

News about Snakemake

Johannes Köster

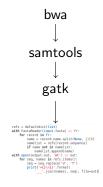
February 5, 2013

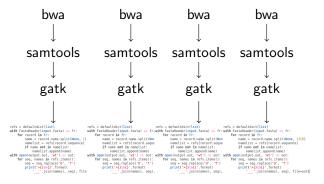
gatk

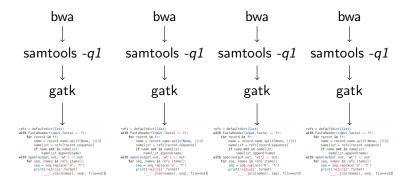
bwa

samtools

refs = defaultdict(list)
with FarithWadder(input.fait) as fr:
fr = mes = record.case.split(Nese, 1)[0]
masslist = refs!record.segence]
fr = mes = refs.record.segence
it = mes = refs.record.segence
it = mes = refs.record.segence
it = mes = refs.record.segence
for seg. nemes in refs.record.segence
print("s())wid).remail
print("s()wid).remail
print("s()wid).segence
pr







Why Snakemake?



GNU Make provided us with...

- a language to write rules to create each output file from input files
- wildcards for generalization
- implicit dependency resolution
- implicit parallelization
- fast and collaborative development on text files

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GNU Make provided us with...

- a language to write rules to create each output file from input files
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but we missed...

- easy to read syntax
- simple scripting inside the workflow
- creating more than one output file with a rule
- multiple wildcards in filenames

Structure



1 Idea

Scheduling

3 Key Features





```
rule map_reads:
   input: "hg19.fasta", "{sample}.fastq"
   output: "{sample}.sai"
   shell: "bwa aln {input} > {output}"
```



```
rule sai_to_bam:
  input: "hg19.fasta", "{sample}.sai", "{sample}.fastq"
  output: "{sample}.bam"
  shell:
    "bwa samse {input} | samtools view -Sbh - > {output}"

rule map_reads:
  input: "hg19.fasta", "{sample}.fastq"
  output: "{sample}.sai"
  shell: "bwa aln {input} > {output}"
```



```
Example: for samples \{500, \ldots, 503\} map reads to hg19.
SAMPLES = ["500", "501", "502", "503"]
rule all:
  input: expand("{sample}.bam", sample=SAMPLES)
rule sai_to_bam:
  input: "hg19.fasta", "{sample}.sai", "{sample}.fastq"
  output: "{sample}.bam"
  shell:
    "bwa samse {input} | samtools view -Sbh - > {output}"
rule map_reads:
  input: "hg19.fasta", "{sample}.fastq"
  output: "{sample}.sai"
  shell: "bwa aln {input} > {output}"
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SAMPLES = ["500", "501", "502", "503"]
rule all:
  input: expand("{sample}.bam", sample=SAMPLES)
rule sai_to_bam:
  input: "hg19.fasta", "{sample}.sai", "{sample}.fastq"
  output: protected("{sample}.bam")
  shell:
    "bwa samse {input} | samtools view -Sbh - > {output}"
rule map_reads:
  input: "hg19.fasta", "{sample}.fastq"
  output: "{sample}.sai"
  shell: "bwa aln {input} > {output}"
```



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  shell:
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rule map_reads:
  input: "hg19.fasta", "{sample}.fastq"
  output: temp("{sample}.sai")
  shell: "bwa aln {input} > {output}"
```



Example: for samples $\{500, \dots, 503\}$ map reads to hg19.

rule all 500.bam, 501.bam, 502.bam, 503.bam

```
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  shell:
    "bwa samse {input} | samtools view -Sbh - > {output}"

rule map_reads:
  input: "hg19.fasta", "{sample}.fastq"
  output: temp("{sample}.sai")
  shell: "bwa aln {input} > {output}"
```



Example: for samples $\{500, \dots, 503\}$ map reads to hg19.

```
rule all
500.bam, 501.bam, 502.bam, 503.bam
rule sai_to_bam
500.sai
```

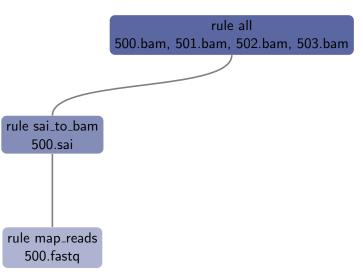
```
rule map_reads:
```

input: "hg19.fasta", "{sample}.fastq"

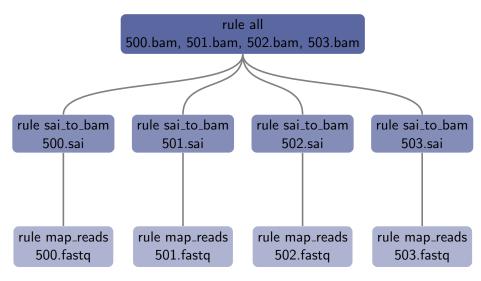
output: temp("{sample}.sai")

shell: "bwa aln {input} > {output}"

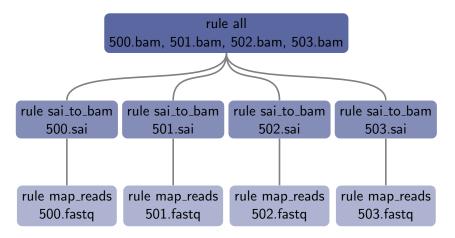












- DAG of jobs
- each path needs to be executed serially
- two disjoint paths can be executed in parallel

Building the DAG



an edge between two jobs A,B if input of A is matched by output of B, e.g.

"500.bam" matches "{sample}.bam"
$$\Leftrightarrow$$
 "500.bam" $\in L($ ".+\.bam" $)$

In case of ambiguity:

- Constrain wildcards: "{sample, [0-9]+}.bam"
- Order rules: ruleorder: sai_to_bam > sort_bam

Scheduling



execute the set of jobs E^* among all $E \subseteq J$ that maximizes under lexicographical order

$$\sum_{j\in E} (p_j,i_j)$$

such that

$$\sum_{j\in E} t_j \leq I$$

```
J set of jobs ready to execute T provided cores I idle cores I threads of job I priority of job I input size of job I
```

Python Rules



```
rule plot_coverage_histogram:
  input: ...
  output: ...
  run:
    R("""
    # some R code
    """)
```





```
rule all:
  input: dynamic("{cluster}.pdf")
rule plot:
  input: "{cluster}.csv"
  output: "{cluster}.pdf"
  shell: "gnuplot ..."
rule cluster:
  input: ...
  output: dynamic("{cluster}.csv")
  shell: "cluster ..."
```



In case of unknown number of output files... dynamically update the DAG to process these

rule all dynamic(" {cluster}.pdf')

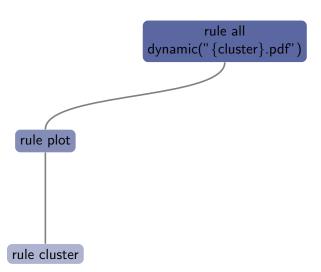
```
rule plot:
    input: "{cluster}.csv"
    output: "{cluster}.pdf"
    shell: "gnuplot ..."

rule cluster:
    input: ...
    output: dynamic("{cluster}.csv")
    shell: "cluster ..."
```

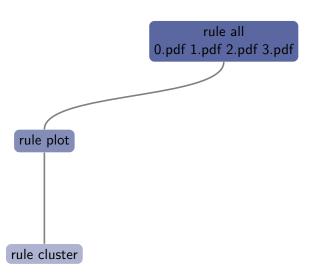


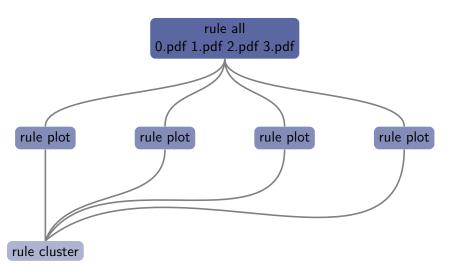
```
rule all
                   dynamic("{cluster}.pdf")
rule plot
rule cluster:
  input: ...
  output: dynamic("{cluster}.csv")
  shell: "cluster ..."
```











HTML Reports



```
rule report:
 input: "table.csv", "plot.pdf"
 output: "report.html"
 run:
  report("""
  ______
  Report of some project
  ================
  Some text containing a formula
```

:math:`\sum_j \in E t_j \leq I`
and embedding a table F1_ and a figure F2_.
""", output[0], F1=input[0], F2=input[1])

HTML Reports



Report of some project ${\bf Report\ of\ some\ project}$ Some text containing a formula $\sum_{j\in E}t_j\leq I$ and embedding a table E1 and a figure F2.





Combining expressions of IncRNAs measured by qPCR with HuEx exon arrays

HuEx and qPCR datasets and their combination

Two datasets were given: An incRNA assay using gPCR (T1) and 274 primary tumors analysed with Affymetrix HuEx exon arrays whithin the NRC.

ENSEMBL Gene IDs for IncRNAs were extracted from given IncRNA qPCR assay description (T2). For these, exonic loci were derived from the ENSEMBL hg19 v69 annotation track. Affymetrix HuEx 1.0 probes that lie within these loci were identified, and combined to meta probesets (13). Here, each row depicts an IncRNA given as ENSEMBL ID together with all the HuEx probes that should measure the expression of one of its exons.

Each of these meta-probesets summarizes the expression of one IncRNA. We calculated and normalized the expressions for the given 274 primary tumors. This was done with the Affymetrix Power Tools implementation of RMA with default parameters. It remains to be investigated if the RMA normalization has successful removed batch effects since the tumor data comes from different labs.

Estimation of regulated IncRNAs in the gPCR dataset

We estimate the consistency between the two controls by calculating the fold-change and throwing away all incRNAs that exceed a threshold in this test. For the remaining lncRNAs the fold-change between treatment and the mean of the two

Table T4 shows upregulated incRNAs sorted by strength of fold-change, Table T5 shows the smae for downregulated IncRNAs. Figure [1] shows the histogram of fold-changes

Counting tumors expressing the regulated IncRNAs in the HuEx dataset

We assume incRNAs with an absolute logarithmic fold-change greater than 0.69 to be regulated.

We only consider those incRNAs that can be measured by exonarray probes (see Table 16). For these we calculate from the HuEx setup described above the number of tumors with a minimum probeset expression of 6. Table T shows the results, Figure P2 shows a histogram of the observed counts.

Mathematical background

The provided qPCR analysis yielded -cq values that are on a logarithmic scale compared to the real molecule counts. This is because each PCR cycle in theory doubles the amount of molecules. Since this rate is not reached in practice we assume a factor of 1.8 here.

Consequently a fold-change on these logscaled values has to be computed as subtraction instead of a quotient. Further, a non-logarithmic foldchange of t corresponds to logs at in logarithmic scale. It have to be noted that at the moment it is not yet clear whether the normalization applied before may have an effect on the assumed factor.

We denote the different -cq values as $cq^i_{r'}$, $cq^i_{r'}$ and $cq^i_{r+r'}$ for a given incRNA i. The consistency between the two controls (see above) is now computed as

$$|cq_{0ctrl}^i - cq_{nt}^i| > 0.5$$

The fold-change between treatment and the mean of controls is calculated as

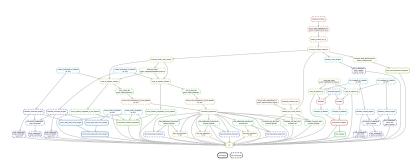
$$cq_{si}^i - \frac{1}{2} \left(cq_{0ctrl}^i + cq_{nt}^i \right)$$

Since the latter -cq values are still logarithmic, the mean here corresponds to the geometric mean of the real molecule counts. This is intended since it avoids domination of the mean by the higher -cg value.

2013-02-01

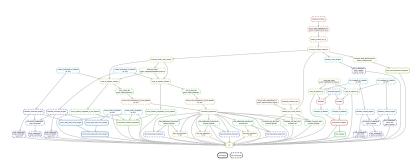
Using the dot language of graphviz:

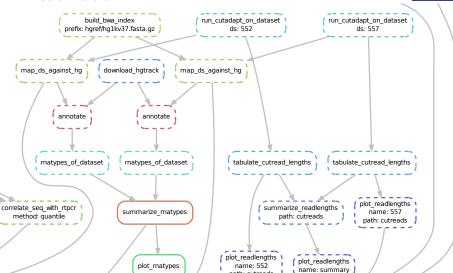
\$ snakemake --dag | dot | display



Using the dot language of graphviz:

\$ snakemake --dag | dot | display





path: cutreads

Conclusion



Snakemake is a new workflow system that provides:

- an easy pythonic textual representation
- multiple wildcards in filenames
- dynamic update of job DAG
- implicit parallelization and dependency resolution
- job scheduling considering threads, priorities and input size
- cluster and batch support

https://bitbucket.org/johanneskoester/snakemake $\mbox{depends on Python} \geq 3.2$

Early adopters: Shirley Liu's lab, Broad Institute Genome of the Netherlands Project