

# Reproducibility

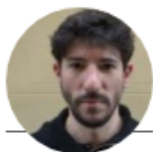


# **REPRODUCIBILITY: WHY DOES IT MATTER?**

PESQUISAS CIENTÍFICAS ›

# Ciência vive uma epidemia de estudos inúteis

Cientistas dos EUA, Reino Unido e Holanda denunciam que a pesquisa está perdendo parte de sua credibilidade



NUÑO DOMÍNGUEZ



19 JAN 2017 - 16:09 BRST

Há séculos, não bastava a Newton e Galileu realizarem descobrimentos capazes de mudar a história. Deveriam também repetir suas experiências diante de todos os seus colegas, e esses, por sua vez, as repetiam por sua conta antes de ficarem completamente convencidos. Esse princípio de reprodutibilidade foi fundamental para o avanço da [ciência](#) desde então. Na atualidade, essa garantia essencial está se perdendo, e coloca em dúvida a validade de muitos estudos em quase todas as disciplinas.

[nature](#) > [news](#) > article

NEWS | 25 April 2025

# Huge reproducibility project fails to validate dozens of biomedical studies

**Unique reproducibility effort in Brazil focuses on common methods rather than a single field — and prompts call for reform.**

By [Rodrigo de Oliveira Andrade](#)



Fonte: <https://www.nature.com/articles/d41586-025-01266-x>



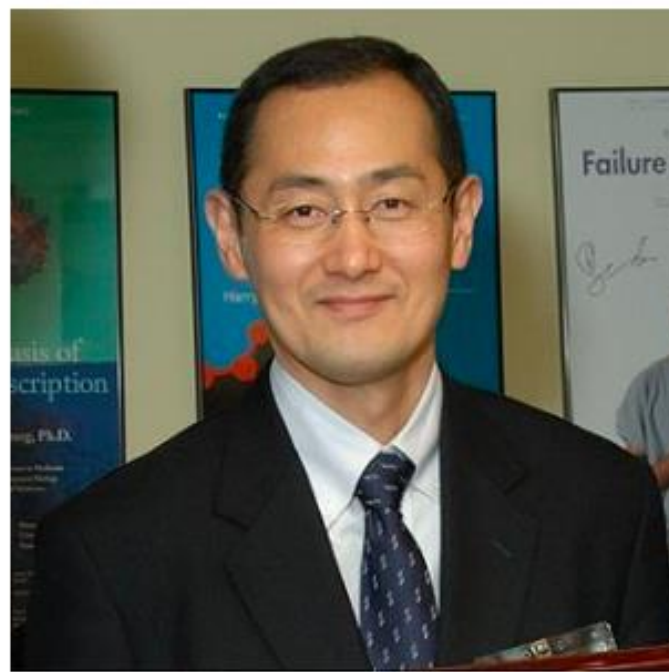
Nature Methods has retracted a 2017 paper suggesting a common gene editing technique may cause widespread collateral damage to the genome.

The notice has a long backstory: After the paper was published, it immediately drew an outcry from critics (including representatives from companies who sell the tool, whose stock fell after publication). Some critics argued that the authors, led by Vinit B. Mahajan at Stanford University, hadn't employed sufficient controls, so they couldn't be sure that the observed mutations stemmed from the tool, rather than normal background variation between mice. Only months after the paper appeared, the journal issued an expression of concern about the article. In a new preprint posted on BioRxiv on Monday, the authors concede that their critics may be right.



Kyoto University has “punitively dismissed” a researcher found guilty of falsifying nearly all of the figures in a 2017 stem cell paper.

According to an announcement Wednesday, the university fired the paper’s corresponding author, Kohei Yamamizu, after determining he had fabricated and falsified data in all but one figure in the 2017 *Stem Cell Reports*



Shinya Yamanaka

paper. The findings of the investigation, which were announced in January, found that Yamamizu, who worked at the Center for iPS Cell Research and Application (CiRA), was the only person responsible for the manipulation.

But CiRA’s director, Shinya Yamanaka—who shared a Nobel Prize for his work in stem cell biology—has taken responsibility for the incident as well. In an official statement, Yamanaka said he felt “a strong responsibility for not having prevented research misconduct at our institute:”



## 2) Novel Mechanism of Inhibition of Dendritic Cells Maturation by Mesenchymal Stem Cells via Interleukin-10 and the JAK1/STAT3 Signaling Pathway:

*Following publication of this article [1], concerns were raised regarding the presented data.*

*In Figure 5, the P-JAK1 and STAT3 Western blot panels are duplicates.*

*Four pairs of panels are duplicated in Figure 7:*

*7A panels CD86 and OX62.*

*7B panels CD86 and OX62.*

*7A panel CD11b/c and 7C panel CD11b/c.*

*7A panel MHC-II and 7B panel CD80.*

(...) continues on the next slide





*In view of the concerns regarding the reliability of the results and the absence of the raw data images, the authors and PLOS ONE Editors retract this article. The authors wish to apologize to readers.*





# Common to theses cases

- They were all peer reviewed papers
- Most of the problems were found by scientists **trying to reproduce the research**
- Comments are usually sent to the editors, or published on the Web
  - PubPeer
  - BioRxiv



# About PubPeer

## The PubPeer Foundation

The **PubPeer** Foundation is a California-registered public-benefit corporation with 501(c)(3) nonprofit status in the United States. The overarching goal of the Foundation is to improve the quality of scientific research by enabling innovative approaches for community interaction. The bylaws of the Foundation establish pubpeer.com as a service run for the benefit of its readers and commenters, who create its content. Our current focus is maintaining and developing the **PubPeer** online platform for post-publication peer review.



The PubPeer database contains all articles. Search results return articles with comments.

Mario Schietroma



To leave the first comment on a specific article, paste a unique identifier such as a **DOI**, **PubMed ID**, or **arXiv ID** into the search bar.

Search publications for: **Mario Schietroma**

2  
months  
ago

**RETRACTED: High-concentration supplemental perioperative oxygen and surgical site infection following elective colorectal surgery for rectal cancer: a prospective, randomized, double-blind, controlled, single-site trial**

Mario Schietroma, Emanuela M. Cecilia, Federico Sista, Francesco Carlei, Beatrice Pessia, Gianfranco Amicucci

ERRATUM

The American Journal of Surgery (2014)

1 comment

8  
months  
ago

**Dexamethasone for the prevention of recurrent laryngeal nerve palsy and other complications after thyroid surgery: a randomized double-blind placebo-controlled trial**

Mario Schietroma, Emanuela Marina Cecilia, Francesco Carlei, Federico Sista, Giuseppe De Santis, Laura Lancione, Gianfranco Amicucci

JAMA Otolaryngology–Head & Neck Surgery (2013)

1 comment

Home / Publications

## Dexamethasone for the prevention of recurrent laryngeal nerve palsy and other complications after thyroid surgery: a randomized double-blind placebo-controlled trial

JAMA Otolaryngology–Head & Neck Surgery (2013) - 1 Comment

doi: 10.1001/jamaoto.2013.2821 issn: 2168-6181 pubmed: 23681030 issn: 2168-619X

Mario Schietroma, Emanuela Marina Cecilia, Francesco Carlei, Federico Sista, Giuseppe De Santis, Laura Lancione, Gianfranco Amicucci

**#1 Polyommatus Arasbarani** commented 9 months ago

2017 expression of concern. <http://jamanetwork.com/journals/jamaotolaryngology/fullarticle/2645374>

 report  permalink

Reply



# WHAT IS REPRODUCIBILITY?

# What is Reproducibility?

- There is no consensus
- Scientists use slightly different definitions for reproducibility
- We will adopt one that seems to be well accepted



Dagstuhl Seminar 16041 , January 2016  
Reproducibility of Data-Oriented Experiments in e-Science  
<http://www.dagstuhl.de/en/program/calendar/semhp/?semnr=16041>

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# Definition of Reproducible Experiment in Computational Science

- An experiment composed by a sequence of steps  $S$  that has been developed at time  $T$ , on environment (hardware and OS)  $E$ , and on data  $D$  **is reproducible if** it can be executed with a sequence of steps  $S'$  (different or the same as  $S$ ) at time  $T' > T$ , on environment  $E'$  (different or the same as  $E$ ), and on data  $D'$  (different or the same as  $D$ ) with **consistent results** ( $R$  and  $R'$  consistent)



# Definition of Reproducible Experiment in Computational Science

- This definition includes both **exact reproducibility** and **approximate reproducibility**
- **Exact Reproducibility** (a.k.a. **repeatability**): requires reproducing **the exact same result**
  - $S' = S$  and  $E' = E$  and  $D' = D \Rightarrow R = R'$
- **Approximate Reproducibility**: involves producing **similar results** as the original ones
  - $S' \neq S$  or  $E' \neq E$  or  $D' \neq D \Rightarrow R \sim R'$

# Reproduce x Replicate

- **Reproduce:** to execute **the exact same experiment (same code, same data)** in a different environment
- **Replicate:** **independent investigators** address a scientific hypothesis and build up evidence for or against it (**different code, different data**)

# Replication: not easy!

- Depending on the type of the experiment, and the resources it requires, **replication may be nearly impossible**
  - May require lots of computing power
  - May require access to big telescopes
  - May require access to a particle accelerator
  - May require decades of following up subjects (e.g. drug tests)
  - ...

PENG, R. Reproducible Research in Computational Science. **Science**. V. 443:1226-1227, 2011.

# Reproducibility in Computational Science

“An attainable **minimum standard** for assessing the value of scientific claims, particularly **when full independent replication of a study is not feasible**”

“A result is said to be **reproducible** if another researcher can take the original **code** and **input data**, execute it, and re-obtain the same result.”

PENG, R. Reproducible Research in Computational Science. **Science**. V. 443:1226-1227, 2011.

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. **Frontiers in Neuroinformatics**. V.11, article 69, 2018.



Reproduce

Repeat

Re-run

Reuse

Replicate

# The R\* brouhaha

- For a program to contribute to science, it should be rerunnable ( $R^1$ ), repeatable ( $R^2$ ), reproducible ( $R^3$ ), reusable ( $R^4$ ), and replicable ( $R^5$ )

*Reproduce*

*Repeat*

*Reuse*

*Re-run*

*Replicate*

GOBLE, C. What is reproducibility? The Rbrouhaha, In: First International Workshop on Reproducible Open Science (Hannover), 2016.

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# R<sup>1</sup> - Rerunnable

- A **rerunnable** code is one that can be run again when needed
  - It becomes intrinsically difficult as code ages
  - It implies we need knowledge of the original environment (***E***), access to the code (***S***) and data (***D***)
  - ***S'*** = ***S*** and ***E'*** ~ ***E*** and ***D'*** = ***D***
  - Note that nothing is said about the **result**

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Example: Random Walk ( $R^0$ )

LISTING 0: Random walk ( $R^0$ )

[raw code](#), [archive](#)

```
import random

x = 0
for i in xrange(10):
    step = random.choice([-1,+1])
    x += step
    print x,
```

Output

```
-1, 0, -1, 0, -1, 0, -1, 0, 1, 2
# with the steps being -1,+1,-1,+1,-1,+1,-1,+1,+1,+1
```

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. *Frontiers in Neuroinformatics*. V.11, article 69, 2018.



# Example: Random Walk ( $R^0$ )

LISTING 0: Random walk ( $R^0$ )

[raw code](#), [archive](#)

```
import random

x = 0
for i in xrange(10):
    step = random.choice([-1,+1])
    x += step
    print x,
```

*Environment info is unknown.  
Does it work on any Python version?*

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. *Frontiers in Neuroinformatics*. V.11, article 69, 2018.

# Example: Random Walk ( $R^0$ )

LISTING 0: Random walk ( $R^0$ )

[raw code](#), [archive](#)

```
import random

x = 0
for i in xrange(10):
    step = random.choice([-1,+1])
    x += step
    print x,
```

*xrange and print are deprecated in Python 3*

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. *Frontiers in Neuroinformatics*. V.11, article 69, 2018.

# Example: Rerunnable Random Walk ( $R^1$ )

LISTING 1: Re-runnable random walk ( $R^1$ )

[raw code](#), [archive](#)

```
# Tested with Python 3
import random

x = 0
walk = []
for i in range(10):
    step = random.choice([-1, +1])
    x += step
    walk.append(x)

print(walk)
```

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Example: Rerunnable Random Walk ( $R^1$ )

LISTING 1: Re-runnable random walk ( $R^1$ )

[raw code, archive](#)

```
# Tested with Python 3
import random

x = 0
walk = []
for i in range(10):
    step = random.choice([-1, +1])
    x += step
    walk.append(x)

print(walk)
```

*Environment info  
Scientist is responsible  
for keeping this info*

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Repeatable ( $R^2$ )

- A **repeatable** code is one that can be rerun and that produces the same result on **successive runs**
  - Program needs to be **deterministic**
  - Control the initialization of pseudo-random number generators
  - **Previous results** need to be available (so it is possible to compare with current results)
  - $S' = S$  and  $E' \sim E$  and  $D' = D$  and  $R = R'$

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Example: Repeatable Random Walk ( $R^2$ )

LISTING 2: Re-runnable, repeatable random walk ( $R^2$ )

[raw code](#), [archive](#)

```
# Tested with Python 3
import random

random.seed(1) # RNG initialization

x = 0
walk = []
for i in range(10):
    step = random.choice([-1,+1])
    x += step
    walk.append(x)

print(walk)
# Saving output to disk
with open('results-R2.txt', 'w') as fd:
    fd.write(str(walk))
```



# Example: Repeatable Random Walk ( $R^2$ )

LISTING 2: Re-runnable, repeatable random walk ( $R^2$ )

[raw code, archive](#)

```
# Tested with Python 3
import random

random.seed(1) # RNG initialization
```

*Random seed initialization*

```
x = 0
walk = []
for i in range(10):
    step = random.choice([-1,+1])
    x += step
    walk.append(x)
```

*Save output to allow  
comparing different runs  
(again scientist is responsible  
for recording provenance)*

```
print(walk)
# Saving output to disk
with open('results-R2.txt', 'w') as fd:
    fd.write(str(walk))
```

# Initialization of Random Seeds

- Verifying that the qualitative aspects of the results and the conclusions that are made are **not tied to a specific initialization of the pseudo-random generator** is an integral part of any scientific undertaking in computational Science
- This is usually done by **repeating the simulations multiple times with different seeds**

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Reproducible ( $R^3$ )

- A result is said to be **reproducible** if **another researcher** can take the original code and input data, execute it, and re-obtain the **same (compatible) result**
- An  $R^2$  program **will not necessarily** produce the same results all the time over **different execution environments**
- $S' = S$  and  $E' = E$  and  $D' = D$  and  $R \sim R'$

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Example: Repeatable Random Walk ( $R^2$ )

LISTING 2: Re-runnable, repeatable random walk ( $R^2$ )

[raw code, archive](#)

```
# Tested with Python 3
import random

random.seed(1) # RNG init

x = 0
walk = []
for i in range(10):
    step = random.choice
    x += step
    walk.append(x)

print(walk)
# Saving output to disk
with open('results-R2.txt', 'w') as fd:
    fd.write(str(walk))
```

*Due to a change that occurred in the pseudo-random number generator between Python 3.2 and Python 3.3, executing this code in Python 3.3 will NOT generate the same results when compared to the Python 3.2 execution*

# Repeatable Random Walk Example is not reproducible

- Executed with Python 2.7–3.2, the code will produce the sequence  
-1, 0, 1, 0, -1, -2, -1, 0, -1, -2
- But with Python 3.3–3.6, it will produce  
-1, -2, -1, -2, -1, 0, 1, 2, 1, 0
- With future versions of the language, it may change still

# Reproducibility ( $R^3$ )

- Executability ( $R^1$ ) and determinism ( $R^2$ ) are necessary but not sufficient for reproducibility
- The **exact execution environment** used to produce the results must also be specified

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# Reproducibility ( $R^3$ )

- Having environment info is not enough
  - In our example, **should the code change** after the production of the results, someone provided with the last version of the code **will not be able to know which seed was used** to produce the results
  - **Result files should come alongside their context**, i.e., an exhaustive list of the parameters used as well as a precise description of the execution environment
  - The **code itself is part of that context**: the version of the code must be recorded

# Example: Reproducible Random Walk ( $R^3$ )

LISTING 3: Re-runnable, repeatable, reproducible random walk ( $R^3$ )

[raw code](#), [archive](#)

```
# Copyright (c) 2017 N.P. Rougier and F.C.Y. Benureau
# Release under the BSD 2-clause license
# Tested with 64-bit CPython 3.6.2 / macOS 10.12.6
import sys, subprocess, datetime, random

def compute_walk():
    x = 0
    walk = []
    for i in range(10):
        if random.uniform(-1, +1) > 0:
            x += 1
        else:
            x -= 1
        walk.append(x)
    return walk
```

```
# If repository is dirty, don't run anything
if subprocess.call(("git", "diff-index",
                    "--quiet", "HEAD")):
    print("Repository is dirty, please commit first")
    sys.exit(1)
```

```
# Get git hash if any
hash_cmd = ("git", "rev-parse", "HEAD")
revision = subprocess.check_output(hash_cmd)
```

```
# Unit test
random.seed(42)
assert compute_walk() == [1, 0, -1, -2, -1, 0, 1, 0, -1, -2]
```

```
# Random walk for 10 steps
seed = 1
random.seed(seed)
walk = compute_walk()
```

*Use git to keep track of code versions*

```
# If repository is dirty, don't run anything
if subprocess.call(("git", "diff-index",
                    "--quiet", "HEAD")):
    print("Repository is dirty, please commit first")
    sys.exit(1)
```

```
# Get git hash if any
hash_cmd = ("git", "rev-parse", "HEAD")
revision = subprocess.check_output(hash_cmd)
```

```
# Unit test
random.seed(42)
assert compute_walk() == [1, 0, -1, -2, -1, 0, 1, 0, -1, -2]
```

```
# Random walk for 10 steps
seed = 1
random.seed(seed)
walk = compute_walk()
```

*Test for reproducibility*

```
# Display & save results
print(walk)
results = {
    "data"      : walk,
    "seed"      : seed,
    "timestamp": str(datetime.datetime.utcnow()),
    "revision"  : revision,
    "system"    : sys.version}
with open("results-R3.txt", "w") as fd:
    fd.write(str(results))
```

*Record environment  
with output data*

# Quick Recap

- **Reproducibility** implies **re-runnability** and **repeatability** and **availability**, yet imposes additional conditions
- Dependencies and platforms must be described as precisely and as specifically as possible
- Parameters values, the version of the code, and inputs should accompany the result files
- The **data and scripts behind the graphs must be published**

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.



# Reusability (R<sup>4</sup>)

- Making your program **reusable** means it can be easily used, and **modified**, by you and other people, inside and outside your lab
- The easier it is to use your code, the lower the threshold is for other to study, modify and extend it
  - This implies it should be **well documented!**

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Reusability (R<sup>4</sup>)

- Scientists constantly face the constraint of time
  - if a model is **available**, **documented**, and can be installed, run, and **understood** all in a few hours, it will be preferred over another that would require weeks to reach the same stage
- A **reproducible** and **reusable** code offers a platform both verifiable and easy-to-use, fostering the development of derivative works by other researchers on solid foundations
- Those derivative works **contribute to the impact of your original contribution (citations!!)**

# Reusability (R<sup>4</sup>)

- Reusability is not as indispensable a requirement as re-runability, repeatability, and reproducibility
- But it can contribute to strengthen reproducibility and re-runability over the long-term

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. *Frontiers in Neuroinformatics*. V.11, article 69, 2018.

# Example: Reusable Random Walk ( $R^4$ )

LISTING 4: Re-runnable, repeatable, reproducible, reusable random walk ( $R^4$ )  
raw code, archive

```
# Copyright (c) 2017 N.P. Rougier and F.C.Y. Benureau
# Release under the BSD 2-clause license
# Tested with 64-bit CPython 3.6.2 / macOS 10.12.6
import sys, subprocess, datetime, random

def compute_walk(count, x0=0, step=1, seed=0):
    """Random walk
    count: number of steps
    x0    : initial position (default 0)
    step  : step size (default 1)
    seed  : seed for the initialization of the
            random generator (default 0)
    """
```

```
random.seed(seed)
x = x0
walk = []
for i in range(count):
    if random.uniform(-1, +1) > 0:
        x += 1
    else:
        x -= 1
    walk.append(x)
return walk
```

```
def compute_results(count, x0=0, step=1, seed=0):
    """Compute a walk and return it with context"""
    # If repository is dirty, don't do anything
    if subprocess.call(("git", "diff-index",
                        "--quiet", "HEAD")):
        print("Repository is dirty, please commit")
        sys.exit(1)

    # Get git hash if any
    hash_cmd = ("git", "rev-parse", "HEAD")
    revision = subprocess.check_output(hash_cmd)

    # Compute results
    walk = compute_walk(count=count, x0=x0,
                        step=step, seed=seed)
```



```
return {  
    "data"      : walk,  
    "parameters": {"count": count, "x0": x0,  
                   "step": step, "seed": seed},  
    "timestamp" : str(datetime.datetime.utcnow()),  
    "revision"   : revision,  
    "system"     : sys.version}
```

```
if __name__ == "__main__":
    # Unit test checking reproducibility
    # (will fail with Python<=3.2)
    assert (compute_walk(10, 0, 1, 42) ==
            [1, 0, -1, -2, -1, 0, 1, 0, -1, -2])

    # Simulation parameters
    count, x0, seed = 10, 0, 1
    results = compute_results(count, x0=x0, seed=seed)

    # Save & display results
    with open("results-R4.txt", "w") as fd:
        fd.write(str(results))
    print(results["data"])
```

# Tips for Producing Reusable Code

- **Avoid** hardcoded or magic numbers
- **Magic numbers** are those present directly in the source code (no name, no semantics)
- **Hardcoded values** are variables that cannot be changed through an argument or a parameter configuration file
- In the  $R^3$  Random Walk example, the seed is hardcoded, and the number of steps is a magic number

# Tips for Producing Reusable Code

- Code behavior **should not** be changed by commenting/uncommenting code
- Instead, it should be **explicitly set** through parameters that are accessible to the end user
- This improves reproducibility in two ways
  - it allows those conditions to be recorded as parameters in the result files, and
  - it allows to define separate scripts to run or configuration files to load to produce each of the figures of the published paper

# Replicability ( $R^5$ )

“the replication of important findings by **multiple independent investigators** is fundamental to the accumulation of scientific evidence”

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Replicability ( $R^5$ )

- **Replicability** is the implicit assumption that an article that does not provide the source code makes: that the description it provides of the algorithms is sufficiently precise and complete to re-obtain the results it presents
- Replicating implies **writing a new code** matching the conceptual description of the article, in order to obtain the same **(compatible) results**
- $S' \neq S \text{ and } (E' \neq E \text{ or } D' \neq D) \Rightarrow R \sim R'$

# Replicability (R<sup>5</sup>)

- Replication affords robustness to the results
  - should the original code contain an error, a different codebase creates the possibility that this error will not be repeated

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. *Frontiers in Neuroinformatics*. V.11, article 69, 2018.

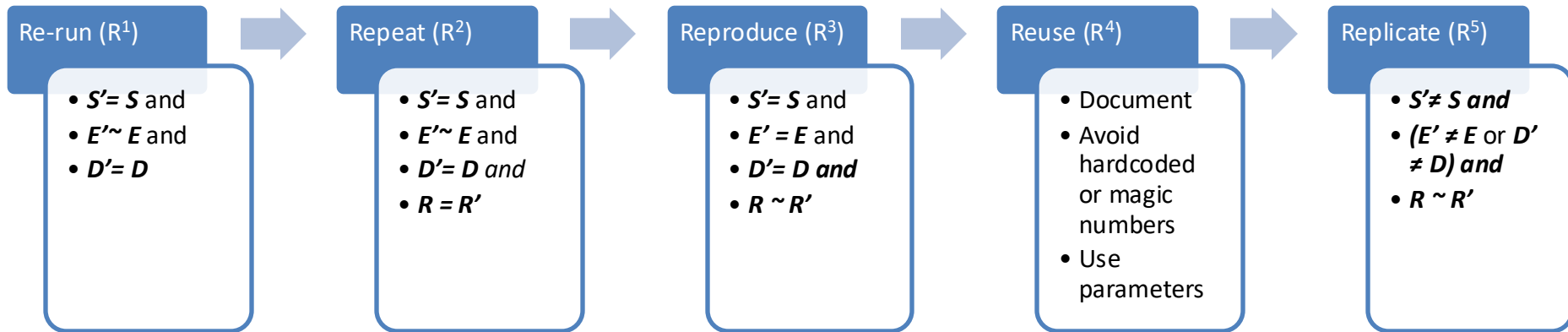


# Replicability ( $R^5$ )

- Every paper is a mistake or a forgotten parameter away from irreplicability
- Replication efforts use the paper first, and then the reproducible code that comes along with it whenever the paper falls short of being precise enough

BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. Frontiers in **Neuroinformatics**. V.11, article 69, 2018.

# Summary



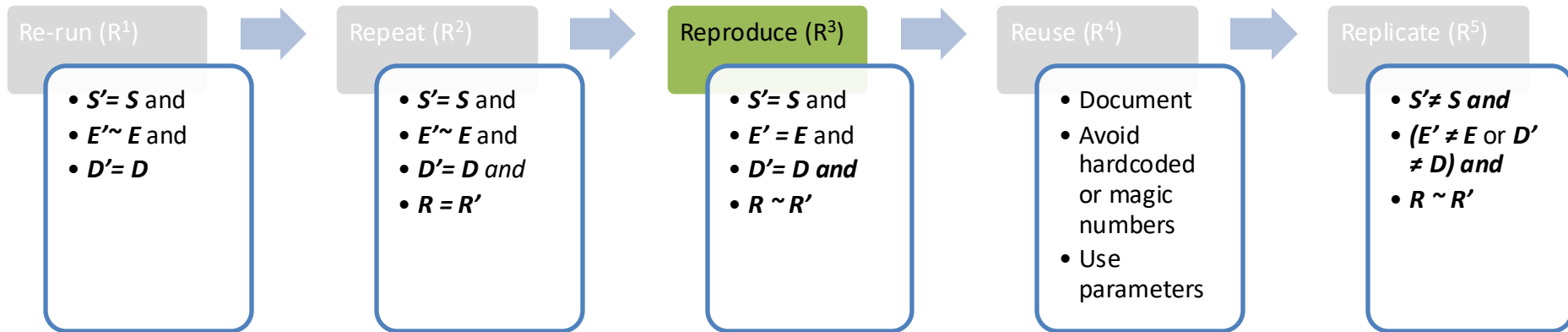
Code (local) + Environment + Input Data

Same (Compatible) Output

Same Environment

Publicly available +  
Documentation

# Summary



*Minimum  
Scientific  
Standard*

# But we are not there yet...

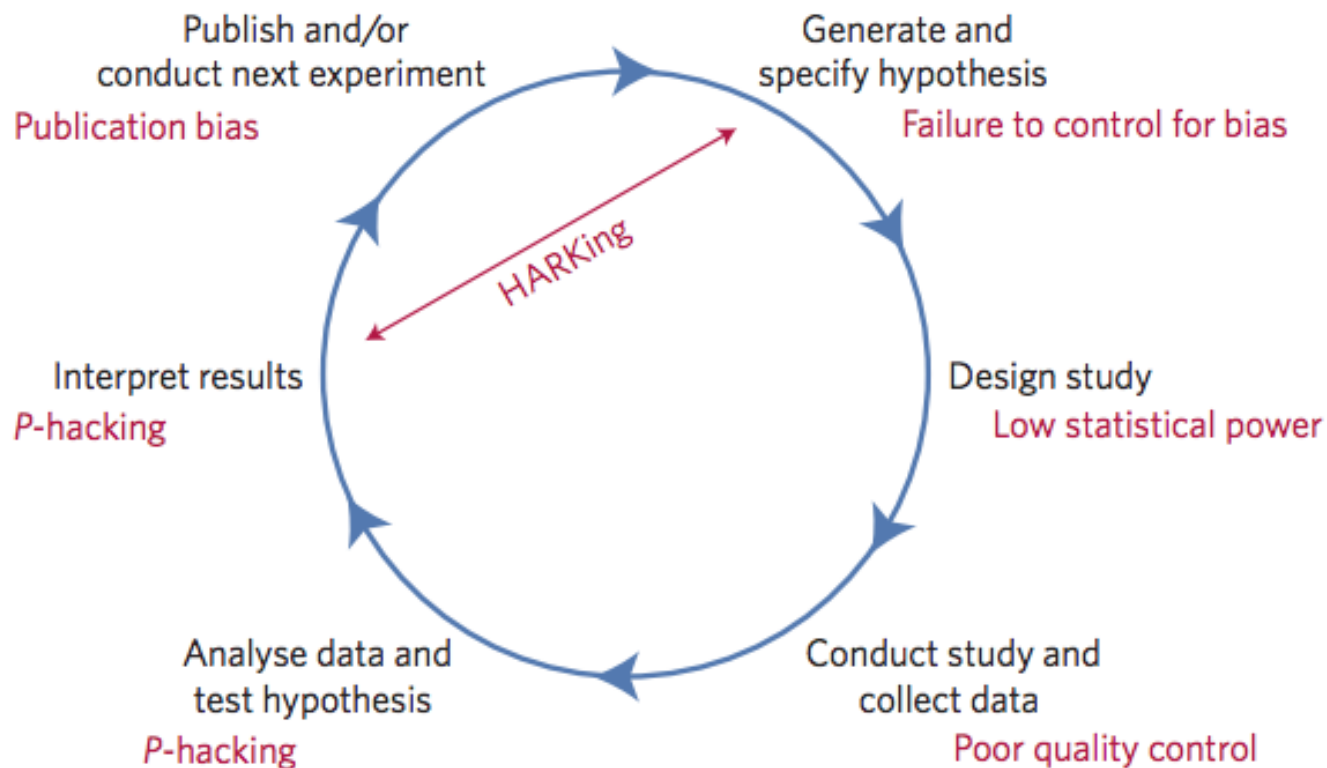
- Reproducibility **is still not the norm** for computational experiments
- Scientists argue that it is **time-consuming** to create reproducible experiments
- **Usability** is an important requirement for a broader adoption of reproducibility
- “An independent user should be able to reproduce the results with a single mouse click”

FREIRE, J.; CHIRIGATI, F. Provenance and the Different Flavors of Computational Reproducibility. **IEEE Data Engineering Bulletin**. V. 41:15-26, 2018.

# Making Reproducibility Easier

- Scientist should focus on research rather than making their code capture its own provenance
- There are several tools to easy reproducibility
  - noWorkflow, Sumatra, Reprozip, etc.
- Improvements still needed to make them “one mouse click away from reproducibility”

# CAUSES OF NON-REPRODUCIBLE RESULTS



**Figure1 | Threats to reproducible science.** An idealized version of the hypothetico-deductive model of the scientific method is shown. Various potential threats to this model exist (indicated in red), including lack of replication<sup>5</sup>, hypothesizing after the results are known (HARKing)<sup>7</sup>, poor study design, low statistical power<sup>2</sup>, analytical flexibility<sup>51</sup>, P-hacking<sup>4</sup>, publication bias<sup>3</sup> and lack of data sharing<sup>6</sup>. Together these will serve to undermine the robustness of published research, and may also impact on the ability of science to self-correct.

# p-hacking

While collecting and analyzing data, researchers have many decisions to make, including whether to collect more data, which outliers to exclude, which measure(s) to analyze, which covariates to use, and so on. If these decisions are not made in advance but rather are made as the data are being analyzed, then researchers may make them in ways that self-servingly increase their odds of publishing (Kunda, 1990). Thus, rather than placing entire studies in the file-drawer, researchers may file merely the subsets of analyses that produce nonsignificant results. We refer to such behavior as *p-hacking*.<sup>1</sup>

SIMONSOHN, U., NELSON, L., SIMMONS, J. P-Curve: A Key to the File-Drawer. Journal of Experimental Psychology: General. V. 143(2):534-547, 2014.



**Table1 | A manifesto for reproducible science.**

Theme	Proposal	Examples of initiatives/potential solutions (extent of current adoption)	Stakeholder(s)
Methods	Protecting against cognitive biases	All of the initiatives listed below (* to ****) Blinding (**)	J, F
	Improving methodological training	Rigorous training in statistics and research methods for future researchers (*) Rigorous continuing education in statistics and methods for researchers (*)	I, F
	Independent methodological support	Involvement of methodologists in research (**) Independent oversight (*)	F
	Collaboration and team science	Multi-site studies/distributed data collection (*) Team-science consortia (*)	I, F
Reporting and dissemination	Promoting study pre-registration	Registered Reports (*) Open Science Framework (*)	J, F
	Improving the quality of reporting	Use of reporting checklists (**) Protocol checklists (*)	J
	Protecting against conflicts of interest	Disclosure of conflicts of interest (***) Exclusion/containment of financial and non-financial conflicts of interest (*)	J
Reproducibility	Encouraging transparency and open science	Open data, materials, software and so on (* to **) Pre-registration (**** for clinical trials, * for other studies)	J, F, R
Evaluation	Diversifying peer review	Preprints (* in biomedical/behavioural sciences, **** in physical sciences) Pre- and post-publication peer review, for example, Publons, PubMed Commons (*)	J
Incentives	Rewarding open and reproducible practices	Badges (*) Registered Reports (*) Transparency and Openness Promotion guidelines (*) Funding replication studies (*) Open science practices in hiring and promotion (*)	J, I, F

Estimated extent of current adoption: \*, <5%; \*\*, 5–30%; \*\*\*, 30–60%; \*\*\*\*, >60%. Abbreviations for key stakeholders: J, journals/publishers; F, funders; I, institutions; R, regulators.

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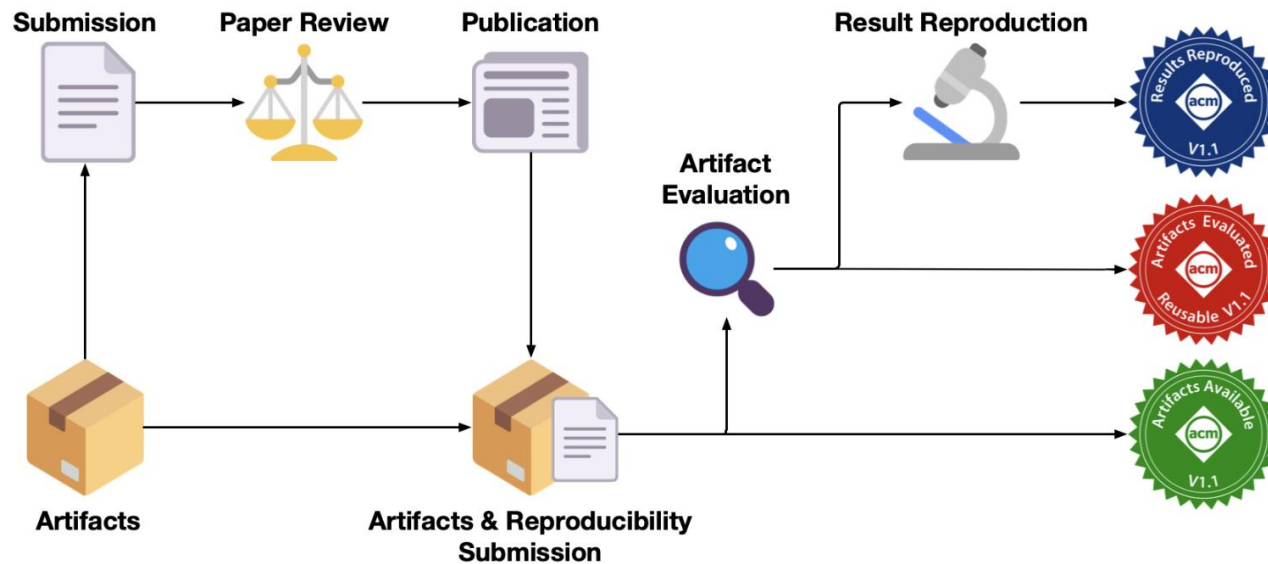
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# Incentives

- ACM SIGMOD Most Reproducible Paper Award
- ACM SIGMOD Availability and Reproducibility

How it works?

On the high-level, the workflow of the SIGMOD Availability and Reproducibility process is shown below. For more details keep scrolling to the [Process](#) and [Guidelines](#).















<https://reproducibility.sigmod.org/reports.html>

## Badges and Reproducibility Reports

The reviewers submit their reports, which are published in this page. Following is the list of papers that passed the reproducibility test, the functionality test, and/or have made their artifacts available, along with the reproducibility reports (when applicable) and the badges awarded. The reproducibility reports are maintained here starting from reproduced papers of SIGMOD 2020.



To access all reproduced papers please click [here](#).

## ACM SIGMOD 2023 Badges and Reproducibility Results

Paper Title	Report	Badges Awarded
<a href="#">Polaris: Enabling Transaction Priority in Optimistic Concurrency Control</a>		  
<a href="#">SafeBound: A Practical System for Generating Cardinality Bounds</a>		  
<a href="#">Efficient Star-based Truss Maintenance on Dynamic Graphs</a>		  

# Incentives

- ICSE “Artifacts Evaluated Reusable”

14:00 - 14:20	☆	<b>Big Bangs and Small Pops: On Critical Cyclomatic Complexity and Developer Integration Behavior</b> Daniel Ståhl Ericsson AB, Antonio Martini University of Oslo, Norway, Torvald Mårtensson Saab AB	INDUSTRY PROGRAM	SEIP
14:20 - 14:40	☆	<b>Predictive Test Selection</b> Mateusz Machalica Facebook, Inc., Alex Samykin Facebook, Inc., Meredith Porth Facebook, Inc., Satish Chandra Facebook	INDUSTRY PROGRAM	SEIP
14:40 - 15:00	☆	<b>Assessing Transition-based Test Selection Algorithms at Google</b> Claire Leong Google / UNSW, Abhayendra Singh Google, Inc, Mike Papadakis University of Luxembourg, Yves Le Traon University of Luxembourg, John Micco Netflix	INDUSTRY PROGRAM	SEIP
15:00 - 15:20	☆	<b>Automated Reporting of Anti-Patterns and Decay in Continuous Integration</b> Carmine Vassallo University of Zurich, Sebastian Proksch University of Zurich, Harald Gall University of Zurich, Massimiliano Di Penta University of Sannio <a href="#">Pre-print</a>	TECHNICAL TRACK	 

# Incentives

- Reproducibility Section of Information Systems Journal
  - <https://www.elsevier.com/journals/information-systems/0306-4379/guide-for-authors>

# Provenance of these slides

- BENUREAU, F., ROUGIER, N. Re-run, Repeat, Reproduce, Reuse, Replicate: Transforming Code into Scientific Contributions. *Frontiers in Neuroinformatics*. V.11, article 69, 2018.
- FREIRE, J.; CHIRIGATI, F. Provenance and the Different Flavors of Computational Reproducibility. *IEEE Data Engineering Bulletin*. V. 41:15-26, 2018.
- GOBLE, C. What is reproducibility? The Rbrouhaha, In:First International Workshop on Reproducible Open Science (Hannover), 2016.
- MUNAFÒ, M. et al. A manifesto for reproducible science. *Nature Human Behaviour*. V. 1: article 21, 2017.
- SIMONSOHN, U., NELSON, L., SIMMONS, J. P-Curve: A Key to the File-Drawer. *Journal of Experimental Psychology: General*. V. 143(2):534-547, 2014.