

Reproducible data analysis with Snakemake

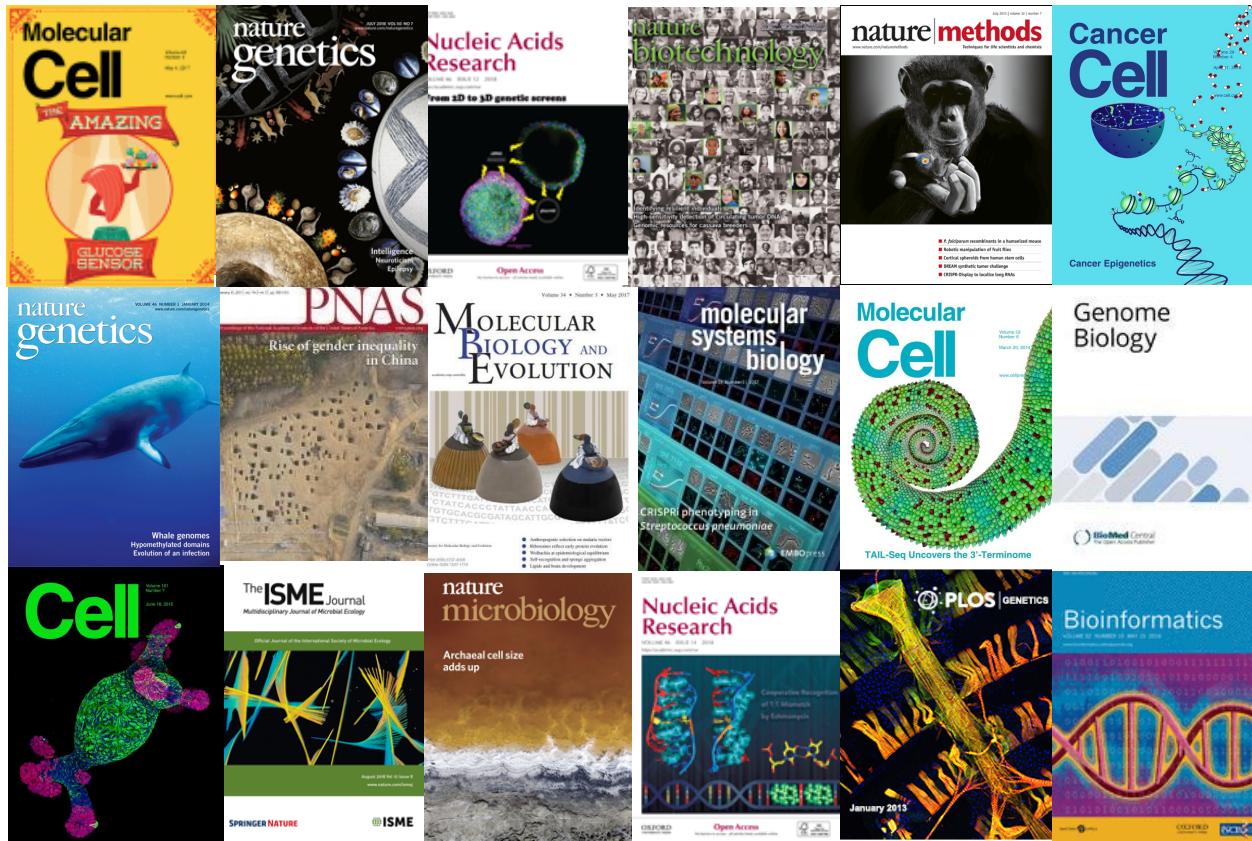
Johannes Köster
2019

<https://koesterlab.github.io>

Snakemake is popular

150k downloads since 2015

497 citations (+150 in 2019)



Concise DSL

```
rule mytask:  
    input:  
        "data/{sample}.txt"  
    output:  
        "result/{sample}.txt"  
    shell:  
        "some-tool {input} > {output}"
```

Python scripts

```
rule mytask:  
    input:  
        "data/{sample}.txt"  
    output:  
        "result/{sample}.txt"  
    script:  
        "scripts/mytask.py"
```

R scripts

```
rule mytask:  
    input:  
        "data/{sample}.txt"  
    output:  
        "result/{sample}.txt"  
    script:  
        "scripts/mytask.R"
```

Julia scripts

```
rule mytask:  
    input:  
        "data/{sample}.txt"  
    output:  
        "result/{sample}.txt"  
    script:  
        "scripts/mytask.jl"
```

No boilerplate

```
rule mytask:  
    output: "data/{sample}.txt"  
    script: "result/{sample}.txt"  
    "scripts/mytask.py"  
  
import matplotlib.pyplot as plt  
import pandas as pd  
  
d = pd.read_table(snakemake.config["hist-bins"])  
  
d.hist(bins=snakemake.config["hist-bins"])  
  
plt.savefig(snakemake.output[0])
```

<https://snakemake.readthedocs.io>

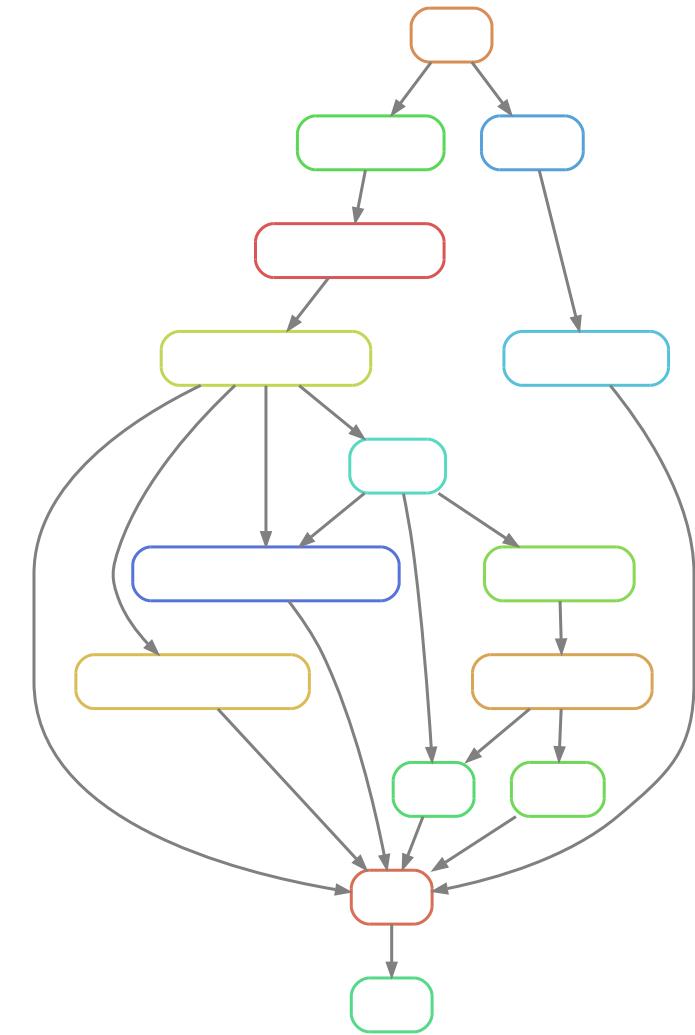
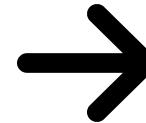
No boilerplate

```
rule mytask:  
    output: "data/{sample}.txt"  
    script: "result/{sample}.txt"  
    "scripts/mytask.py"  
  
import matplotlib.pyplot as plt  
import pandas as pd  
  
d = pd.read_table(snakemake.config["hist-bins"])  
  
d.hist(bins=snakemake.config["hist-bins"])  
  
plt.savefig(snakemake.output[0])
```

<https://snakemake.readthedocs.io>

implicit dependencies

```
rule mytask:  
    input:  
        "path/to/{dataset}.txt"  
    output:  
        "result/{dataset}.txt"  
    script:  
        "scripts/myscript.R"  
  
rule myfiltration:  
    input:  
        "result/{dataset}.txt"  
    output:  
        "result/{dataset}.filtered.txt"  
    shell:  
        "mycommand {input} > {output}"  
  
rule aggregate:  
    input:  
        "results/dataset1.filtered.txt",  
        "results/dataset2.filtered.txt"  
    output:  
        "plots/myplot.pdf"  
    script:  
        "scripts/myplot.R"
```



Scalability

workstation



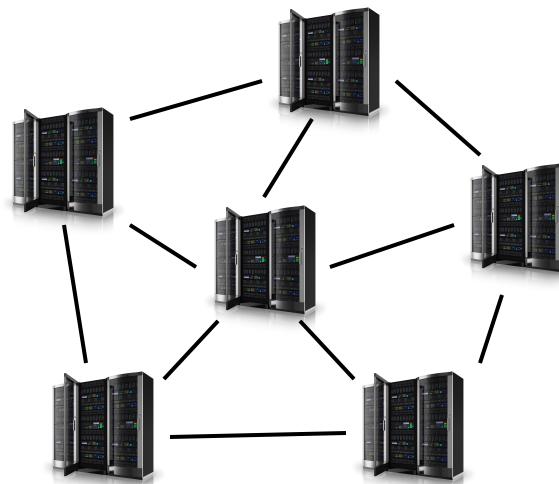
compute server



cluster



grid computing



cloud computing



kubernetes



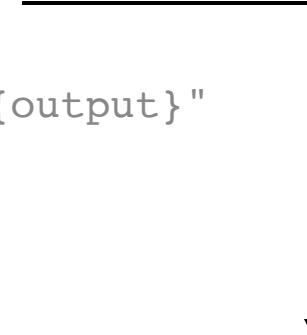
Google Cloud Platform



amazon
web services

Conda integration

```
rule mytask:  
    input:  
        "path/to/{dataset}.txt"  
    output:  
        "result/{dataset}.txt"  
    conda:  
        "envs/mycommand.yaml"  
    shell:  
        "mycommand {input} > {output}"
```



```
channels:  
    - bioconda  
    - conda-forge  
dependencies:  
    -mycommand =2.3.1
```

Singularity integration

```
rule mytask:  
    input:  
        "path/to/{dataset}.txt"  
    output:  
        "result/{dataset}.txt"  
    singularity:  
        "docker://some/container"  
    shell:  
        "mycommand {input} > {output}"
```

Singularity + Conda

```
rule mytask:  
    input:  
        "path/to/{dataset}.txt"  
    output:  
        "result/{dataset}.txt"  
    conda:  
        "envs/mycommand.yaml"  
    singularity:  
        "docker://some/os"  
    shell:  
        "mycommand {input} > {output}"
```

Snakemake

Snakemake Report

Fri Jul 19 16:58:33 2019 CET
Snakemake 5.5.3+6.g668fec20

Workflow

Statistics

Configuration

Rules

RESULTS

Allele Frequency Estimation

Concordance

FDR Control

Precision and Recall

Score Distribution

Results

Allele Frequency Estimation

Show 10 entries

Search:

File	Size	Description	Job properties	
simulated-bwa.DEL.svg	328.3 kB	Allele frequency estimation error vs. true allele frequency on simulated data.	Rule plot_allelefreq Wildcards run=simulated-bwa, vartype=DEL Params varlociraptor_callers=['delly', 'lancet', 'manta', 'strelka'], len_ranges=[[1, 30], [30, 50], [50, 100], [100, 250]]	
simulated-bwa.DEL.svg	1.9 MB	Allele frequency estimation error vs. true allele frequency on simulated data. The dashed lines depict the standard deviation, solid line depicts the mean. The grey area shows the standard	Rule plot_allelefreq_scatter Wildcards run=simulated-bwa, vartype=DEL Params depth_ranges=[[1, 20], [20, 40]], callers=['delly', 'lancet', 'manta', 'strelka']	

Snakemake

Snakemake Report	
	Fri Jul 19 16:58:33 2019 CET Snakemake 5.5.3+6.g668fec20
Workflow	
Statistics	
Configuration	
Rules	
RESULTS	
Allele Frequency Estimation	
Concordance	
FDR Control	
Precision and Recall	
Score Distribution	

Rule	Jobs	Output	Singularity	Conda environment	Code
adhoc_varlociraptor	40	+		+	+
aggregate_concordance	30	+		+	+
				python =3.6	<pre>1 from common import load_variants 2 import networkx as nx 3 import pandas as pd 4 import numpy as np 5 6 vartype = snakemake.wildcards.vartype 7 8 index_cols = ["CHROM", "POS", "SVLEN"] if var 9 10 all_variants = [load_variants(f, vartype=vart 11 12 G = nx.Graph() 13 for calls, (i, j) in zip(all_variants, snakem 14 calls["component"] = None 15 for call in calls.itertuples(): 16 a = (i, call.Index) 17 G.add_node(a) 18 if call.MATCHING >= 0: 19 b = (j, call.MATCHING) 20 G.add_node(b) 21 G.add_edge(a, b) 22 23 # get a set of calls for each dataset (we do not 24 representatives = {snakemake.params.dataset_c 25 26 if snakemake.wildcards.mode != "varlociraptor": 27 varlociraptor_variants = [load_variants(f, v 28 for calls in varlociraptor_variants: 29 calls.set_index(index_cols, inplace=True) 30 varlociraptor_representatives = {snakemake.p 31</pre>
				pandas =0.23	
				matplotlib =3.0	
				seaborn =0.9.0	
				pysam =0.13.0	
				svgutils =0.2	
				pybedtools =0.7.10	
				networkx =2.2	

scalability

