

AGILE Conference

Reproducibility Review 2021

Daniel Nüst, AGILE 2021 Reproducibility Chair
daniel.nuest@uni-muenster.de

<https://doi.org/10.5281/zenodo.4926269>





<https://reproducible-agile.github.io/>

2017, '18 & '19: Workshops on reproducibility

2019: Reproducible publications at AGILE conferences (initiative)

2020: First AGILE reproducibility review

2021: Second AGILE reproducibility review



<https://giphy.com/gifs/usnationalarchives-nasa-scientist-scientists-1F1JyGzhiSAA8Vuhn>

THE CONVERSATION

COVID-19 Arts · Culture · Business · Economy Cities · Education · Environment · Energy Health · Medicine Politics · Society **Science · Technology**

How computers broke science - and what we can do to fix it

November 9, 2015 11:55am GMT

Computer ... or back to the 1940s US Army

Reproducibility is one of the cornerstones of science. Made popular by British scientist Robert Boyle in the 1660s, the idea is that a discovery should be reproducible before being accepted as scientific knowledge.

Author: Ben Marwick, Associate Professor of Technology, University of Washington

<https://theconversation.com/how-computers-broke-science-and-what-we-can-do-to-fix-it-49938>

Claerbout's claim:

An article about computational science in a scientific publication is not the scholarship itself, it is merely advertising of the scholarship. The actual scholarship is the complete software development environment and the complete set of instructions which generated the figures.

<https://doi.org/10.1190/1.1822162>
https://doi.org/10.1007/978-1-4612-2544-7_5

CC-BY 3.0, Sebastian Bertalan, Wikimedia Commons

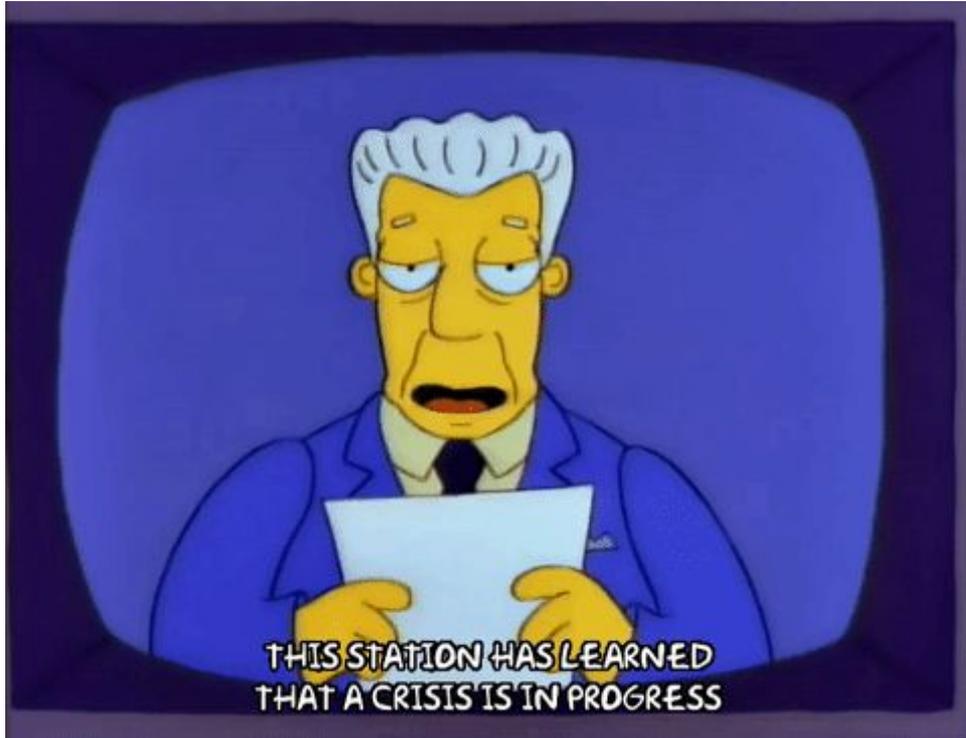


<https://giphy.com/gifs/with-computers-fascination-PxSFAnuubLkSA>



<https://giphy.com/gifs/david-hasselhoff-M3o3fl9nrxG4o>

Crisis? Crisis of what?



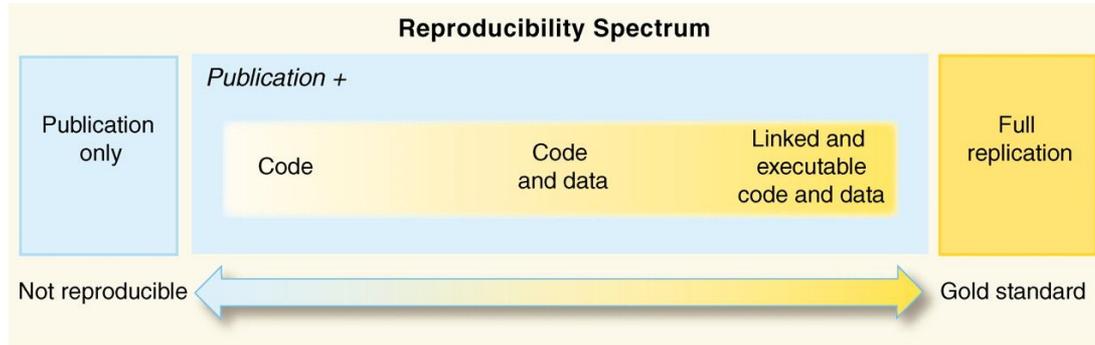
- Credibility crisis?
- Replicability crisis?
- Reproducibility crisis?
- Robustness crisis?
- Generalisability crisis?



		Data	
		Same	Different
Analysis	Same	Reproducible	Replicable
	Different	Robust	Generalisable

<https://the-turing-way.netlify.app/reproducible-research/overview/overview-definitions.html>

Reproducible Research & Open Science



<https://doi.org/10.1126/science.1213847>



"Science should be 'show me', not 'trust me'; it should be 'help me if you can', not 'catch me if you can'."

Rather than reproducibility, should we be looking at preproducibility? [@Nature_wellc.me/2IMNuiq](https://twitter.com/Nature_wellc.me/2IMNuiq)

151 15:55 - 28. Mai 2018

Preproducibility

Before reproducibility must come preproducibility
Instead of arguing about whether results hold up, let's push to prov...
nature.com

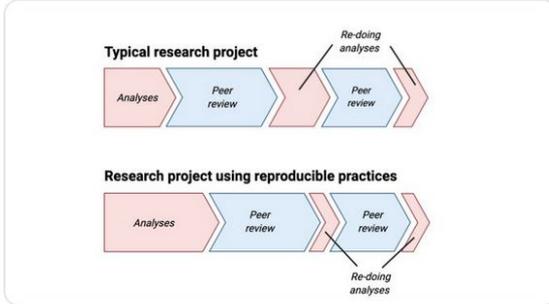
<https://doi.org/10.1038/d41586-018-05256-0>



Dan Quintana
@dsquintana

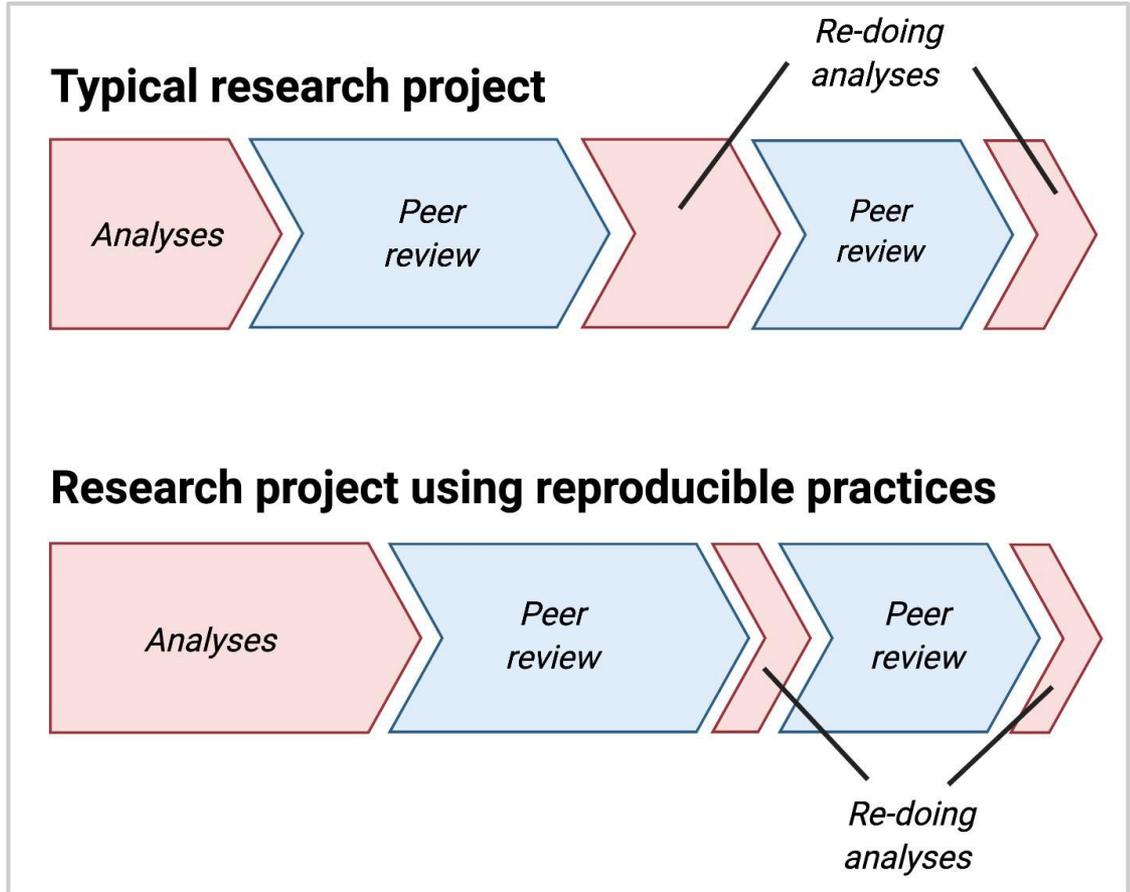
In my experience, you don't lose time doing reproducible science—you just *relocate* how you're spending it

[Tweet übersetzen](#)



4:13 nachm. · 26. Nov. 2020 · TweetDeck

107 Retweets 20 Zitierte Tweets 536 „Gefällt mir“-Angaben



Quintana, D. S. (2020, November 28). Five things about open and reproducible science that every early career researcher should know. <https://doi.org/10.17605/OSF.IO/DZTVQ>

AGILE Reproducible Paper Guidelines: Contents & First Revision

AGILE Reproducible Paper Guidelines



<https://doi.org/10.17605/OSF.IO/CB7Z8>

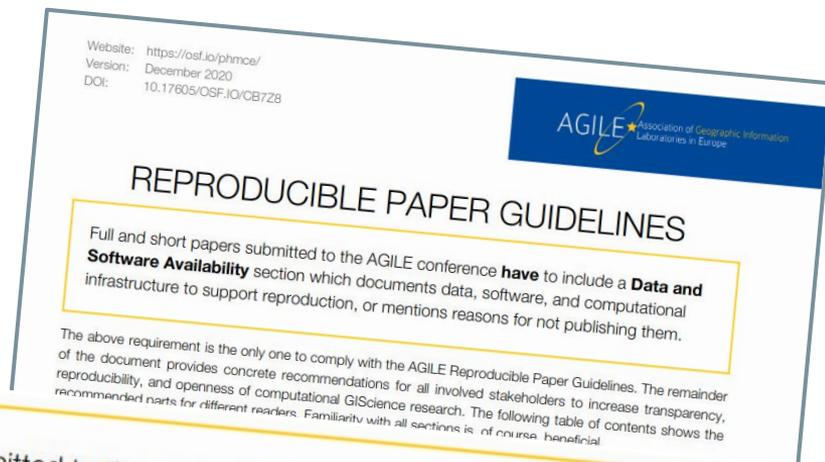
Created by AGILE Initiative in 2019, see report at <https://osf.io/hupxr/> and updated in 2020

Transparency

Promotion

Acknowledge spectrum

GIScience



	Reproducibility Checklist Helps to ensure authors and reviewers do not miss anything important.	2
	Author Guidelines Show how to write the Data and Software Availability Section and give practical recommendations to make data and computational workflows reproducible.	4
	Writing the Data and Software Availability Section Including Data in Research Papers Including Computational Workflows in Research Papers	
	Scientific Reviewer Guidelines Describe role in evaluating plausibility and completeness of the data and software availability documentation.	7
	Reproducibility Reviewer Guidelines Describe role in evaluating reproducibility of the data and software availability documentation.	

The guidelines

<https://doi.org/10.17605/OSF.IO/CB7Z8>

Reproducibility checklist

Author guidelines

Writing DASA section

Data in Research Papers

Computational workflows in Research Papers

Reviewer guidelines

Reproducibility reviewer guidelines

Background

REPRODUCIBLE PAPER GUIDELINES

Full and short papers submitted to the AGILE conference **have** to include a **Data and Software Availability** section which documents data, software, and computational infrastructure to support reproduction, or mentions reasons for not publishing them.

The above requirement is the only one to comply with the AGILE Reproducible Paper Guidelines. The remainder of the document provides concrete recommendations for all involved stakeholders to increase transparency, reproducibility, and openness of computational GIScience research. The following table of contents shows the recommended parts for different readers. Familiarity with all sections is, of course, beneficial.

Author
Reproducibility Reviewer
Scientific Reviewer

	Reproducibility Checklist Helps to ensure authors and reviewers do not miss anything important.	2
	Author Guidelines Show how to write the Data and Software Availability Section and give practical recommendations to make data and computational workflows reproducible. Writing the Data and Software Availability Section Including Data in Research Papers Including Computational Workflows in Research Papers	4
	Scientific Reviewer Guidelines Describe role in evaluating plausibility and completeness of the data and software availability documentation.	7
	Reproducibility Reviewer Guidelines Describe role and approach to execute workflows and clarify efforts.	8
	Background	10

Further resources

These guidelines can not cover all details of the reproducibility review at AGILE conferences. For more information for authors, translations, and practical examples see the [guidelines wiki](#). For more information about the review process and deadlines, see the [process description](#). For any questions, please visit the AGILE Discourse server's [forum for the Reproducible Paper Guidelines](#).

Notable updates in version 2 (December 2020)

Updated checklist (clarity + structure, extra items for after acceptance)

Reordering of contents (important stuff first!) and details moved to Wiki

TOC + visual aides for reader roles

Updated resources + language improvements

Removed “intermediate” level for computational workflows

NEW: Reproducibility reviewer guidelines

<https://doi.org/10.17605/OSF.IO/CB7Z8>

Author	Reproducibility Reviewer	Scientific Reviewer		
			Reproducibility Checklist Helps to ensure authors and reviewers do not miss anything important.	2
			Author Guidelines Show how to write the Data and Software Availability Section and give practical recommendations to make data and computational workflows reproducible. Writing the Data and Software Availability Section Including Data in Research Papers Including Computational Workflows in Research Papers	4
			Scientific Reviewer Guidelines Describe role in evaluating plausibility and completeness of the data and software availability documentation.	7
			Reproducibility Reviewer Guidelines Describe role and approach to execute workflows and clarify efforts.	8
			Background	10

Checklist and writing the DASA section

REPRODUCIBILITY CHECKLIST

For all **datasets** included/produced in the paper, check if data:

- Is provided in a non-proprietary format
- Is documented for third parties to reuse
- Is accessible in a public repository and has an open data licence

For all **software tools/libraries/packages** and **computational workflows** included/produced, check if:

- Reproduction steps are explained in a README (plain text file), flowchart, or script
- Computational environments (including hardware) are documented or provided
- Versions of relevant software components (libraries, packages) are provided
- All parameters and expected execution times for the computational workflow are provided
- Software developed by the authors is available in a public repository and has an open licence
- There is a clear connection between **tables, figures, maps, and statistical values** and the data and code that they are based on, e.g., using file names or documentation in the README

In the **Data and Software Availability section**, check if you include:

- Data and software statements (see examples below)
- The reasons, if any, for not being able to share (parts of) data or code

For all **data and software** check that:

- All datasets and code (used or mentioned) are assigned DOIs
- Datasets and code are cited throughout the paper

After acceptance in the **camera-ready paper** check that:

- If data has been shared privately or anonymously for peer review, they are updated with all metadata and accessible via a DOI and referenced from the paper
- If a reproducibility review report will be published for your paper, a DOI URL in the Data and Software Availability section is included using the following template:
A reproducibility report for this paper is available confirming that [considerable parts of the computational workflow / all results / Figures 1 and 4] could be independently reproduced, see https://doi.org/link_to_report.

WRITING THE DATA AND SOFTWARE AVAILABILITY SECTION

The DASA section provides references to where data, software and documentation is available (e.g., paper section or README file) and under what conditions (e.g., copyright, licenses or access procedures for protected data). It should be concise and contain persistent links to repositories using Digital Object Identifiers' (DOI). You may remove links for anonymity during peer review ("xxx"), or share anonymized links^s if your repository supports them. Data, software and (third-party) tools should be cited following recommended citation or standard citation guidelines. Possible statements for the DASA section are provided below. You may include one of these statements or draft your own.

Statements for non-computational or conceptual work

No data or code was collected, developed, or used in this work.

The full list of reviewed literature is available at [link to attachment or citable deposit of bibliography].

The full concept maps are available at [link] and the ideas were first sketched in a blog post at [link].

Research data/code supporting this publication ...

... is available in [name of the repository(-ies)] and is accessible via the following DOI [DOI link(s)]

... was accessed on [date of dataset access/download] with the following [query parameters, if applicable] under the license [dataset license].

... was downloaded manually using the services at [name of organisation] (using a departmental subscription for costs) and [name of organisation]. The compiled dataset cannot be redistributed due to licensing restrictions.

... is not available due to [indicate reasons, e.g., licenses, sensitive data on human subjects, privacy statements; if there are processes to obtain the data, describe them].

The computational workflow supporting this publication ...

... is executed via [choose, e.g., a single command/file, a workflow management software, a set of numbered scripts] published under license [the license] at [DOI of repository].

... is published in a [language] module/package at [link of software project]. The used version is archived at [DOI of repository].

... is provided as a [container/VM] published at [DOI of repository] with instructions included in the file README.md in the repository.

The guidelines for data



“What if...” and
Examples not shown

INCLUDING DATA IN RESEARCH PAPERS

	Minimum requirements	Recommended practices
What?	<ul style="list-style-type: none">• All input data and configuration• Data description/documentation, including provenance, field or column types, etc.• If data is retrieved from an external source, documentation on collection queries and download steps	<ul style="list-style-type: none">• Standardised, discipline-specific metadata⁸ and ontologies to describe your data• Data download scripts
Where?	<ul style="list-style-type: none">• Publish data in a public repository providing a DOI• Cite data (including date and version) in the paper	<ul style="list-style-type: none">• Discipline- or data type-specific repository⁹• Include recommended citation in dataset description (unless already provided by repository)• Create a registration for OSF projects¹⁰ and use the DOI to cite it
How?	<ul style="list-style-type: none">• Use open data formats; export from proprietary format for publication• Specify the license	<ul style="list-style-type: none">• Use plain text-based file formats

The guidelines for computational workflows



INCLUDING COMPUTATIONAL WORKFLOWS IN RESEARCH PAPERS

	Minimum requirements	Recommended practices
What? Computational environment	<ul style="list-style-type: none"> Describe the used environment and computational infrastructure, e.g., hardware specs, operating system List software versions Cite used software¹⁴ 	<ul style="list-style-type: none"> Provide the actual environment, e.g., a Dockerfile + container¹⁵ or a Virtual Machine (e.g., using OSGeo-Live) Provide a pinned freeze of your dependencies (structured configuration files with dependency information) Add a colophon or “reproducibility receipt”¹⁶ to your notebooks Installation and execution instructions for different operating systems
Computation steps	<ul style="list-style-type: none"> Document the detailed steps in a text file and/or flowchart (every action/click) Document expected execution times given computing power unless negligible Ask a colleague to try out the instructions 	<ul style="list-style-type: none"> Scripts/models and a README file that explains their use All figures are fully scripted and a peer has read your README’s instructions (incl. interactive visualisations and interactive adjustments) Multi-panel plots are composited with scripts¹⁷ Software package with structured metadata¹⁸, tests/CI¹⁹, and a pipeline framework²⁰ or workflow language²¹ Live documents for analyses, e.g., Binder²² Live demo of APIs/online applications (e.g., anonymous cloud resources, such as Google Cloud Run or AWS) Subset or a synthetic dataset for quick evaluation
Where?	<ul style="list-style-type: none"> Repository providing a persistent identifier, e.g., a DOI or SWHID²³ 	<ul style="list-style-type: none"> Versioned code repository, such as GitHub or GitLab, and ongoing open development
How? Tools used	<ul style="list-style-type: none"> Use generally available tools (avoid proprietary tools that are not available to reviewers and other researchers) 	<ul style="list-style-type: none"> Use and create Open Source tools Cite core modules/tools/language used
Development practices	<ul style="list-style-type: none"> Use clear licenses²⁴ that fit your environment Follow one of “Good enough practices in scientific computing”²⁵ 	<ul style="list-style-type: none"> Follow all “Good enough practices..” Use development guidelines for your environment / language of choice (e.g., for R²⁶)

Scientific reviewer guidelines... concerning the reproducibility review only!



SCIENTIFIC REVIEWER GUIDELINES

This section clarifies the expectations and role of the scientific reviewer with respect to the reproducible paper guidelines. For information for the Reproducibility Reviewer, please see the following section.

Reproducibility is considered good scientific practice that provides input for the quality assessment of a paper. Therefore, reviewers of AGILE papers should be aware of the **author guidelines on reproducibility** and be familiar with the **reproducibility checklist**, as well as the expected content of the **mandatory data and software availability section**. Using this information, reviewers should evaluate the plausibility and completeness of the data and software availability documentation, and whenever possible and readily available **include feedback on reproducibility aspects** in their comments. Scientific reviewers are free to but **are not expected to attempt reproductions of computations**.

Data and software availability documentation provide an additional set of information for assessing the quality of research presented in a manuscript. Reviewers are asked to know about the AGILE reproducible paper guidelines and to consider the level of reproducibility reached in a manuscript. To do so, they shall assume the position of someone who would like to reproduce the submitted work to assess whether the provided material is likely to allow reproduction of the submitted work. Based on this impression, reviewers may challenge authors regarding the level of reproducibility reached, if any statements are made regarding reproducibility in a manuscript.

Scientific reviewers are not required to actually reproduce a manuscript, but, if the data and code are provided in an anonymous format, and if a reviewer attempts to reproduce all or parts of the submitted work, then they are asked to document the process and outcomes (see Reproducibility Reviewer Guidelines below). Please reach out to the reproducibility chair if you are keen on conducting a reproducibility review for a paper you are reviewing.

The peer review of AGILE papers is a fully anonymous peer review, i.e. authors and reviewers do not know each other's identity. Reviewers should be supportive to authors and consider potential limitations in access to resources due to anonymisation. Since the provision of information to help reproduction of a paper can accidentally lead to disclosure of an author's identity, the reviewers should not use any such additional information to the disadvantage of the authors. The reviewers' comments provided to the authors are expected to be neutral²⁸ and contribute to improved reproducibility of the reported findings.

The guidelines for reproducibility reviewers

Ideal vs. realistic

Role & skills

Examples for “Do’s and Don’ts”:

- Do shift burden to author
- Do encourage and set examples
- Do *not* accept private data sharing
- Document your work in report (impact)
- Be kind (career stage, knowledge, privileges)
- No rummaging

Reproducibility reviewer guidelines

Reproducibility reviewers conduct a complimentary review of the computational workflow that is published with a full paper that is provisionally accepted after the scientific review process. They read the paper insofar as needed to **reproduce the computation, using the abstract and the Data and Software Availability section** (DASA) as starting points. Ideally, these sections of the paper together with a README file are sufficient for the reproduction. When reproducibility reviewers get stuck, they take advantage of the option to **communicate** with the authors early and often. Reproducibility reviewers should be aware of the different reproducibility levels (see Author Guidelines above) to **recommend Improvements** to the authors, but they are not responsible for making a workflow transparent or executable. Reproducibility reviewers **write a reproducibility report** documenting the results of their reproduction attempt and their communication with the authors. The report is published if the reproduction was, at least in part, successful. It is shared with the authors if the reproduction attempt was stopped but already contains relevant feedback.

Reproducibility review coordination

The reproducibility chair will be your contact person regarding supporting infrastructure and getting access to the private discussion forum for reproducibility reviewers on the AGILE Discourse server²³. This forum is used to assign, under the leadership of the reproducibility chair, respective topical and technical skills, and share mat report.

Goals and scope

While the AGILE reproducible paper guidelines are reproducibility success rate for accepted papers, understanding, and ultimately community adoption the tasks as reproducibility reviewer harder and progress review is an extra merit for an accepted paper, but acceptance. The reproducibility reviewer should be awi might “take the extra few steps” needed. This non-ex one reproducibility reviewer is assigned per paper. } scientific reviewer on the same paper, but the roles of in of the reproducibility review is roughly in line with t community is worth exploring for further examples and reproduction, e.g., the recreation of some but not all of t though what is “good enough” may change over time, or the reproducibility committee chair in case of doubt.

Reproducibility reviewer skills

A reproducibility review is a learning experience for bc AGILE community to increase openness and transpare amount of time you should spend on a reproduction at as the research you are tasked to reproduce. However few minutes of being stuck and not spending more t depends also on your interest, time budget, and skills i get basic familiarity with package managers and virtus DESCRIPTION files and renv for R, npm for JavaSc reproducibility reviewer discussion forum early and often

Do	Don't
Quick pre-repro-review checks and ask authors to fix before continuing; even if not all of these are technically required, authors who are willing to work reproducibly can show their engagement right from the start: <ol style="list-style-type: none">1. Do the links to data sets and materials resolve?2. Is there a README with clear step-by-step instructions?3. Is there a clear mention of to be expected execution times?4. Is there a LICENSE file to ensure openness?	Dig across badly or un-documented collections of files and functions to identify which part of the code/data creates which figure/table/output; find or build the “start button” yourself.
Encourage authors by pointing out promising intermediate results or concrete benefits of reproducibility.	Run workflows requiring considerable computational resources (unless interesting for you) but ask for data subsets for demonstration purposes.
Accept sample datasets to run a workflow and compare the outcome with the expected sample results; check the sources of the full datasets, if available.	Accept private sharing of data or code, unless strictly required for protection of sensitive data. All changes by the author should update to the public reproduction material.
Clearly document the extent of the reproduction in your reproduction report and suggest potential improvements; if you provide intermediate feedback, to include a history of your interactions in the report so that the ideas you contributed are preserved when the submission’s material is improved.	Attempt to install software without any instructions, install binary software of unknown origin, or try to fix installation problems you encounter on your machine; try to install without (a) asking for help from a fellow reproducibility reviewer who is familiar with the software, or (b) asking the author to help, providing a minimal reproducible example of your problem.
Get in touch with fellow reproducibility reviewers if specific expertise (tool, programming language, ...) is needed.	Point out or even fix problems that are not specific to the submission, e.g., general problems in a software tool.
Set an example when communicating about computational problems, e.g., by clearly defining your system (OS version, language version, etc.).	Create accounts on any service or platform to access code, data, or other resources.
Ask specific questions or point out concrete problems that may lead authors to improve their material, including referencing these guidelines or concrete tools/methods that you already (i) know about, especially if you suspect that the author might now be familiar with them (e.g., version pinning/dependency management, absolute paths).	Fix anything (unless you really enjoy doing so), e.g., <ul style="list-style-type: none">● compiler problems,● outdated libraries,● broken paths, or● Incomplete computing environment specifications, especially if the author can fix them even quicker.
Make sure that you are aware of any templates or specific resources provided for reproducibility reviewers from the reproducibility committee chair before starting your review.	
Consider the author’s background, career stage, and position to be aware of (a lack of) privileges or institutional power to decide how much support you provide and how you communicate; your reproducibility review can be a contribution to improve equity and inclusion in academia.	Be a bro .

Thanks!

Daniel Nüst

Frank Ostermann

Rusne Sileryte

Barbara Hofer

Carlos Granell

Marta Teperek

Anita Graser

Karl Broman

Kristina Hettne

Connie Clare

Frederique Belliard

Yan Wang



Review process

Proceedings:

https://www.agile-giscience-series.net/review_process.html

Process documentation:

<https://osf.io/7rjpe/>

Reproducibility review *after* accept/reject decisions

Reproducibility review & communication

Community conference & volunteers

Badges on proceedings website, article website with link, and first article page
(NEW! Thanks you, Copernicus!)

The screenshot displays the AGILE GIScience Series website. The main article is titled "Window Operators for Processing Spatio-Temporal Data Streams on Unmanned Vehicles" by Tobias Werner and Thomas Brinkhoff. The article includes an abstract, keywords, and a citation. A prominent badge indicates the article is "reproducible". The website also features a search bar, navigation menus, and a list of other articles in the series. The footer contains the AGILE GIScience Series logo, the Copernicus Publications logo, and a Creative Commons Attribution 4.0 International License notice.

Reproducibility Review at AGILE Conference 2021

Reproducibility review results

9 reproducibility reports published
(2020: 6)

8 not reproducible:

- 3 conceptual papers
- **data not shared (choice, licence)**
 - synthetic data! subsets!
- **code** not shared (choice) or proprietary software (repro reviewer matching failed)

Reproducibility review of: Building Change Detection of Airborne Laser Scanning and Dense Image Matching Point Clouds using Height and Class Information

Friese
Reproduction report and material.

Reproducibility review of: Investigating drivers' geospatial abilities in unfamiliar environments

Friese
Reproduction report and material.

Reproducibility review of: Extraction of linear structures from digital terrain models using deep learning

Nüst & Graser

Reproducibility review of: A Comparative Study of Typing and Speech For Map Metadata Creation

Ostermann & Nüst

Reproducibility review of: A Socially Aware Huff Model for Destination Choice in Nature-based Tourism

Krukar

Reproducibility review of: Automated Extraction of Labels from Large-Scale Historical Maps

Nüst

Reproducibility review of: Flood Impact Assessment on Road Network and Healthcare Access – at the example of Jakarta, Indonesia

Graser

Reproducibility review of: H-TFIDF: What makes areas specific over time in the massive flow of tweets related to the covid pandemic?

Nüst

Reproducibility review of: An Approach to Assess the Effect of Currentness of Spatial Data on Routing Quality

Nüst & Kmoch

Reproducibility Reports

Published on OSF with a DOI
Title page, cites the paper

Paper links to report via URL
(no citation)

Automatically added to ORCID profile

Eventually indexed in GS

Reproducibility review of: Investigating drivers' geospatial abilities in unfamiliar environments

Philipp A. Friese

2021-06-07



This report is part of the reproducibility review at the AGILE conference. For more information see <https://reproducible-agile.github.io/>. This document is published on OSF at <https://osf.io/dx92a>. To cite the report use

Friese, Philipp A. (2021, May). Reproducibility review of: Investigating drivers' geospatial abilities in unfamiliar environments. <https://doi.org/10.17605/OSF.IO/DX92A>

Reviewed paper

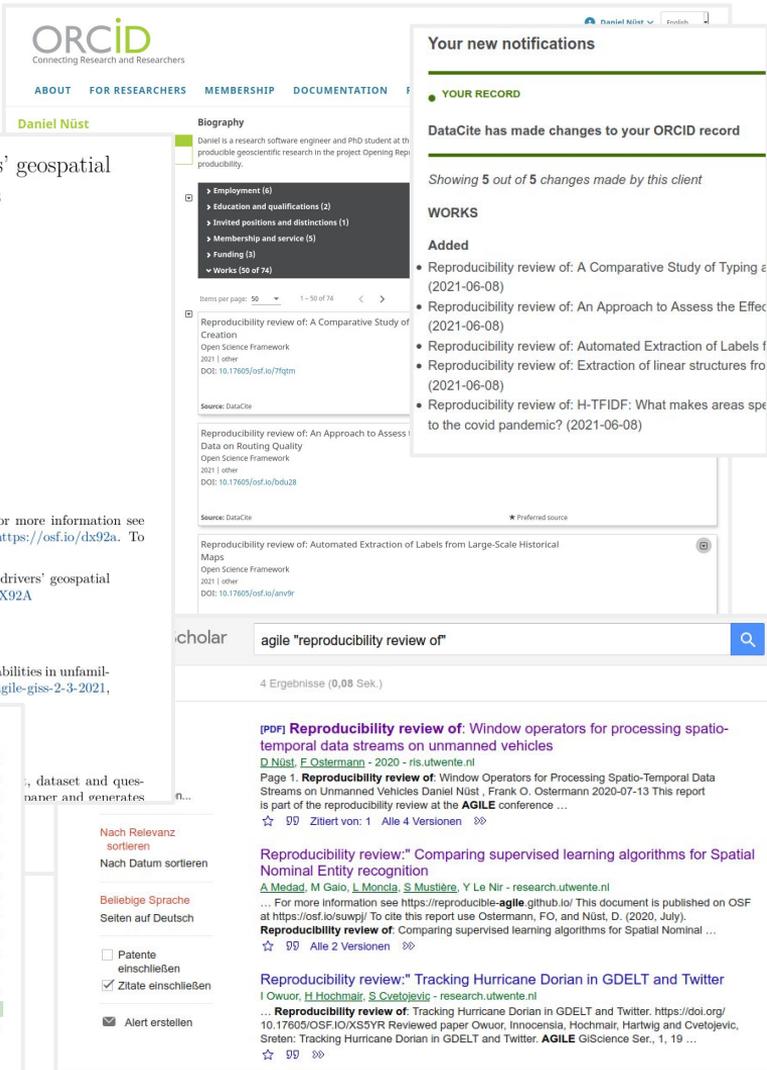
Karkasina, D., Kokla, M., and Tomai, E.: Investigating drivers' geospatial abilities in unfamiliar environments, *AGILE GIScience Ser.*, 2, 3, <https://doi.org/10.5194/agile-giss-2-3-2021>, 2021.

2.4 Data and Software Availability

Questionnaires and sketches were collected anonymously. All statistical analyses, which results are detailed in the following section, have been performed in R (R Core Team, 2021) using the tidyverse package (Wickham et al., 2019). Driving directions given to participants, an Exemplary Questionnaire in English, the collected survey data in tabular form, and all necessary metadata supporting this publication, are available on figshare and are accessible via the following DOI: <https://doi.org/10.6084/m9.figshare.14460102.v4>. The workflow underlying this paper was successfully reproduced by an independent reviewer during the AGILE reproducibility review and a reproducibility report was published at <https://doi.org/10.17605/OSF.IO/DX92A>.

Summary

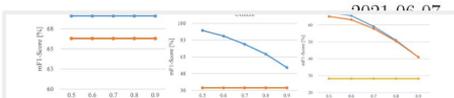
The updated submissionnaires. The provided dataset and questionnaire and generates...



The screenshot shows the ORCID profile of Daniel Nüst, including a biography, a list of works, and a search for 'agile reproducibility review of'. The search results show a paper by Daniel Nüst and Frank O. Ostermann, with a reproducibility review link. The ORCID profile also shows a notification about DataCite changes and a list of works, including the reproducibility review of the paper 'Automated Extraction of Labels from Large-Scale Historical Maps'.

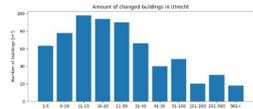
Reproducibility review of: Building Change Detection of Airborne Laser Scanning and Dense Image Matching Point Clouds using Height and Class Information

Philipp A. Friese 



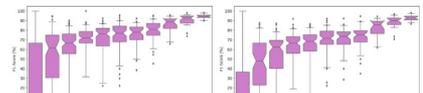
Graphic 1: Mean F1 Scores from Excel Sheet - corresponds to Figure 2-7 and methodologically to 12-15 in reproduced paper

After the authors provided an additional visualization script Figure 1 was reproduced. The generated image is shown in graphic 2.



Graphic 2: Generated Amounts of changed buildings - corresponds to Figure 1 in reproduced paper

The script generates Figure 10 and 11 automatically for each parameter permutation. An excerpt of the generated images are shown in graphic 3.



Reviewed paper

Politz, F., Sester, M., and Brenning and Dense Image Matching
GIScience Ser., 2, 10, <https://doi.org/10.17605/OSF.IO/RSF4M>



Graphic 4: Generated Prediction Images, visualised using QGIS, parameters 'jstl prob ct 0.7' - corresponds to Figure 8, 9, and 16 in reproduced paper

Reproducibility review of: Investigating drivers' geospatial abilities in unfamiliar environments

This report is part of the reproducible-agile.github.io cite the report use

Friese, Philipp A. (2021, May) abilities in unfamiliar environments

Reviewed paper

Karkasina, D., Kokla, M., and Iar environments, AGILE GIScience Ser., 2, 10, <https://doi.org/10.17605/OSF.IO/RSF4M>

Initial execution of the analysis script raised errors while generating Fig. 4. This was however resolved after contacting the authors and Fig. 4 was reproduced. The generated image is shown in Figure 1.

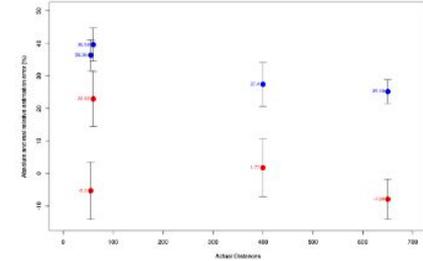


Figure 1: Average absolute (blue) and real relative (red) estimation error - corresponds to Figure 1 in reproduced paper

Table 1 and 2 were not generated automatically but instead the corresponding values were printed to the console. After extracting these values and exporting them to .csv files, both tables were reproduced. Refer to Table 1 and 2 in this report, which contain the raw values generated by the script.

Table 1: Spearman correlations Group 1 - corresponds to Table 1 in reproduced paper

type	S statistic	rho	p-value
SBSOD - Map Errors	463.24	-0.6197837	0.03160
SBSOD - Landmarks omitted	364.51	-0.2745132	0.38790
SBSOD - Road Segments mistakes	471.27	-0.6477894	0.02274
Landmarks omitted - Road Segments mistakes	160.86	0.4375473	0.15490
SBSOD - Direction estimates	233.45	0.1837559	0.56750
SBSOD - Distance estimates	342.00	-0.5545455	0.07065
Map Errors - Direction estimates	278.88	0.0249112	0.93570
Map Errors - Distance estimates	102.19	0.5354817	0.08958
Distance estimates - Direction estimates	205.81	0.0645223	0.85050

Table 2: Spearman correlations Group 2 - corresponds to Table 2 in reproduced paper

type	S statistic	rho	p-value
SBSOD - Map Errors	578.43	-0.5890887	0.03414
SBSOD - Landmarks omitted	520.67	-0.4304142	0.14210

Reproducibility review of: A Comparative Study of Typing and Speech For Map Metadata Creation

F.O. Ostermann  and Daniel Nüst 

Reproducibility reviewer notes

This review focuses on the reproduction of the analysis results. No in-depth examination code was conducted, but it was confirmed that the provided code can be run and see application used in the study. Using

```
# with npm version 6.14.8 and node version 14.13.0
npm install
npm start
```

we could run the application on <http://localhost:8080>, as shown in the screenshot below



Figure 1: Screenshot of application executed locally

A clear license is missing in the repository. The most important information (soft overview, exact questionnaire, maps used in the experiment) is also provided as supplement

This report is part of the reproducibility <https://reproducible-agile.github.io/>. To cite the report use

Ostermann, F. O., & Nüst, D. (2021). A Comparative Study of Typing and Speech For Map Metadata Creation, AGILE GIScience Ser., 2, 7, <https://doi.org/10.5194/agile-giss-2-7-2021>, 2021.

Reviewed paper

Lai, P.-C. and Degbelo, A.: A Comparative Study of Typing and Speech For Map Metadata Creation, AGILE GIScience Ser., 2, 7, <https://doi.org/10.5194/agile-giss-2-7-2021>, 2021.

Summary

The paper presents the results of a user experiment to improve GI-metadata using speech. A complete reproduction is practically impossible to achieve. This reproducibility report therefore investigated two components: First, whether sufficient information is provided to replicate the experiment elsewhere with

Reproducibility review of: A Socially Aware Huff Model for Destination Choice in Nature-based Tourism

Jakub Krukar 

2021.05.07

Construction.ipynb and can be verified by comparing the values from the output of chunk [12] to Table 8 in the paper. I expected that the values in the Number of photos column would stay the same (because the number of photos does not change), but the values in outgoing/incoming trips columns would change.

The location of the threshold variable is marked with an in-code comment in the file Trip Construction.ipynb:

```
if length.days > 4: #time threshold: average length of stay in both HPS
```

Table 8 Summary of attractions in Acadia National Park

Attraction	Number of photos	Outgoing trips	Incoming trips
Schoodic Institute	1119	53	64
Bass Harbor	2298	260	288
Southwest Harbor	723	109	111
Northeast Harbor	605	67	76
Bar Harbor	6259	433	357
Wild Gardens of Acadia	550	60	66
Cadillac Mountain	3285	349	345
Penobscot Peak	776	16	15
Bubble Rock	703	83	89
Jordan Pond	1250	227	250
Boulder Beach	536	85	102
Thunder Hole	977	167	185
Sand Beach	1253	216	177

Figure 1: Original Table 8 from the paper.

This report is part of the reproducibility <https://reproducible-agile.github.io/>, the report use

Krukar, J. (2021, May 7). Reproducibility in Nature-based Tourism. <https://doi.org/10.5194/agile-giss-2-7-2021>, 2021.

Reviewed paper

Shi, M., Janowicz, K., Cai, L., Mai, G. in Nature-based Tourism, AGILE GIScience Ser., 2, 7, <https://doi.org/10.5194/agile-giss-2-7-2021>, 2021.

Summary

The code, sample API query, and data working Binder link. All files contain

	a	b	c	d	e	f	g	h	i	j	k	l	m	total_out	total_in	cross_boundary	photos	
Places	0	13	7	1	12	1	8	0	4	3	2	6		57	66		123	1119
Schoodic Institute	12	0	34	9	64	13	53	4	8	25	12	15	21	268	295		563	2298
Bass Harbor	3	44	0	6	30	3	15	4	1	4	1	2	2	115	117		232	723
Southwest Harbor	5	16	8	0	13	1	7	0	2	10	1	2	3	68	78		146	605
Northeast Harbor	20	60	25	21	0	17	118	3	12	50	15	40	56	437	367		804	6259
Bar Harbor	1	3	1	2	10	0	6	1	1	6	4	11	15	61	67		128	550
Wild Gardens of Acadia	8	57	12	13	102	16	0	0	14	51	12	24	46	354	350		704	3285
Cadillac Mountain	2	3	3	2	2	0	0	0	0	2	0	1	1	16	15		31	776
Penobscot Peak																		

Reproducibility review of: Automated Extraction of Labels from Large-Scale Historical Maps

Daniel Nüst

2021-06-07

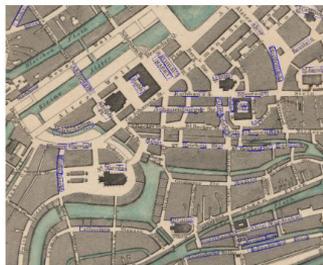


Figure 1: One output file from text recognition run.

This report is part of the reproducible-agile.github.io

Nüst, D. (2021, May 6). Reproducibility review of: Automated Extraction of Labels from Large-Scale Historical Maps.

Reviewed paper

Schlegel, I.: Automated Extraction of Labels from Large-Scale Historical Maps. *ISPRS International Journal of Geo-Information*, 2021, 10, 12, <https://doi.org/10.3390/ijgi10122021>.

Summary

The provided workflow could be partially reproduced. Some manual steps were included which were not documented in the repository.

I could run all cells in `String_Similarity_by_Levenshtein_Distance.ipynb`, and found the documentation to be extensive and a little bit raw, though very transparent, including tests by the author while developing the workflow etc. At first I got an error reading the `OCR_results.xlsx` file: `XLDRLError: Excel xlsx file; not supported`, so I changed the data loading to use `openpyxl`.

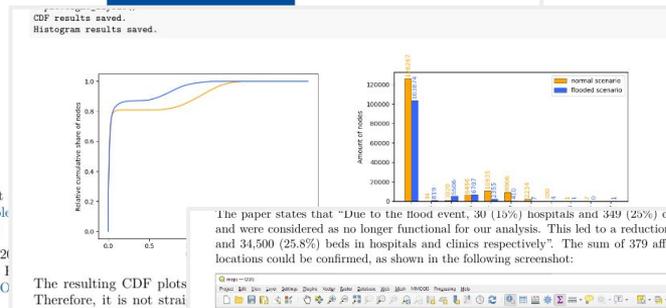
Figure 6 seems to be created using QGIS, but no project file or georeferenced version of the base map was included in the repository.

I did not run the final notebook of step "06 Approximate georeferencing", because of the advertised run error. I found out that the advertised workflow is not reproducible in my QGIS installation and that the error message is not clear. I tried to run the workflow in my QGIS installation and found that the error message is not clear. I tried to run the workflow in my QGIS installation and found that the error message is not clear.

Reproducibility review of: Flood Impact Assessment on Road Network and Healthcare Access – at the example of Jakarta, Indonesia

Anita Graser

2021-06-07

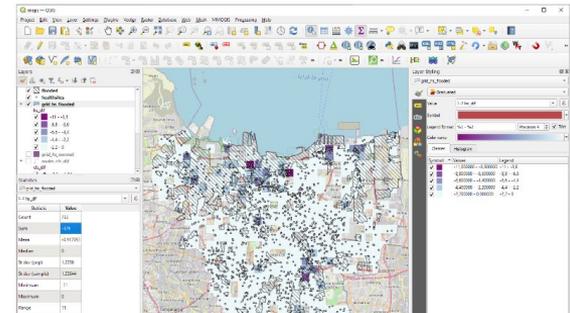


This report is part of the reproducible-agile.github.io

Graser, A. (2021, May 6). Reproducibility review of: Flood Impact Assessment on Road Network and Healthcare Access – at the example of Jakarta, Indonesia.

The resulting CDF plots are not stratified. Therefore, it is not stratified.

The paper states that "Due to the flood event, 30 (15%) hospitals and 349 (25%) clinics were affected and were considered as no longer functional for our analysis. This led to a reduction of 12,000 (16.6%) and 34,500 (25.8%) beds in hospitals and clinics respectively". The sum of 379 affected health service locations could be confirmed, as shown in the following screenshot:



Reviewed paper

Klipper, I. G., Zipf, A., and Lautenbach, S.: Flood Impact Assessment on Road Network and Healthcare Access at the Example of Jakarta, Indonesia. *AGILE 2021*, 2021, 1, 1, https://doi.org/10.1007/978-3-030-58000-0_1.

Summary

The provided workflow was partially reproducible. Some manual steps were included which were not documented in the repository.

Reproducibility review of: H-TFIDF: What makes areas specific over time in the massive flow of tweets related to the covid pandemic?

Daniel Nüst 

2021-06-07



outbreak
coronavirus
people
lockdown
news
viral
disneyland
cases
likely
committing
line
new
health
lockdown
deadly
news
viral
disneyland
cases

Figure 4: Reproduction of Figure 3 (only one of two weeks). Wordcloud of tweets-mood-tets/experiments/agile21/results/jan_2weeks_week/country 01-19.png.



Figure 7: Reproduction of Figure 6a. Projection of H-TFIDF representation in a t-SNE space; file covid19-tweets-mood-tets/experiments/agile21/results/jan_2weeks_week/tsne_bert_embeddings_H-TFIDF.png

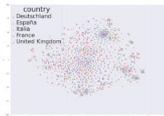


Figure 8: Reproduction of Figure 6b. Projection of TF-IDF representation in a t-SNE space; file covid19-tweets-mood-tets/experiments/agile21/results/jan_2weeks_week/tsne_bert_embeddings_TF-IDF on corpus by Country.png

This report
<https://reproducible-agile.github.io/>
cite the report

Nüst, D., over 17605

Review

Decomposition of the massive flow of tweets related to the covid pandemic over time in the massive flow of tweets related to the covid pandemic. *AGIScience Ser.*, 2, 2, <https://doi.org/10.5194/agile-giss-2>

Summary

The authors provide a well documented workflow analysing a large time span. Because of the data size, the authors provided instructions on how to execute the code successfully and the created figures match the original report.

Reproducibility review of: An Approach to Assess the Effect of Currentness of Spatial Data on Routing Quality

Alexander Knoch , Daniel Nüst 

2021-06-07

REPRODUCIBLE
AGILE

This report is part of the reproducibility review at <https://reproducible-agile.github.io/>. This document is a reproduction of the original report. Please cite the report as:

Nüst, D., & Knoch, A. (2021, May 19). Reproducibility Review of: Effect of Currentness of Spatial Data on Routing Quality, *AGIScience Ser.*, 2, 2, <https://doi.org/10.5194/agile-giss-2-13-2021>, 2021.

Reviewed paper

Schmidl, M., Navratil, G., and Giannopoulos, I.: Currentness of Spatial Data on Routing Quality, *AGIScience Ser.*, 2, 2, <https://doi.org/10.5194/agile-giss-2-13-2021>, 2021.

Summary

The reproduction was successful. All provided scripts could be executed using the provided data. Some manual steps could not be automated, but the results were consistent with the original report.

```
signal: TRUE  
snprievr::mapshot(n, file = "figure1.png")  
knitr::include_graphics("figure1.png")
```



Figure 1: Reproduction of Figure 1: 'Distribution of the 1000 origin and destination points used in the experiment'

Table 1

A version of Table 1, naturally with different values, could be recreated from any of the generated GeoJSON files:

```
knitr::kable(sf::read_sf("001_fatmatst_geojson"),  
caption = "Reproduction of Table 1")
```

Table 1: Reproduction of Table 1

year	duration	distance	geometry
2014	1173.1	11093.7	LINSTRING (16.29116 48.166...
2015	1195.8	11180.5	LINSTRING (16.29116 48.166...
2016	1176.9	11508.9	LINSTRING (16.29116 48.166...
2017	1175.7	11511.4	LINSTRING (16.29116 48.166...
2018	1174.9	11511.6	LINSTRING (16.29116 48.166...
2019	1181.8	11510.5	LINSTRING (16.29116 48.166...
2020	1183.0	11513.0	LINSTRING (16.29116 48.166...

Check route completeness

Reproducibility review reports AGILE 2021



Reproducibility review of: Extraction of linear structures from digital terrain models using deep learning

Daniel Nüst and Anita Graser

2021-06-07



This report is part of the reproducibility review at the AGILE conference. For more info <https://reproducible-agile.github.io/>. This document is published on OSF at <https://osf.io/2sc7g> cite the report use

Nüst, D., & Graser, A. (2021, April 30). Reproducibility review of: Extraction of structures from digital terrain models using deep learning. <https://doi.org/10.17605/2sc7g>

Reviewed paper

Satari, R., Kazimi, B., and Sester, M.: Extraction of linear structures from digital terrain models using deep learning, AGILE GIScience Ser., 2, 11, <https://doi.org/10.5194/ag-2-11-2021>, 2021.

Summary

The provided workflow was **partially reproduced**. Based on the provided test file and it was able to recreate the computing environment and run the segmentation models. Relevant the paper could be recreated. The training and validation part of the workflow is irreproducible.

This finished within a minute! These values match the column `SegNet` of Table 1, within a level of precision to be expected from such a classification.

```
hrnet <- read.csv("agile-submission-2021/HRNetBinarySegmentation/files/evaluation_file.csv")
segnet <- read.csv("agile-submission-2021/SegnetBinarySegmentation/files/evaluation_file.csv")

suppressPackageStartupMessages(library("tidyverse"))
dplyr::full_join(hrnet, segnet) %>%
  knitr::kable()
```

	acc	f1_m	loss	model_type	precision_m	recall_m	specific_name
	0.9069678	0.7762662	0.2579104	hrnet	0.8187426	0.7393778	simple_binary
	0.8787796	0.6977967	0.3094499	segnetCustomized	0.7688990	0.6399856	simple_binary

Run the next segmentation:

```
cd multiClassSegmentation
python3 evaluate.py --evaluation_file=evaluation_file.csv
```

This completes and recreates the data in Table 3 within reasonable numerical precision based on the file `multiClassEvaluation.csv`. It is unclear to me how Table 2 can be constructed from `evaluation_file.csv` of this segmentation, but I assume it can be.

```
multi <- read.csv("agile-submission-2021/multiClassSegmentation/files/multiClassEvaluation.csv")

rows <- lapply(c(0:5), function(class) {
  classValues <- multi %>%
    dplyr::select(dplyr::ends_with(as.character(class)))
  names(classValues) <- c("sparse_iou", "prediction", "recall", "f1.score", "support")
  c("Class label" = as.character(class), classValues)
})

dplyr::bind_rows(rows) %>%
  knitr::kable()
```

Class label	sparse_iou	prediction	recall	f1.score	support
0	0.8952831	0.9270560	0.9630006	0.9446866	12342971
1	0.2073642	0.4009994	0.3745212	0.3873083	247492
2	0.0324831	0.3139039	0.0373107	0.0666941	282037
3	0.7648531	0.8797419	0.8552982	0.8673479	1961849
4	0.4453632	0.6953347	0.5580455	0.6191711	830866
5	0.1593569	0.2697686	0.2787574	0.2741893	145345

General observations and lessons learned

- **Further improvement over last years submissions - better prepared workflows!** Biggest hurdles remain: insufficient documentation, no “quick” variant or lack of expected data size/runtime, links Figures < > Scripts
- **Community understanding better, but needs time:** Had to remind authors to add DASA section - how can we be clearer in the communication? Camera-ready papers by authors possible, but exhausting.
- Additional **reproducibility questions for scientific reviewers worked better**, but triggering only by regular reviewers doesn't work well - fortunately not too many submission to check for repro chair
- Repro reviews **were less strict than original ideal but on par with last year**
> promote positive examples and don't expect perfection
- **Non-blindness** served its purpose, but unblinding also delayed procedures
- **Schedule** still very much a challenge, partly because infrastructure (EasyChair) does not enable reviewer roles and communication > working around that with scripts and scraping
- **Improvements to process were good:** clarity in communication for authors that **DASA section is mandatory**, not attempting short papers, do not offer authors to object to report publications (no problems!)
- **Reproduction not attempted != bad science**, reproducibility is not binary but a spectrum
> continue education on reproducibility, increase requirements while practices spread in community

The guidelines for reproducibility reviewers

Ideal vs. realistic

Role

Skills

Do's & dont's

REPRODUCIBILITY REVIEWER GUIDELINES

Reproducibility reviewers conduct a complimentary review of the workflow that is published with a manuscript. Ideally, reproducibility reviewers only **read the abstract and the Data and Software Availability section** (DASA) of an article. They may read other sections referenced in the latter. Then they follow the authors' instructions for executing the workflow, ideally starting from the DASA or a README file in the referenced reproduction material. When reproducibility reviewers get stuck, they take advantage of the option to **communicate** with the authors early and often. Reproducibility reviewers should be aware of the different levels for making research reproducible in the author guidelines (see above) to be able to **recommend improvements** to the author and at the same time have the skillset and tools to conduct their review efficiently. Reproducibility reviewers are not responsible for making a workflow transparent or executable. Reproducibility reviewers **write a short reproducibility report** documenting their communication and the results of their reproduction attempt. The report is published if the reproduction was, at least in part, successful.

The reproducibility review from a reproducibility reviewer's perspective

While these AGILE reproducible paper guidelines are created with an intention to eventually have 100% of computations of accepted submissions succeed, understanding, and ultimately community adoption, these tasks are harder and progress slower yet hopefully accepted article, but a successful reproduction should be aware of this role and accept that it is a "partial reproduction", she should accept it. The current discussion is one reproducibility reviewer is assigned to a reproduction and the scientific reviewer on the same

The scope of the reproducibility review is rough, but the community is worth taking a look at for further review. A *partial reproduction*, i.e. if you can't reproduce it, is seen as a success at this point, though what you can do is to ask fellow reproducibility editors or the reproducibility

Reproducibility reviewer skills

A reproducibility review ideally is a learning experience for the AGILE community to increase openness and transparency. A concrete amount of time you spend on a reproduction piece of research you are tasked to reproduce is an hour to get a workflow started. Although you may not be good enough for anyone to reproduce a workflow, you can get things to work within minutes (no counting) by using good package managers and getting familiar with DESCRIPTION files and renv for R, npm for JavaScript, etc.

Do	Don't
Quick pre-repro-review checks and ask authors to fix before continuing; even if not all of these are technically required, authors who are willing to work reproducibly can show their engagement right from the start: <ol style="list-style-type: none">1. Do the links to data sets and materials resolve?2. Is there a README with clear step-by-step instructions?3. Is there a clear mention of to be expected execution times?4. Is there a LICENSE file to ensure openness?	Dig across badly or un-documented collections of files and functions to identify which part of the code/data creates which figure/table/output; find or build the "start button" yourself.
Encourage authors by pointing out promising intermediate results or concrete benefits of reproducibility.	Run workflows requiring considerable computational resources (unless interesting for you) but ask for data subsets for demonstration purposes.
Accept sample datasets to run a workflow and compare the outcome with the expected sample results; check the sources of the full datasets, if available.	Accept private sharing of data or code, unless strictly required for protection of sensitive data. All changes by the author should update to the public reproduction material.
Clearly document the extent of the reproduction in your reproduction report and suggest potential improvements; if you provide intermediate feedback, to include a history of your interactions in the report so that the ideas you contributed are preserved when the submission's material is improved.	Attempt to install software without any instructions, install binary software of unknown origin, or try to fix installation problems you encounter on your machine; try to install without (a) asking for help from a fellow reproducibility reviewer who is familiar with the software, or (b) asking the author to help, providing a minimal reproducible example of your problem.
Get in touch with fellow reproducibility reviewers if specific expertise (tool, programming language, ...) is needed.	Point out or even fix problems that are not specific to the submission, e.g., general problems in a software tool.
Set an example when communicating about computational problems, e.g., by clearly defining your system (OS version, language version, etc.)	Create accounts on any service or platform to access code, data, or other resources.



How to put your community on a path towards more reproducibility in 5 easy hard steps

1. Build a team of enthusiasts (workshop, social events)
2. Assess the current state and raise awareness (workshop, paper)
3. Institutional support (🙏 AGILE Council 🙏 + committee chairs)
4. Positive encouragement (no reproduction != bad science)
5. Keep at it!

Next steps



Do it again in 2022 🎉

🔧 Revise guidelines? 🇮🇹 🇫🇷 🇨🇳

Grow reproducibility reviewer team

YOU!, opportunity ECRs
(mentoring/workshops/...)

Continue meta-research 🧐

Ostermann, F., Nüst, D., Granell, C., Hofer, B., & Konkol, M. (2020).
*Reproducible Research and GIScience: an evaluation using
GIScience conference papers.*
EarthArXiv. <https://doi.org/10.31223/x5zk5v>

**Continue community engagement
towards opening scholarship**

Scope

Requirements

Acceptance condition?

Open review if tenured

Format-free first submission

CRedit

Phase out when standard practice...

