

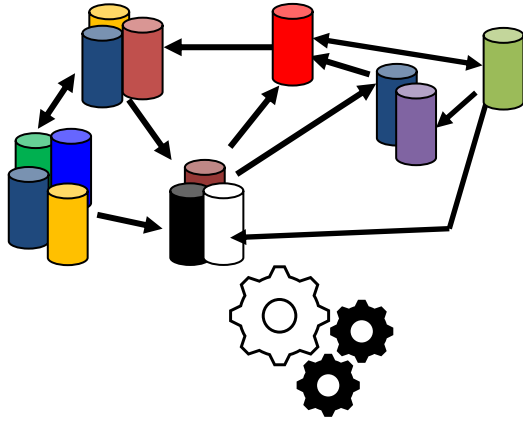
10 ans après la crise de la reproductibilité en bioinformatique : de la reproduction à la réutilisation de workflows scientifiques



Sarah Cohen-Boulakia November, 2025

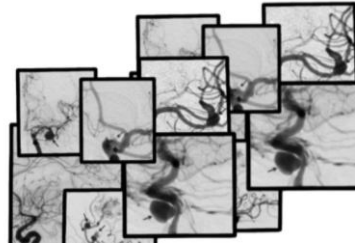
Laboratoire Interdisciplinaire des Sciences du Numérique – LISN
Université Paris-Saclay

Biological and Biomedical data analysis



Public sources

Distributed
Heterogeneous Network
> 1,500 (NAR)



Tools – Scripts

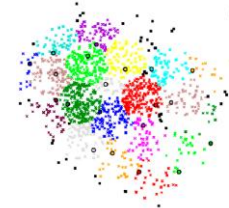
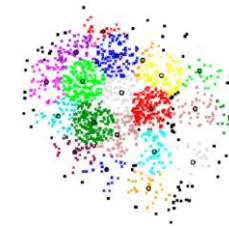
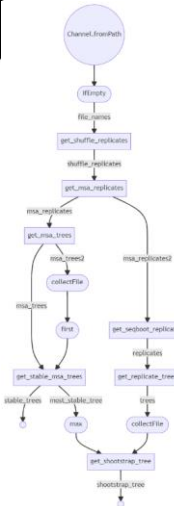
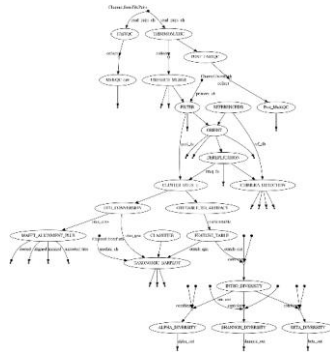
Distributed
Heterogeneous
> 23,000 (bio.tools)



**Which exact dataset
did I use? Which tool
version? Which
parameters...?**

Analysis pipelines

Combining multiple tools
Heterogeneous
Various environnements
& platforms
> 1,000 workflows
(GitHub)



Studies on reproducibility - Reproducibility Crisis...

Nekrutenko & Taylor - [Nature Genetics \(2012\)](#)

[50 papers](#) 2011 using the Burrows-Wheeler Aligner

31/50 ([62%](#)) provide [no information](#)

no version of the tool + no parameters + no genomic ref
sequence

7/50 ([14%](#)) provide all the necessary details

Alsheikh-Ali et al, [PLoS one \(2011\)](#)

10 papers in the top-50 IF journals → [500 papers](#)

149 (30%) were [not subject to any data availability policy](#)
(0% data available)

Of the remaining 351 papers

208 papers (59%) did [not adhere](#) to the data availability instructions

143 make a statement of [willingness to share](#)

47 papers ([9%](#)) deposited full primary raw data online

→ Computational reproducibility

Studies on reproducibility - Reproducibility Crisis...

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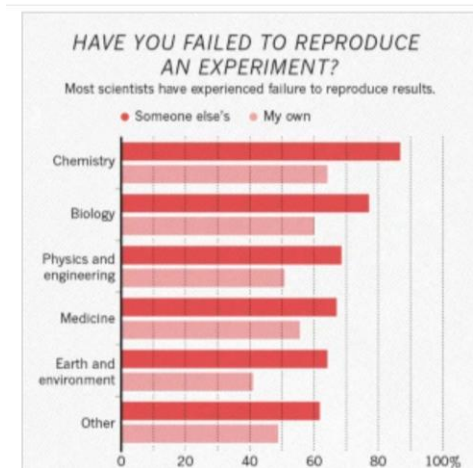
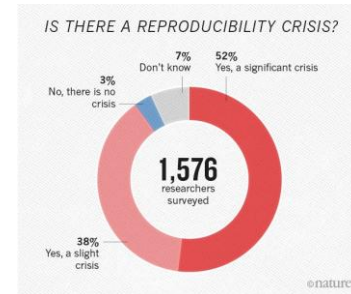
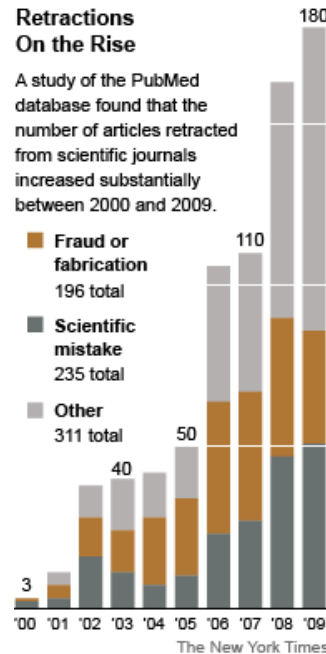
47 papers (9%) deposited full primary raw data online

→ Computational reproducibility



Retractions On the Rise

A study of the PubMed database found that the number of articles retracted from scientific journals increased substantially between 2000 and 2009.



Current solutions to computational reproducibility

Scientific Workflow Management Systems

“Data analysis pipeline ”

Data flow driven

WF specification

connected *processors*

steps of the analysis

WF execution

data consumed/produced

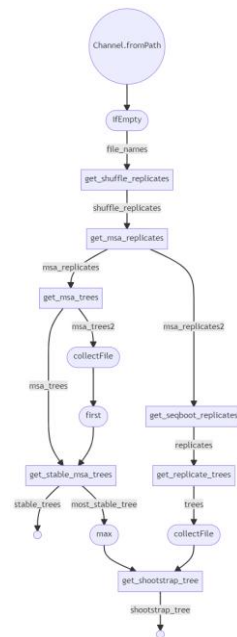
Provenance modules

Scheduling ...

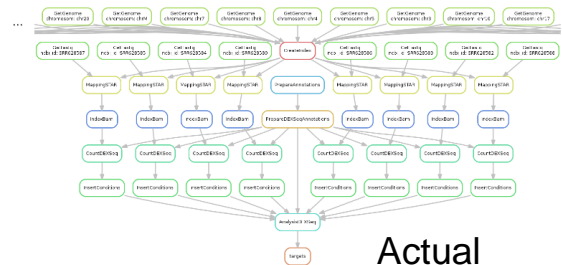


nextflow

Specification Steps



Executed Steps



Possible dataflow

Actual
dataflow

Current solutions to computational reproducibility

Scientific Workflow Management Systems

“Data analysis pipeline ”

Data flow driven

WF specification

connected *processors*

steps of the analysis

WF execution

data consumed/produced

Provenance modules

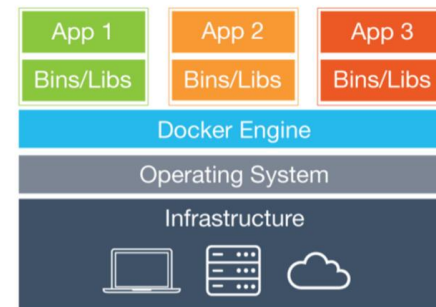
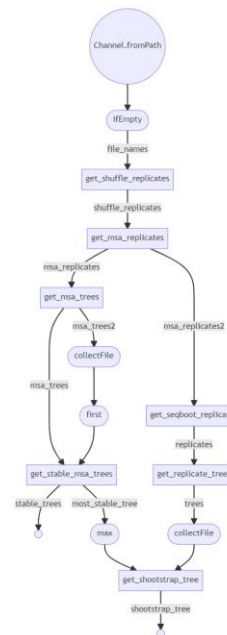
Scheduling ...

WF environment

everything needed to run Libraries, dependencies...

Coupling WF with Docker/AppTainer

BioContainer



Levels of computational reproducibility

From identical output to the same scientific results

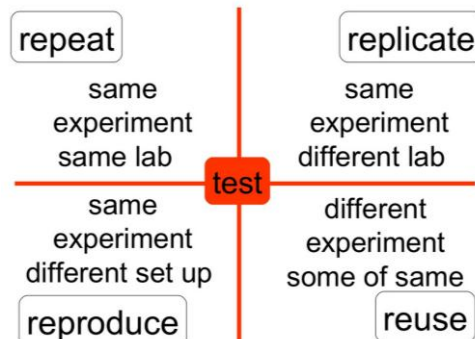
Repeat

Redo - exact same context

Same workflow, execution setting, environment

Identical *output*

Aim = proof for reviewers 😊



Drummond C Replicability is not Reproducibility: Nor is it Good Science, online
Peng RD. Reproducible Research in Computational Science Science 2 Dec 2011: 1226-1227.

Replicate

Variation allowed in the workflows
execution setting environment

Similar *output*

Aim = robustness

Levels of computational reproducibility

From identical output to the same scientific results

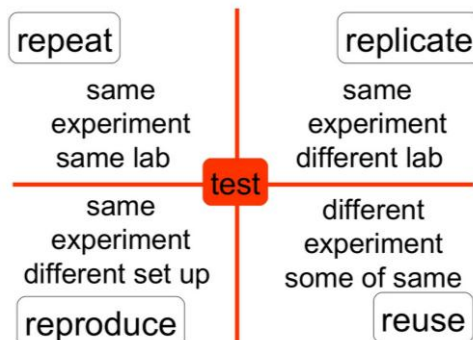
Repeat

Redo - exact same context

Same workflow, execution setting, environment

Identical *output*

Aim = proof for reviewers 😊



Drummond C Replicability is not Reproducibility: Nor is it Good Science, online
Peng RD. Reproducible Research in Computational Science Science 2 Dec 2011: 1226-1227.

Reproduce

Same *scientific result*

But the means used may be changed

Different workflows, execution setting, environment

Different output but in accordance with the result

Replicate

Variation allowed in the workflows
execution setting environment

Similar *output*

Aim = robustness

Reuse

Adapt to new needs

Possibly different scientific result

Reuse in part existing workflows

→ **Cumulative science**

→ **No Reuse without Repeat!**

Reproducibility Networks



French Reproducibility
Network born in 2023
270+ colleagues

Next Scientific Days in Lyon
April 3-4th

<https://www.recherche-reproductible.fr/index-en>


MINISTÈRE
DE L'ENSEIGNEMENT
SUPÉRIEUR
ET DE LA RECHERCHE
*Liberté
Égalité
Fraternité*

Global Networks

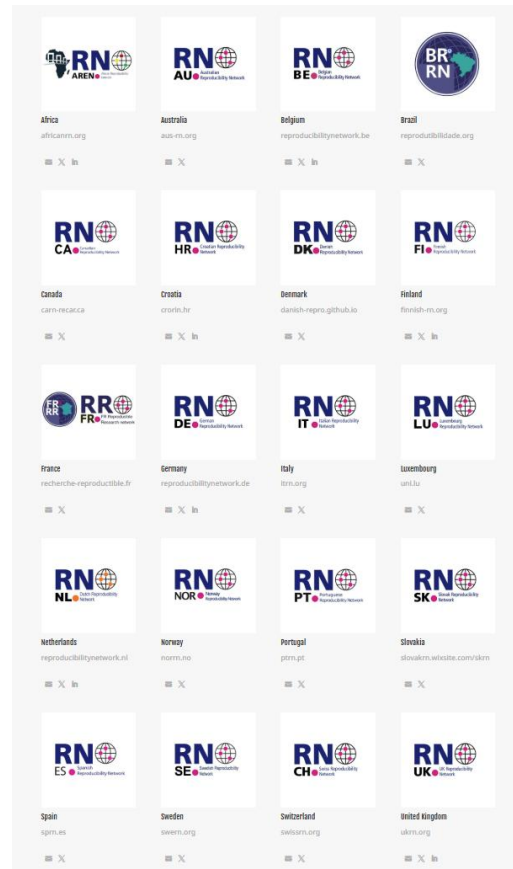
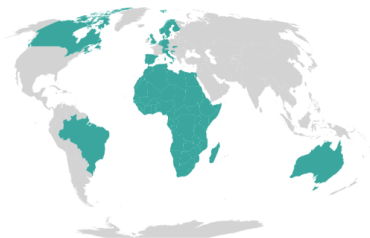
Outside the UK? Find a Reproducibility Network in your area

[See full Global Networks Statement](#)

Global Reproducibility Networks

A Reproducibility Network (RN) is a national, peer-led consortium of researchers that aims to promote and ensure rigorous research practices by establishing appropriate training activities, designing and evaluating research improvement efforts, disseminating best practice and working with stakeholders to coordinate efforts across the sector. RNs aim for broad disciplinary representation and an intensive interdisciplinary dialogue (e.g., with funding agencies, publishers, learned societies and other sectoral organisations, as well as researchers from all disciplines and across all career stages).

To reach as many researchers as possible, and to operate as efficiently as possible, we are keen to support other countries interested in creating similar networks. If you are interested in setting up a national RN, or finding out who in your country is working towards this, please email: contact@ukrn.org.



Outline

The reproducibility landscape

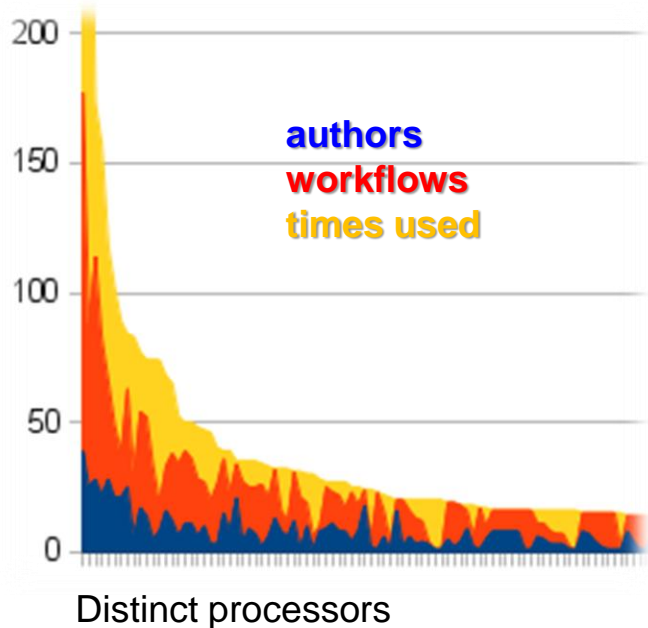
Status of workflow reuse

How to improve workflow reuse?

ZOOM on 2 current contributions

Conclusion

Reuse status - 2010



1,700 Taverna workflows
10 242 processors (analysis steps)
Centralized in myExperiment



Re-use rates have a Zipf-like distribution

The top ten authors published 62% of all workflows

Reuse status - 2023

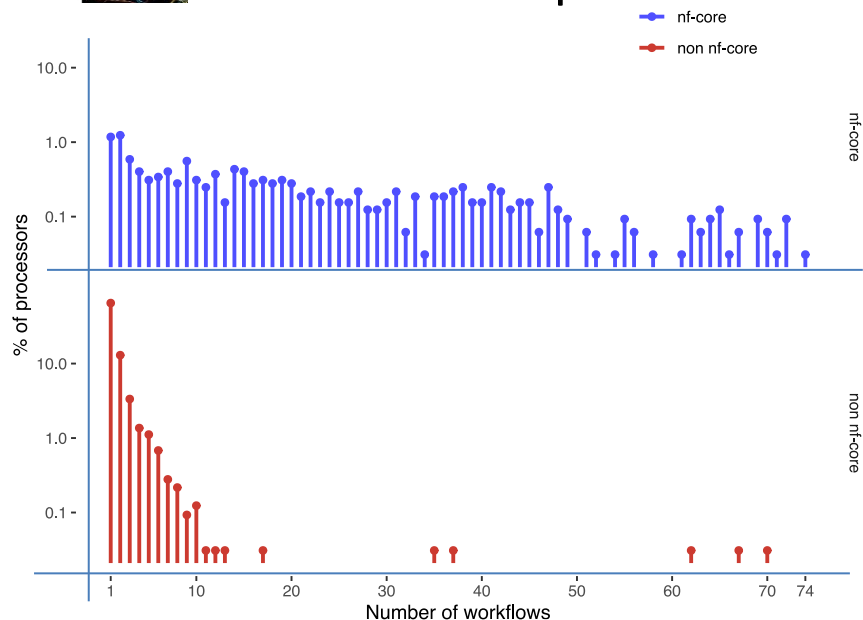


2,443 workflows Nextflow & Snakemake
15 540 processors (analysis steps)

Distributed in github


nf-core: repro of carefully checked Nextflow processors

Marine Djaffardjy



Still low reuse

The top ten authors published only 15% of all workflows

Higher reuse (black box) of processors in **nf-core** 

Outline

The reproducibility landscape

Status of workflow reuse

How to improve workflow reuse?

ZOOM on 2 current contributions

Conclusion

ShareFAIR

ShareFAIR : Sharing reliable workflows to transform datasets into gold standards: Application to Neuro-Vascular Pathologies

PEPR Santé Numérique

Coordinator

- Sarah Cohen-Boulakia

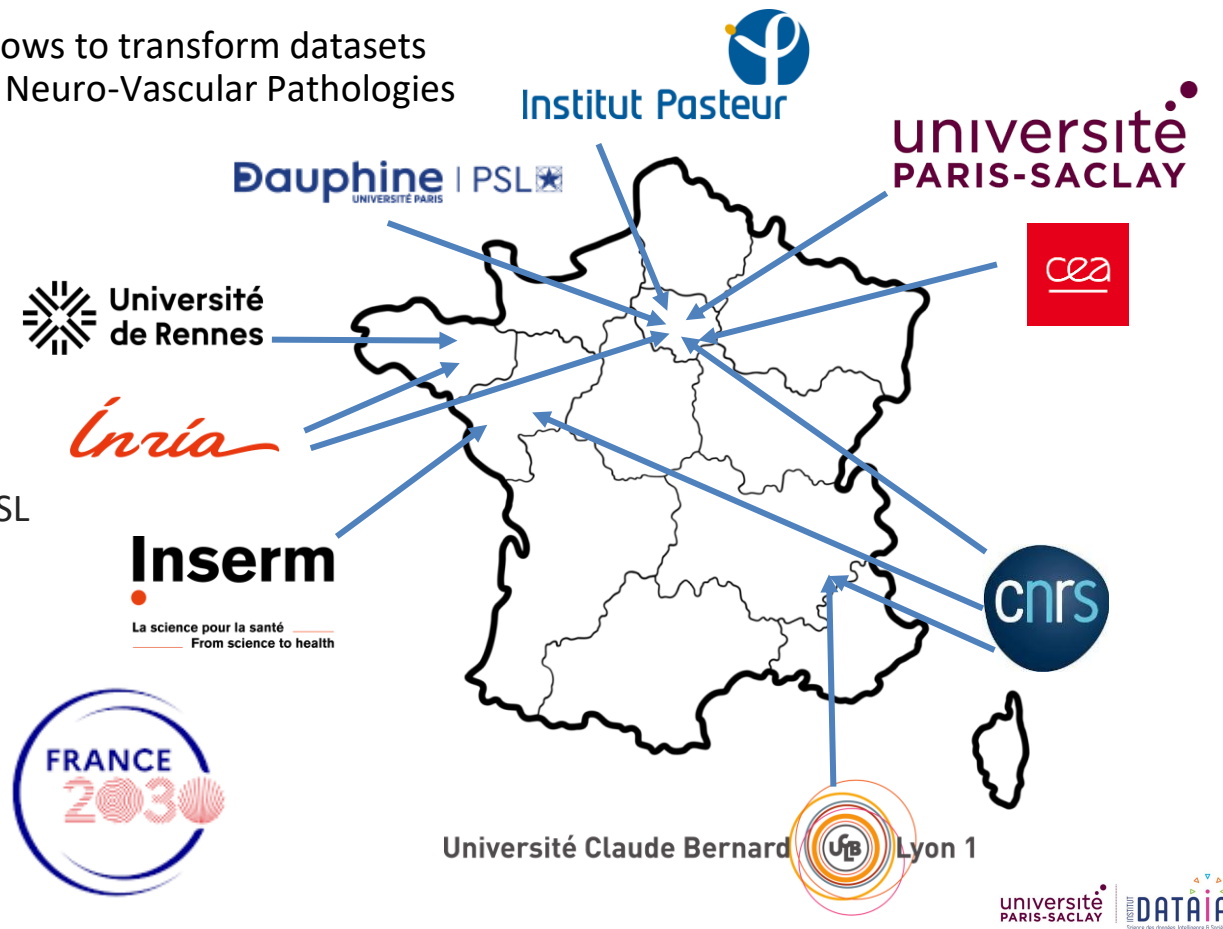
Management structure

- Université Paris-Saclay

Partners

- Université Paris Dauphine PSL
- Institut Pasteur
- Université Lyon
- Université Rennes
- Inria
- CEA
- CNRS, INSERM (ITX)

Duration: 48 mois



Objectives (1/3)

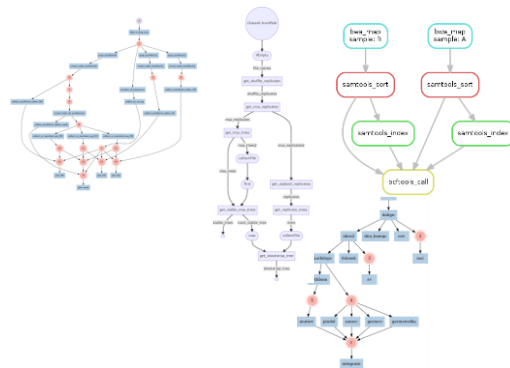
Define **an interoperable framework** for the design, annotation, and sharing of **reproducible and reusable workflows**

Design a **language** to query workflows, and their executions

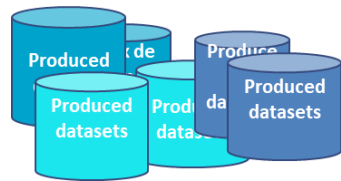
Capture provenance (executions) in an optimized manner

Develop a **FAIRification method** for datasets

Guide workflow developers in **discovering and comparing** workflows



Execution traces:
provenance of
produced data



Objectives (2/3)

Complete standards for uniformly annotating workflows in terms of analysis tools and input/output datasets to produce **FAIR datasets**



Identification of a set of existing standards for annotating protocols, workflows, and datasets: EDAM, EDAM-Biolmaging, MONDO, DUO, etc.

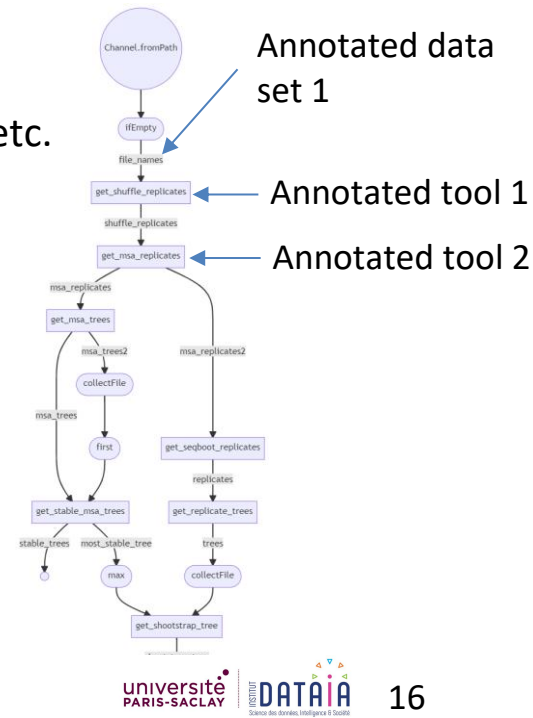


Identification of overlaps, varying levels of precision, and missing concepts



Construction of annotation guides

Development of a Knowledge base



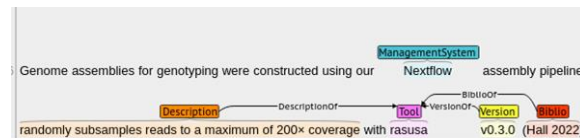
Objectives (3/3)



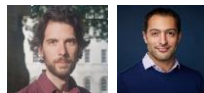
Augment the set of workflows by extracting them from text (litterature) and large datasets from communities



Develop NLP models to extract workflow description from the scientific literature



Learn protocols form clinical data collected all along the care activity (patient data). Comparison between learnt and declared protocols.



Extract protocols and workflows from large shared datasets in neuroimagergy

nature

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Article | [Published: 20 May 2020](#)

Variability in the analysis of a single neuroimaging dataset by many teams

Outline

The reproducibility landscape

Status of workflow reuse

How to improve workflow reuse

ZOOM on 2 current contributions

Conclusion

Exploiting the 3 forms of the workflow

```
#####
# PREPARE ANNOTATION FOR LATER INPUT AND USAGE
#####
process format_annotation {
    label "python3"
    if (workflow.profile.contains('node')) { label 'smallTask' }

    if { params.softlink_results } { publishDir "${params.output}/${params.annotation_dir}", pattern: "**.id2details*" }
    else { publishDir "${params.output}/${params.annotation_dir}", mode: 'copy', pattern: "**.id2details*" }

    input:
    path(annotation)
    val(gtf_str, type)
    val(gtf_feature_type_of_attr_type)

    output:
    path("${annotation.basename}.id2details")

    shell:
    ...
    #!/usr/bin/env python3
    import sys
    target_id_set = set()

    with open("${annotation}", 'r') as gtf, open("${annotation.basename}.id2details", 'w') as out:
        for line in gtf:
            if not line.startswith('#'):
                split_line = line.split('\t')
                if split_line[2] == '${gtf_feature_type_of_attr_type}' or split_line[2] == "pseudogene":
                    if '${gtf_attr_type}' not in line:
                        sys.exit("ERROR: No '${gtf_attr_type}' found in a '${gtf_feature_type_of_attr_type}'-type line\n${line}\nCheck your an
target_id = line.split('${gtf_attr_type}')[1].split(';')[0].replace(';', '').strip()
                    if target_id not in target_id_set:
                        desc = split_line[3]
                        chr = split_line[4]
                        start = split_line[5]
                        stop = split_line[6]
                        strand = split_line[7]
                        if '${gtf_attr_type}' == "transcript_id" and "transcript_name" in desc:
                            target_name = desc.split('transcript_name')[1].split(';')[0].replace(';', '').strip()
                        else:
                            if 'gene_name' in desc:
                                target_name = desc.split('gene_name')[1].split(';')[0].replace(';', '').strip()
                            else:
                                target_name = target_id
                        if target_name == 'NA':
                            target_name = target_id
                        if '${gtf_attr_type}' == "transcript_id" and "transcript_biotype" in desc:
                            target_biotype = desc.split('transcript_biotype')[1].split(';')[0].replace(';', '').strip()
                        else:
                            if 'gene_biotype' in desc:
                                target_biotype = desc.split('gene_biotype')[1].split(';')[0].replace(';', '').strip()
                            else:
                                target_biotype = 'NA'
                        target_id_set.add(target_id)
                        out.write('\t\t'.join([target_id, target_name, target_biotype, chr, start, stop, strand, desc.rstrip()]) + '\n')
    ...
}
```

Code: [GitHub](#) - data science

Code: [GitHub - data science](#)

```
#####
# PREPARE ANNOTATION FOR LATER UPDATE AND USAGE
#####
process format_annotation {
    label python'

    if (!workflow.profile.contains("node")) { label "smalltask" }

    if ( ! param.softlink_results ) { publicOutDir["$param.output"]/$param.annotation_dir/, pattern: "", idetails" }
    else { publicOutDir["$param.output"]/$param.annotation_dir/", mode: "copy", pattern: "".$idetails" }

    input:
        path(annotation)
    val(gtf_attr_type)
    val(gtf_attr_type)
    val(gtf_feature_type_of_attr_type)

    output:
        path("$annotation.baseName".$idetails")

    shell:
    ...
    #!/usr/bin/env python
    import sys
    target_id_set = set()
    with open("{}(annotation)", "r") as gtf, open("{}(annotation.baseName".$idetails)".a") as desc:
        for line in gtf:
            if not line.startswith("#"):
                split_line = line.split("\t")
                if split_line[0] == "(gtf.feature_type_of_attr_type)" or split_line[2] == "transcript":
                    if ("(gtf_attr_type)" not in line:
                        sys.exit("ERROR: No '(gtf_attr_type)' found in a '(gtf.feature_type_of_attr_type)-type line({})".format(line))
                    check you are
                    target_id = line.split("(gtf_attr_type")[1].split(":")[0].replace("", "").strip()
                    if target_id not in target_id_set:
                        desc = split_line[5]
                        chr = split_line[8]
                        start = split_line[9]
                        stop = split_line[10]
                        strand = split_line[16]
                        if "(gtf_attr_type)" == "transcript_id" and "transcript_name" in desc:
                            target_name = desc.split("transcript_name")[1].split(":")[0].replace("", "").strip()
                        else:
                            if "gene_name" in desc:
                                target_name = desc.split("gene_name")[1].split(":")[0].replace("", "").strip()
                            else:
                                target_name = split_id
                            if target_name == "NA":
                                target_name = target_id
                            if "(gtf_attr_type)" == "transcript_id" and "transcript_biotype" in desc:
                                target_biotype = desc.split("transcript_biotype")[1].split(":")[0].replace("", "").strip()
                            else:
                                if "gene_biotype" in desc:
                                    target_biotype = desc.split("gene_biotype")[1].split(":")[0].replace("", "").strip()
                                else:
                                    target_biotype = "NA"
                            target_id_set.add(target_id)
                            out.write("{}\t{}\t{}(target_id), target_name, target_biotype, chr, start, stop, strand, desc.split()) == "\n")

```

Open Access Article

RNAflow: An Effective and Simple RNA-Seq Differential Gene Expression Pipeline Using Nextflow

by Marie Lataretu ¹ and Martin Holzer ^{2,*}

- ² Methodology and Research Infrastructure, MF1 Bioinformatics, Robert Koch Institute, Nordufer 20, 13353 Berlin, Germany

Genes 2020, 11(12), 1487; <https://doi.org/10.3390/genes11121487>

Submission received: 25 October 2020 / Revised: 4 December 2020 / Accepted: 7 December 2020 /
Published: 10 December 2020

(This article belongs to the Special Issue Algorithms and Workflows in RNA Bioinformatics)

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Abstract

RNA-Seq enables the identification and quantification of RNA molecules, often with the aim of detecting differentially expressed genes (DEGs). Although RNA-Seq evolved into a standard technique, there is no universal gold standard for these data's computational analysis. On top of that, previous studies proved the irreproducibility of RNA-Seq studies. Here, we present a portable, scalable, and parallelizable Nextflow RNA-Seq pipeline to detect DEGs, which assures a high level of reproducibility. The pipeline automatically takes care of common pitfalls, such as ribosomal RNA removal and low abundance gene filtering. Apart from various visualizations for the DEG results, we incorporated downstream pathway analysis for two model species as *Homio sapiens* and *Mus musculus*. We evaluated the pipeline using a simulated dataset and a real dataset of rRT-PCR data serving as a reference and observed a very high correlation of the localized transcriptome expression fold changes.

Keywords: RNA-Seq; workflow; Nextflow pipeline; differential gene expression analysis

1. Introduction

More than a decade ago possibility to sequence transcripts (RNA-Seq of any given species, starting from small bacteria, such as *Helicobacter pylori* [1], opened up completely new ways to analyze gene expression and obtain never before possible insights into regulatory mechanisms on an incredibly large scale. Starting from short, single-end reads and few samples per study, we now have access to continuously growing and large transcriptomic data sets of high sequencing quality [2]. A standard and widely distributed computational task comprises the processing of RNA-Seq data to identify differentially expressed features, mainly genes (DEGs), between varying conditions. This task is computationally demanding, as it requires the processing of large data sets. Nowadays, setting up the necessary computational steps to perform an RNA-Seq-based DEG study is still time-consuming and prone to errors, which ultimately leads to results that are difficult to reproduce [3].

Text: scientific papers

Text: scientific papers

More than a decade ago, the possibility to sequence a transcriptome (RNA-Seq) of any given species, starting from small bacteria, such as *Helicobacter pylori* [1], opened up completely new ways to analyze gene expression and obtain never before possible insights into regulatory mechanisms on an incredibly large scale. Starting from short, single-end reads and few samples per study, we now have access to continuously growing and large transcriptomic data sets of high sequencing quality [2]. A standard and widely distributed computational task comprises the processing of RNA-Seq data to identify differentially expressed features, mainly genes (DEGs), between varying conditions, such as an infection or a treatment. However, the increasing size of transcriptomic data nowadays, setting up the necessary computational steps to perform an RNA-Seq based DEG study is still time-consuming and prone to errors, which ultimately leads to results that are difficult to reproduce [3].

A diagram illustrating the interaction between a human and a network. On the left, a stylized human figure is shown in profile, reaching out towards a globe on the right. The globe contains a network of interconnected nodes and lines, representing a complex system. Below the globe and human figure is a large, blue, double-headed horizontal arrow, indicating a bidirectional relationship or interaction between the human and the network.

[illegible]

More than a decade ago possibility to sequence transcripts (RNA-Seq) of any given species, starting from small bacteria, such as *Helicobacter pylori* [1], opened up completely new ways to analyze gene expression and obtain never before possible insights into regulatory mechanisms on an incredibly large scale. Starting from short, single-end reads and few samples per species, we now have access to continuously growing and large transcriptomic data sets of high sequencing quality [2]. A standard and widely distributed computational task comprises the processing of RNA-Seq data to identify differentially expressed features, mainly genes (DEGs), between varying conditions, such as as an example, between different cell types or different developmental stages. Unfortunately, nowadays, setting up the necessary computational steps to perform an RNA-Seq-based DEG study is still time-consuming and prone to errors, which ultimately leads to results that are difficult to reproduce [3].

Text: scientific papers

Aim

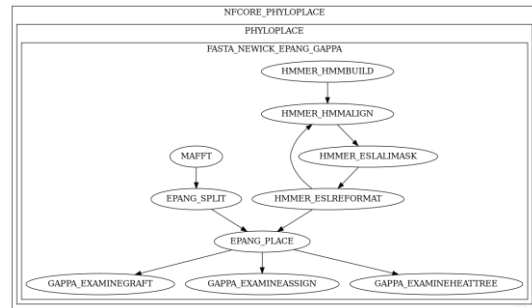
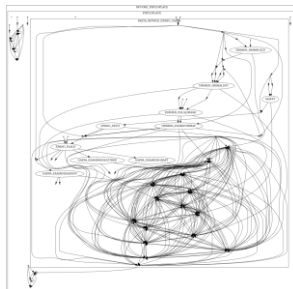
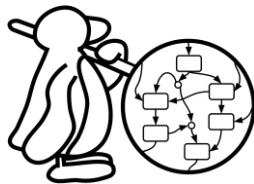
Provide an overview of the main step of a workflow

Coherent with the code...

 nextflow



Workflow Code



George
Marchment



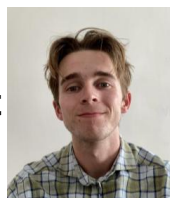
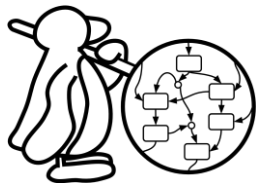
Aim

Provide an overview of the main step of a workflow

Coherent with the code...



Workflow Code



George
Marchment

Functionalities

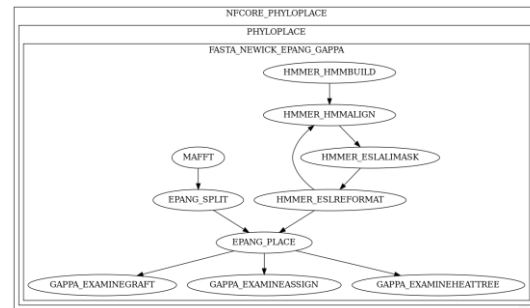
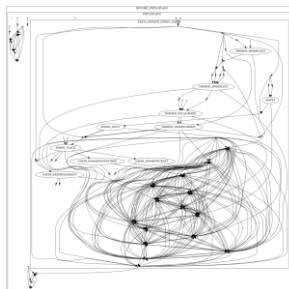
Analyses the code of Nextflow workflows

Generates visual graphs at different **simplification** levels depicting the workflow's structure

Detects errors in the workflow code

Open source: command line or web service

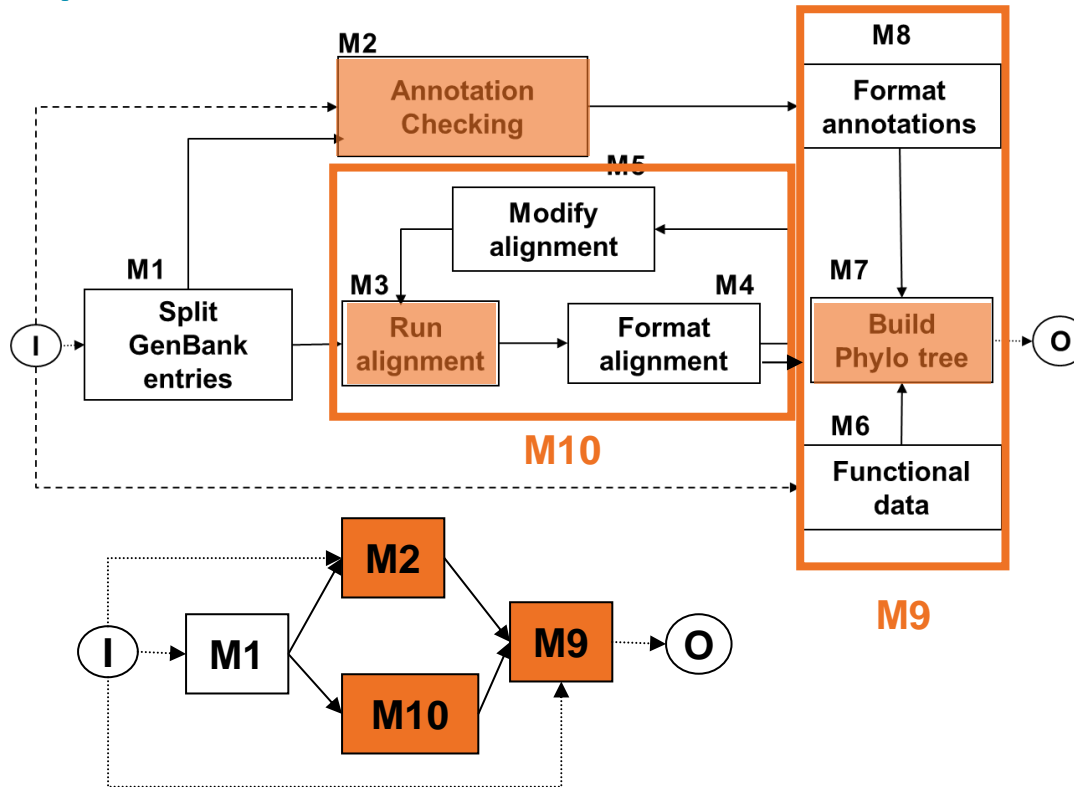
<https://bioflow-insight.pasteur.cloud/>



Simplifying the workflow graph

Choose **relevant modules**

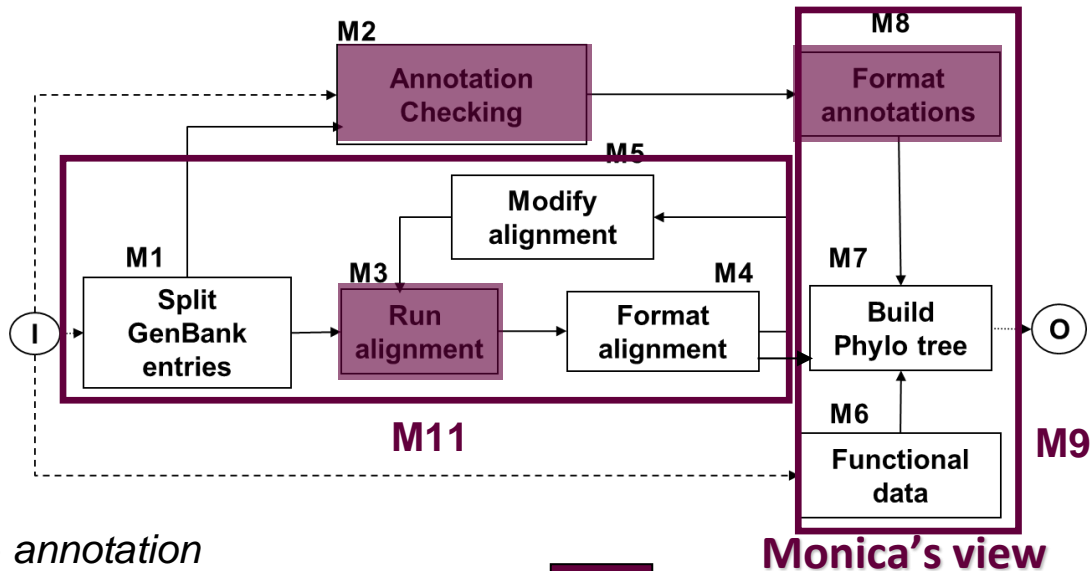
Each **composite** module is constructed around one **relevant module**



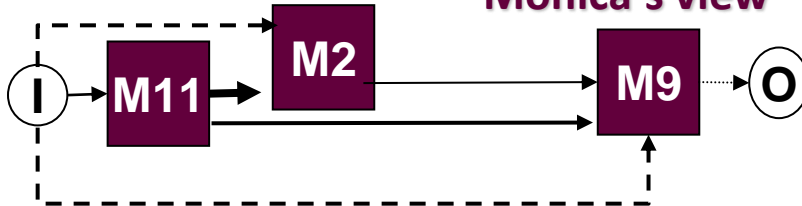
The composite module takes the meaning of the relevant module it contains

Grouping may be error-prone!

Grouping should **preserve the relationships** between **relevant modules**



The *annotation checking* module does not need input from the *run alignment* module!



ZOOM*UserViews in NextFlow Workflows

Formalization of the set of properties to be preserved

Property 1: Given G_w and $R \subseteq N$ relevant modules, U is **well-formed** iff every composite module in U contains at most one element of R .

Property 2: A user view U **preserves dataflow** iff every edge in G_w that induces an edge on an nr -path from $C(r)$ to $C(r')$ in $U(G_w)$ lies on an nr -path r to r' in G_w .

Property 3: A user view U **is complete w.r.t dataflow** iff for every edge e on an nr -path from r to r' in G_w that induces an edge e' in $U(G_w)$, e' lies on an nr -path from $C(r)$ to $C(r')$.

Theorem *ZOOM* is a **polynomial-time** which *preserves Properties 1- 3 and produces a minimal user view*

Implementation of ZOOM

New! Rewriting the workflow code to implement user views

Text: scientific papers

```

graph TD
    In1(( )) --> PGI[PREPARE_STAR_GENOME_INDEX]
    In2(( )) --> PGP[PREPARE_GENOME_PICARD]
    In3(( )) --> PGS[PREPARE_GENOME_SAMTOOLS]
    In4(( )) --> PVP[PREPARE_VCF_FILE]
    In5(( )) --> RSG[RNASEQ_GATK_RECALIBRATE]
    In6(( )) --> RSC[RNASEQ_CALL_VARIANTS]
    In7(( )) --> PPS[POST_PROCESS_VCF]
    In8(( )) --> PVF[PREPARE_VCF_FOR_ASE]
    In9(( )) --> AK[ASE_KNOWNSNPS]

    PGI --> PGP
    PGI --> PGS
    PGI --> RSG
    PGP --> PVP
    PGP --> RSG
    PGS --> RSG
    PGS --> RSC
    PGS --> PPS
    PGS --> PVF
    PGS --> AK
    PVP --> RSG
    PVP --> RSC
    PVP --> PPS
    PVP --> PVF
    PVP --> AK
    RSG --> RSC
    RSG --> PPS
    RSG --> PVF
    RSG --> AK
    RSC --> PPS
    RSC --> PVF
    PPS --> PVF
    PVF --> AK
    
```



1. Introduction

More than a decade ago, the possibility to sequence transcripts (RNA-Seq) of any given species, starting from small bacteria, such as *Helicobacter pylori* [1], opened up completely new ways to analyze gene expression and obtain never before possible insights into regulatory mechanisms on an incredibly large scale. Starting from short, single-end reads and few samples per study, we now have access to continuously growing and large transcriptomic data sets of high sequencing quality [2]. A standard and widely distributed computational task comprises the processing of RNA-Seq data to identify differentially expressed features, mainly genes (DEGs), between varying conditions, such as an infected cell versus a non-infected cell. In the last few years, numerous performance-oriented tools have been developed, setting up the necessary computational steps to perform an RNA-Seq-based DEG study is still time-consuming and prone to errors, which ultimately leads to results that are difficult to reproduce [3].

Entity extraction from scientific papers

Low-resource extraction task

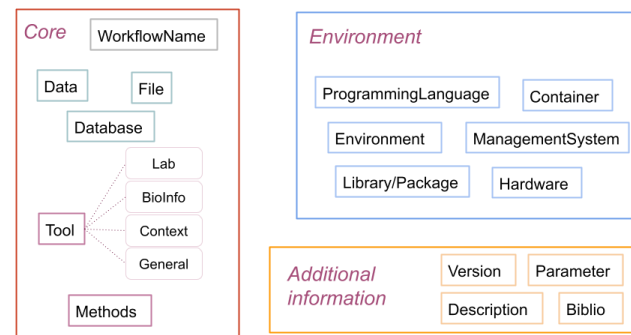
Bioinformatics workflows underrepresented in NLP

No Corpus available

Several strategies tested

Schema of entities representing
key entities related to workflow

Annotated corpus of 52 full papers describing
Snakemake and Nextflow workflows
7 annotators - Inter annot. Agreement (0.70)

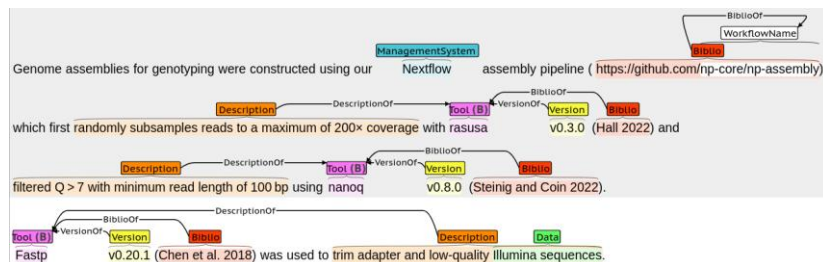


Neuronal approach biLSTM-CRF
Nlstruct python lib
SciBert & BioBert

F1-score ~0.70 all entities
(0.77 on Tools)

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IDA
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Clémence Sebe

Wrap up - Next steps

Lack of reproducibility hurts **cumulative science**

We are still living the **reproducibility crisis**

Several technical solutions exist to help redo/reexecute

Challenges lie at **the reuse level**: repurposing workflow analyses, adapting to own needs

ShareFAIR ambitions to provide a **proof-of-concept of workflow sharing** by providing a **reuse platform**

Current work on

Coupling **workflow code & workflow papers** (NLP-code)

Abstracting workflow **graph structures** (code-graph)



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Lyon

Thanks!

