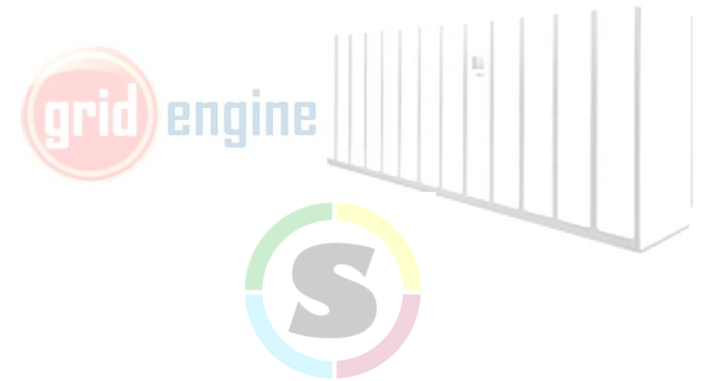


CWL vs Nextflow

- Language specification
- Declarative meta-language (YAML/JSON)
- Verbose
- Committee driven
- Many vendors/implementations (and specification version)
- Language + app. runtime
- DSL on top of a general purpose programming lang.
- Concise, fluent
- Community driven
- Single implementation, quick iterations

Snakemake vs Nextflow

- Command line oriented tool
 - Pull model
 - Rules defined using file name patterns
 - Compute DAG ahead
 - Built-in support for Singularity/Docker
 - Custom scripts for cluster deployments
 - No support for source code management system
 - Python based
- Command line oriented tool
 - Push model
 - Can manage any data structure
 - Compute DAG at runtime
 - All major container runtimes
 - Built-in support for clusters and cloud
 - Built-in support for Git/GitHub, etc., manage pipeline revisions
 - Groovy/JVM based



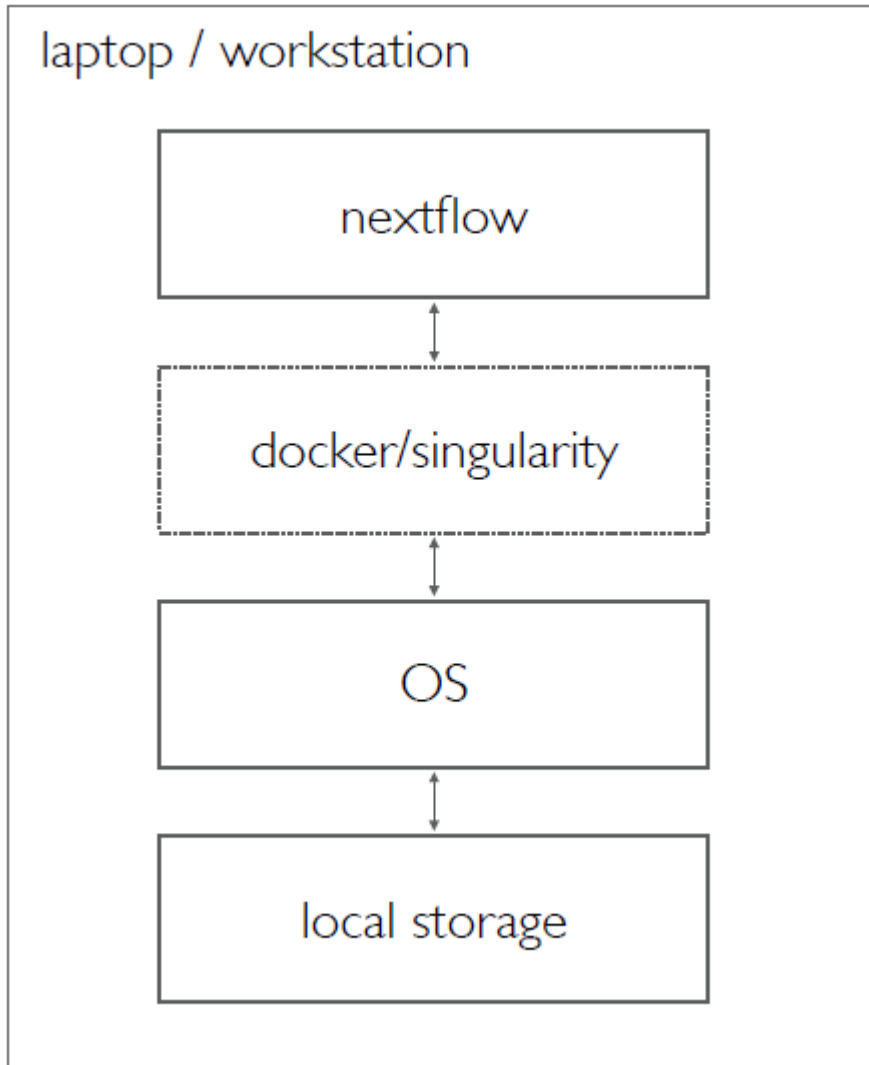
Deployment Scenarios



nextflow



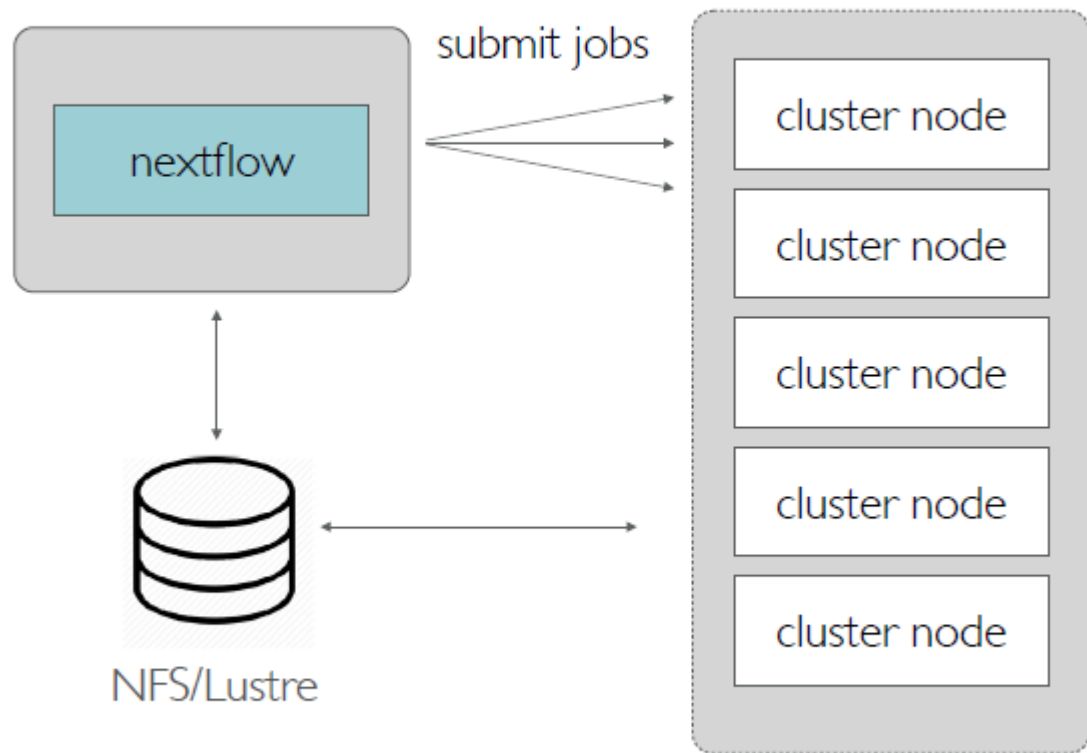
Local execution



- **Common development scenario**
- **Dependencies can be managed using a container runtime**
- **Parallelization is managed by spawning posix processes**
- **Can scale vertically using fat server / shared mem. machine**

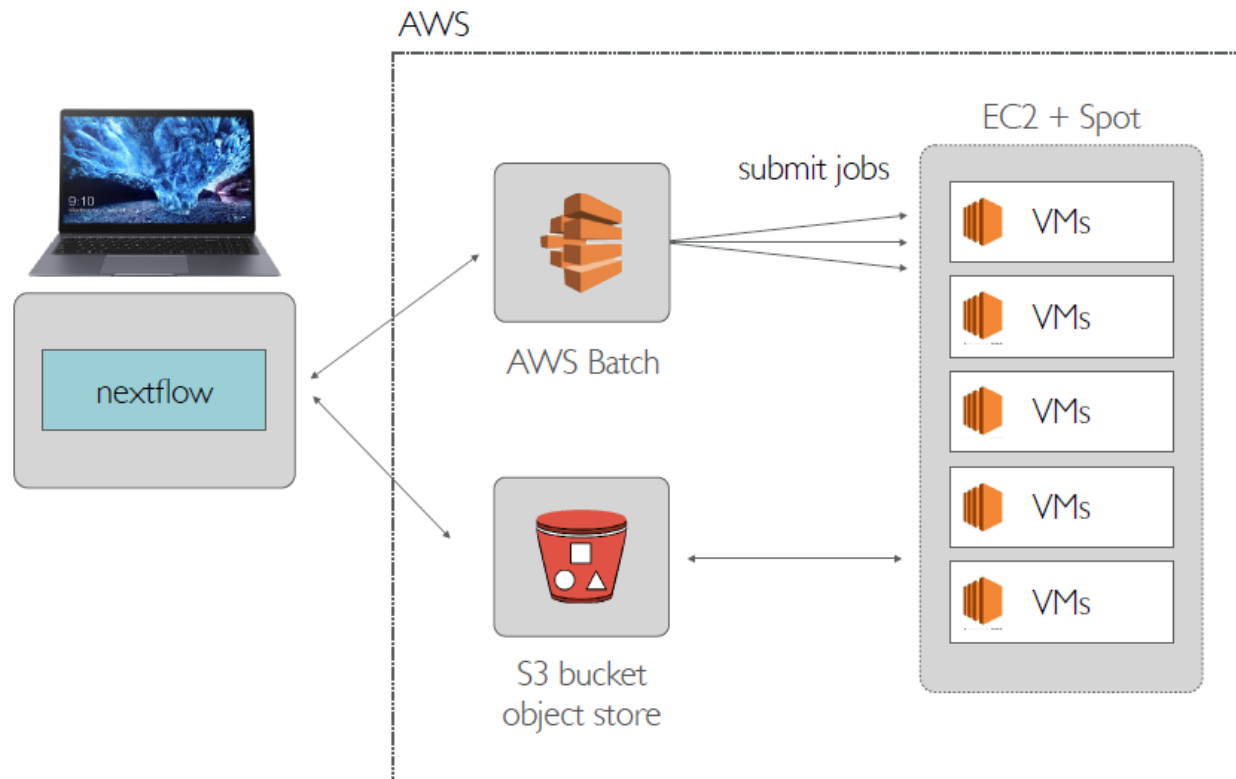
Centralized cluster orchestration

compute cluster



- **Nextflow orchestrates workflow execution submitting jobs to a compute scheduler.**
- **Can run in the head node or a compute node.**
- **Requires a shared storage to exchange data between tasks.**
- **Ideal for coarse-grained parallelism.**

Cloud batch orchestration



- **Nextflow orchestrates workflow execution via AWS**

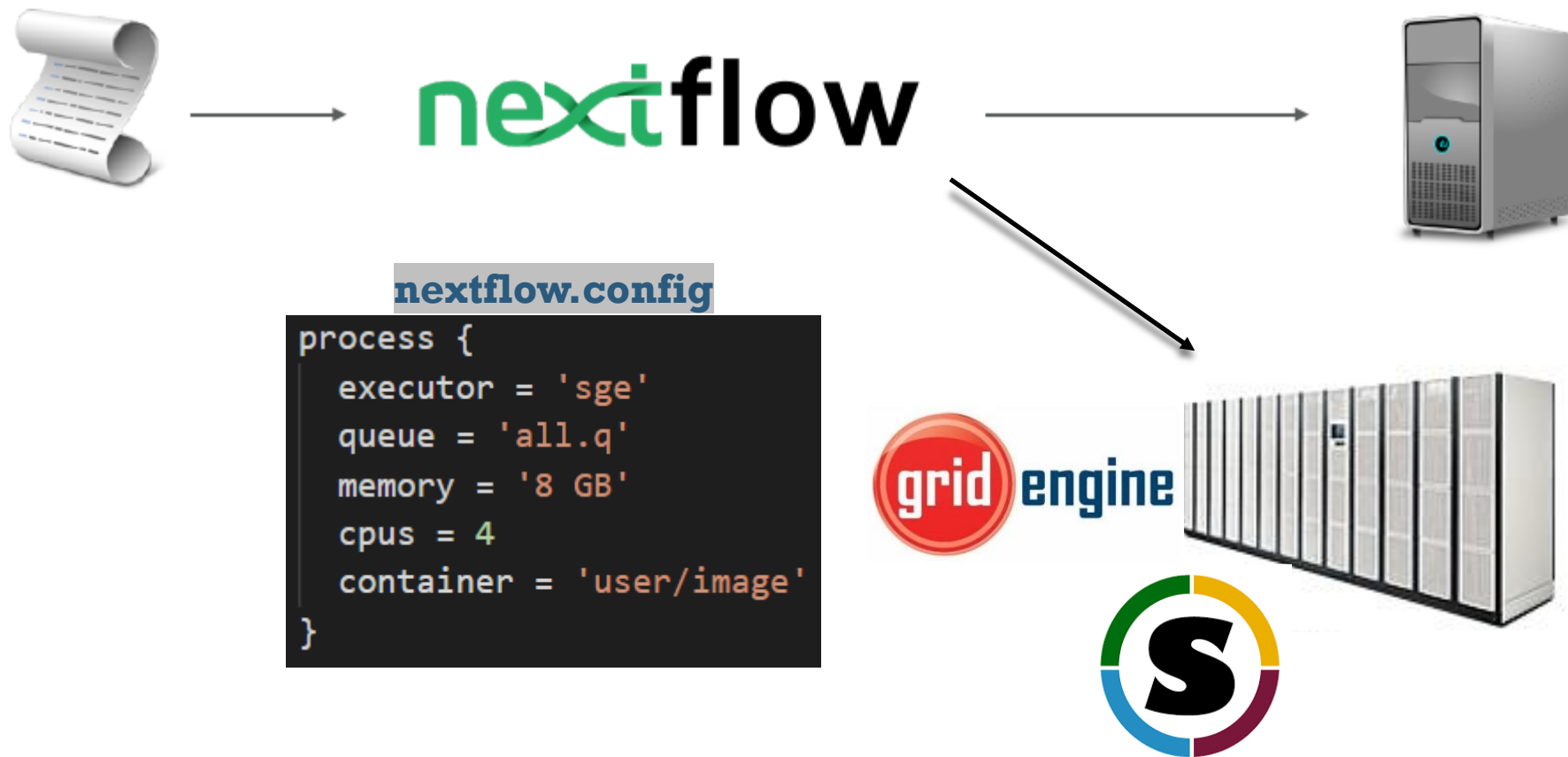
Batch

- **Launch workflow from anywhere into the cloud**
- **Transfer of data between local environment and cloud storage**
- **Requires a shared object storage to exchange data between VMs.**

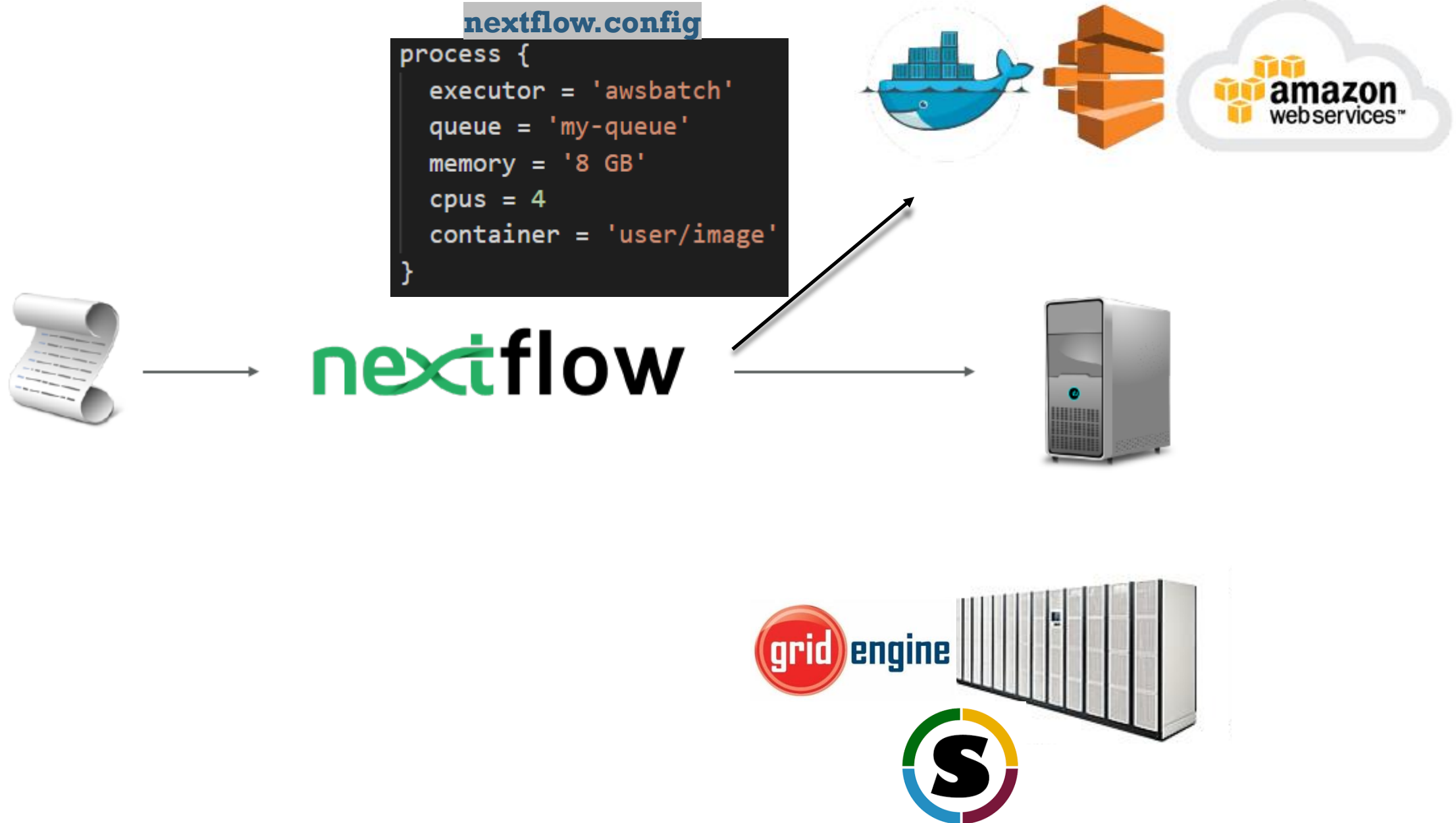
Portability



Portability

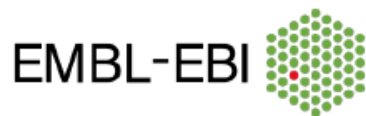


Portability



**Configuration
Decoupling
is the key to
portable deployments**

Widespread enterprise adoption



Container vs. VM

- Lighter: MB vs GB
- Faster startup: ms/secs vs minutes
- Virtualize a process/application instead of OS/Hardware
- Immutable: don't change over time, thus guarantee replicability over executions.
- Composable: the output of one container is directly consumable as input by another container.
- Transparent: they are created with a well-defined automated procedure.



Nextflow Supports:

Conda environments
Docker containers
Shifter Containers
Singularity containers
Podman containers
Charliecloud containers

When use
containers?
Always!

Caching and Checkpointing

A key feature of Nextflow is **re-entrancy** which is the ability to restart a pipeline after an error from the last successful process.

```
nextflow run word_count.nf --input 'data/untrimmed_fastq/*.fastq.gz' -resume
```

```
N E X T F L O W ~ version 21.04.3
Launching `word_count.nf` [chaotic_knuth] - revision: 65cc76d410
[dd/c19734] process > NUM_LINES (5) [100%] 6 of 6, cached: 6 ✓
SRR2589044_2.fastq.gz4428360
SRR2584863_1.fastq.gz6213036
SRR2589044_1.fastq.gz4428360
SRR2584866_1.fastq.gz11073592
SRR2584866_2.fastq.gz11073592
SRR2584863_2.fastq.gz6213036
```


Reporting

Nextflow workflow report

[berserk_raman]

Workflow execution completed successfully!

Run times

16-Jan-2022 19:32:12 - 16-Jan-2022 19:41:13 (duration: **9m 1s**)

22 succeeded

Nextflow command

```
nextflow run variant-calling.nf -profile singularity,sge,short --output results
```

CPU-Hours

0.3

Launch directory

/scicomp/home-pure/rrx8/nextflow_tutorial

Work directory

/scicomp/home-pure/rrx8/nextflow_tutorial/work

Project directory

/scicomp/home-pure/rrx8/nextflow_tutorial

Script name

variant-calling.nf

Script ID

ce089224dca2fceb6160974152132ef

Workflow session

a7d786a3-bb3c-494b-9b26-406e5a322bbf

Workflow profile

singularity,sge,short

Nextflow version

version 21.04.3, build 5560 (21-07-2021 15:09 UTC)

Job Duration

Raw Usage % Allocated

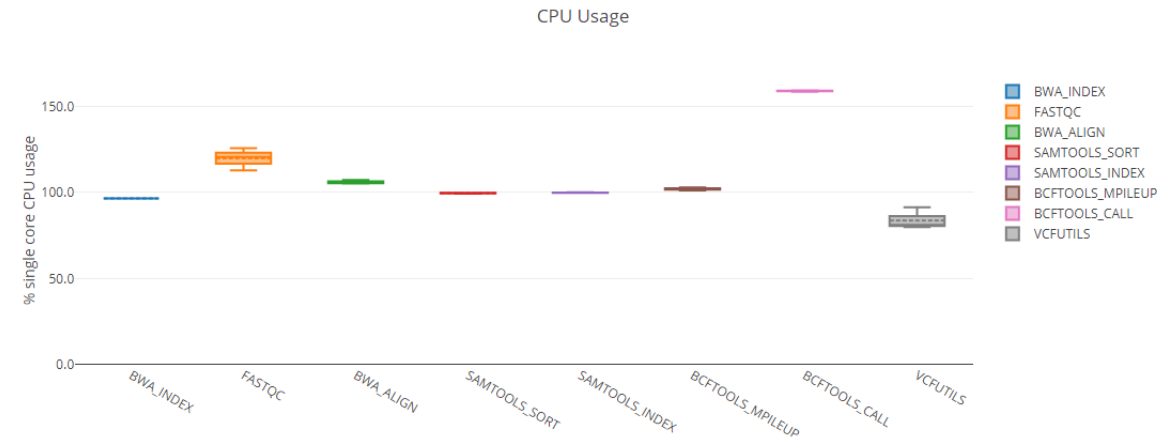


Resource Usage

These plots give an overview of the distribution of resource usage for each process.

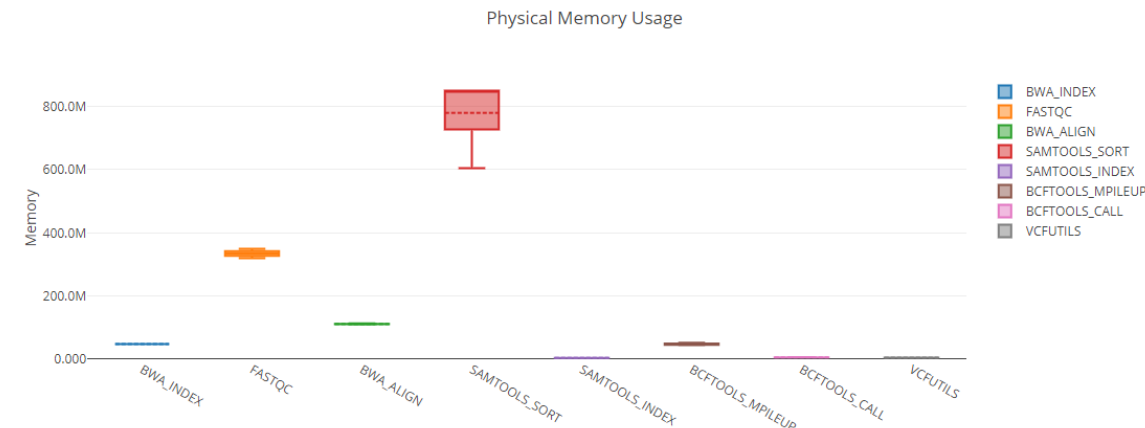
CPU

Raw Usage % Allocated



Memory

Physical (RAM) Virtual (RAM + Disk swap) % RAM Allocated



Reporting

Tasks

This table shows information about each task in the workflow. Use the search box on the right to filter rows for specific values. Clicking headers will sort the table by that value and scrolling side to side will reveal more columns.

Values shown as:

Human readable

Show

25

 entries

Filter:

Metrics

Metadata

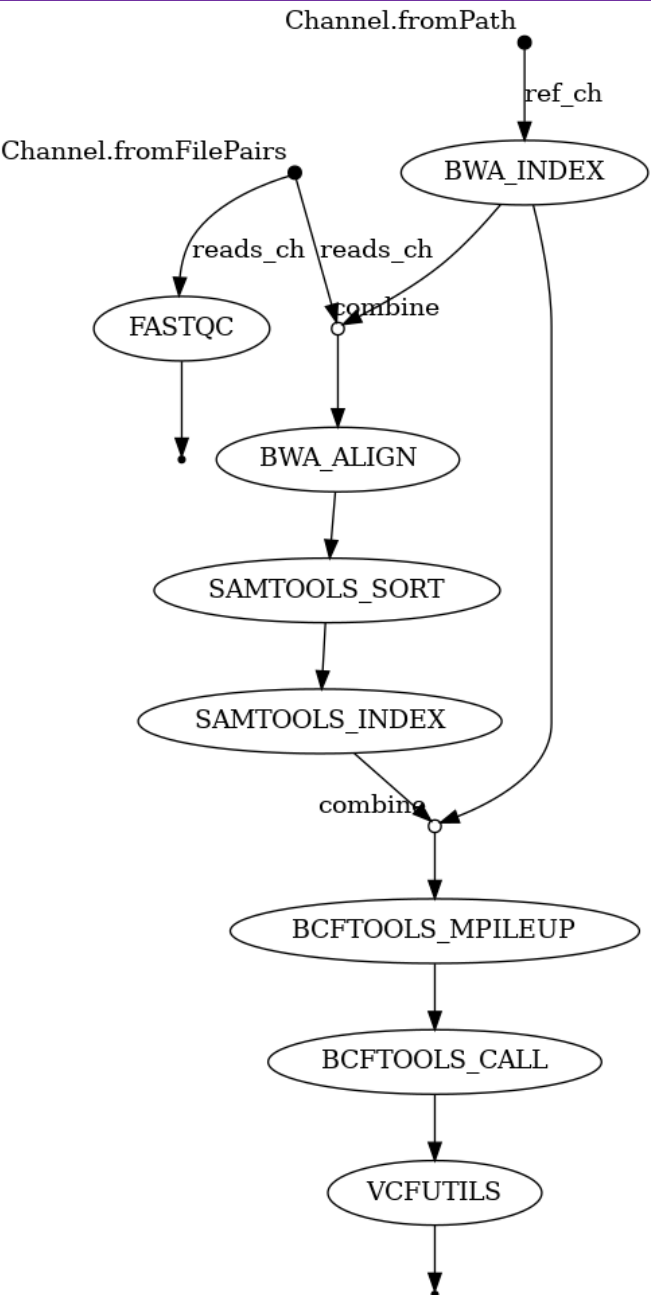
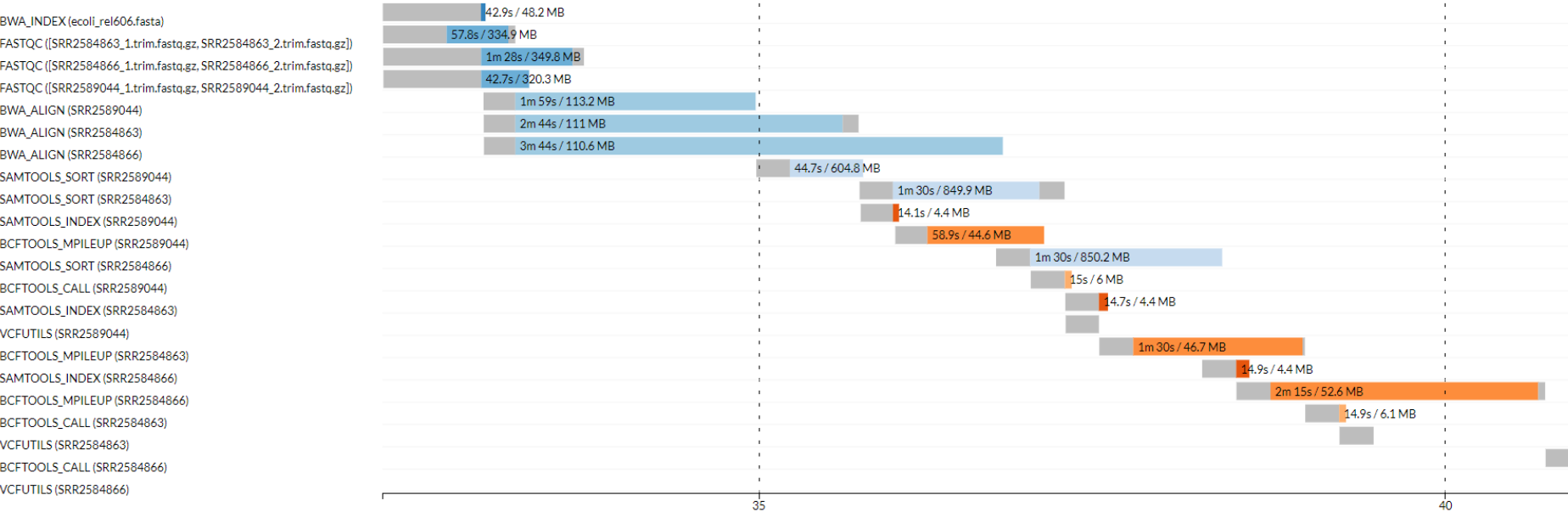
All

 Search:

task_id	process	tag	status	allocated cpus	%cpu	allocated memory	%mem	vmem	rss	peak
1	BWA_INDEX	ecoli_rel606.fasta	COMPLETED	1	96.6	-	0.0	31.230 MB	6.836 MB	69.0
2	FASTQC	[SRR2584863_1.trim.fastq.gz, SRR2584863_2.trim.fastq.gz]	COMPLETED	1	120.6	-	0.1	3.313 GB	327.680 MB	3.3

Processes execution timeline

Launch time: 16 Jan 2022 19:32
Elapsed time: 9m 2s
Legend: job wall time / memory usage (RAM)



<https://nf-co.re/>

nf-core



A community effort to collect a curated set of analysis pipelines built
using Nextflow.

Open-Source Community

55K+

monthly
downloads

10,000+

active developers
/month

35

international
Workshops

150K+

lines of
code

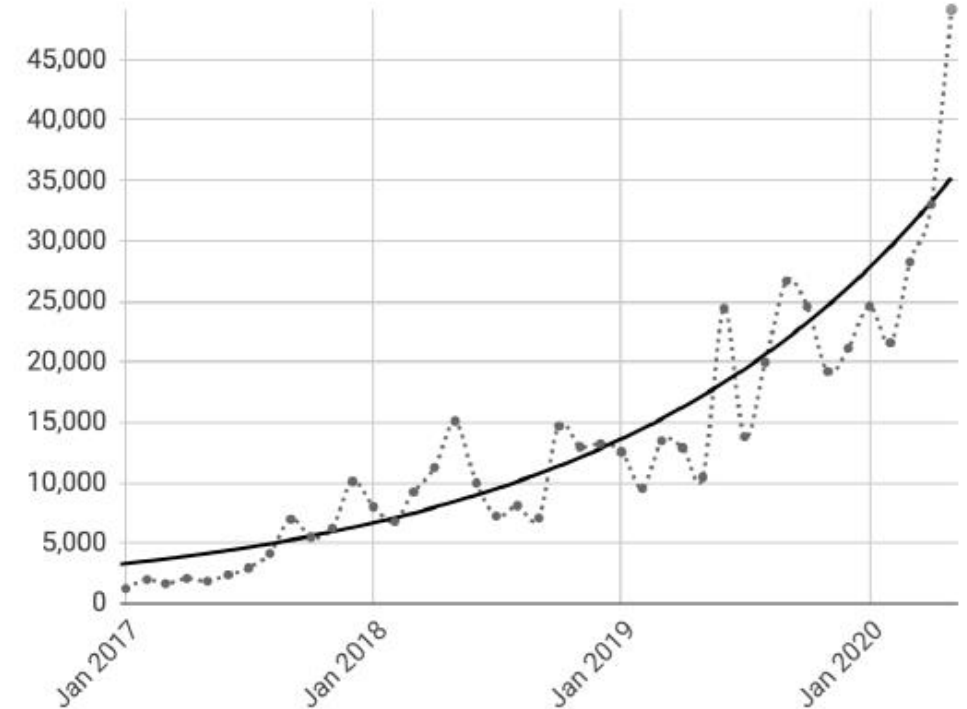
1.6K+

stars on
github

100+

contributors

monthly downloads



Over 2 million downloads

nf-core





Growing community content

Production ready analysis pipelines built with Nextflow.

Available Pipelines

Can you think of another pipeline that would fit in well? [Let us know!](#)

Search keywords: Filter: Released 33 Under development 20 Archived 5 Sort: Last Release Alphabetical Stars Display:  

nf-core/rnaseq ✓ ★ 414 rna rna-seq RNA sequencing analysis pipeline using STAR, RSEM, HISAT2 or Salmon with gene/isoform counts and extensive quality control. Version 3.5 Published 1 month ago	nf-core/sarek ✓ ★ 5 annotation cancer gatk4 genomics germline pre-processing somatic target-panels variant-calling whole-exome-sequencing whole-genome-sequencing Analysis pipeline to detect germline or somatic variants (pre-processing, variant calling and annotation) from WGS / targeted sequencing Version 2.7.1 Published 7 months ago
nf-core/atacseq ✓ ★ 99 atac-seq chromatin-accessibility ATAC-seq peak-calling, QC and differential analysis pipeline Version 1.2.1 Published 1 year ago	nf-core/chipseq ✓ ★ 1 chip chip-seq chromatin-immunoprecipitation macs2 peak-calling ChIP-seq peak-calling, QC and differential analysis pipeline. Version 1.2.2 Published 9 months ago
nf-core/mag ✓ ★ 77 annotation assembly binning long-read-sequencing metagenomes metagenomics nanopore nanopore-sequencing Assembly and binning of metagenomes Version 2.1.1 Published 2 months ago	nf-core/ampliseq ✓ ★ 1 16s amplicon-sequencing metagenomics qiime rrna 16S rRNA amplicon sequencing analysis workflow using QIIME2 Version 2.1.1 Published 3 months ago



<https://nf-co.re>

Deploy



Stable pipelines



Centralized configs



List and update pipelines



Download for offline use

Participate



Documentation



Slack workspace



Twitter updates



Hackathons

Develop



Starter template



Code guidelines



CI code linting and tests



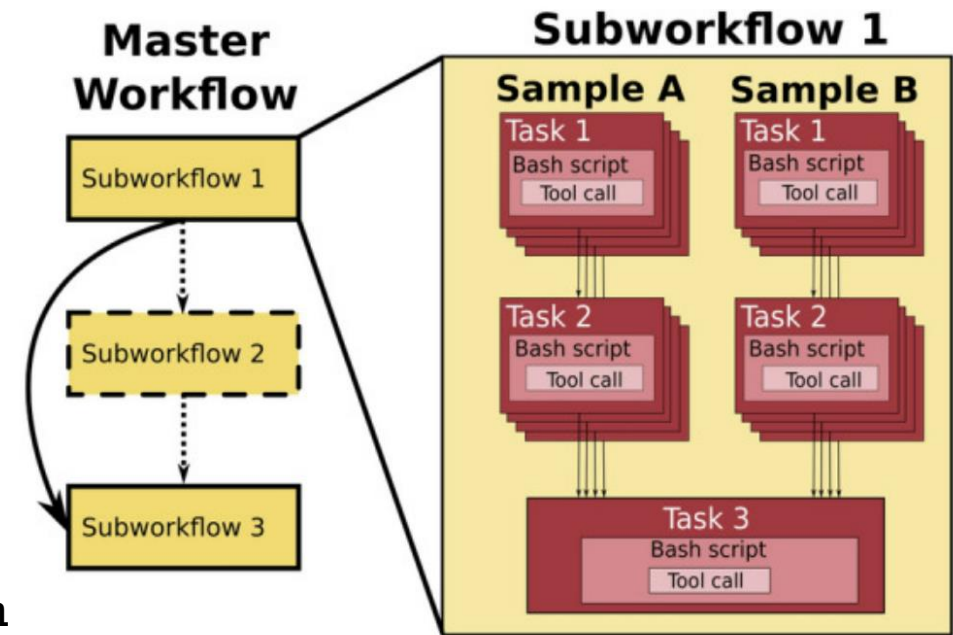
Helper tools

DSL2 Modules & Sub-Workflows = Re-Usability!

- Modularity is a very important design principle for production bioinformatics workflows.
- The core idea is to build a library of reusable modules (tasks or sub-workflows) and assemble them into various master workflows.

This enables -

- (1) performing different analyses without having to refactor the entire workflow
- (2) check-pointing and restart of a workflow run from a task in the middle of analysis if needed
- (3) customizing runtime environments and compute resources which may vary between analysis stages.



```
=====
| Include Modules
|=====
*/
```

```
include { FASTQC }
include { BWA_INDEX }
include { BWA_ALIGN }
include { SAMTOOLS_SORT; SAMTOOLS_INDEX }
include { BCFTOOLS_MPILEUP; BCFTOOLS_CALL; VCFUTILS }

from "./modules/fastqc" addParams(OUTPUT: "${params.outdir}/fastqc")
from "./modules/bwa_index" addParams(OUTPUT: "${params.outdir}/bwa_index")
from "./modules/bwa_align" addParams(OUTPUT: "${params.outdir}/bwa_align")
from "./modules/samtools" addParams(OUTPUT: "${params.outdir}/sorted_bam")
from "./modules/bcftools" addParams(OUTPUT: "${params.outdir}/vcf")
```