Towards Transparency and Knowledge Exchange in AI-assisted Data Analysis Code Generation

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The integration of Large Language Models (LLMs) in scientific research presents both opportunities and challenges for life scientists. Key challenges include ensuring transparency in AIgenerated content and facilitating efficient knowledge exchange among researchers. These issues arise from the in-transparent nature of AI-driven code generation and the informal sharing of AI insights, which may hinder reproducibility and collaboration. This paper introduces git-bob, an innovative AI-assistant designed to address these challenges by fostering an interactive and transparent collaboration platform within GitHub. By enabling seamless dialogue between humans and AI, git-bob ensures that AI contributions are transparent and verifiable. Moreover, it supports collaborative knowledge exchange, enhancing the interdisciplinary dialogue necessary for cuttingedge life sciences research. The open-source nature of git-bob further promotes accessibility and customization, positioning it as a vital tool in employing LLMs responsibly and effectively within scientific communities.

data analysis, bio-image analysis, LLM, code-generation
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Introduction

Generative artificial intelligence (AI) and Large Language Models (LLMs) in particular are changing the way we do data science. Most prominently, scientists use the technology for interacting with scientific data (1), answer data analysis questions (2, 3), generate data analysis code (4–6), and [re-]writing scientific manuscripts (7). Unfortunately, the prompts sent to LLMs are commonly not conserved, and thus, at the time of publication scientific work, it might be hard differentiate human-made and AI-generated parts of the project. A professional peer-review system, for documenting how LLM-generated code was prompted for, and which human reviewed it, is not established in contemporary scientific culture. However, such systems do exist for collaborative code editing involving multiple humans. E.g. the github.com online platform is well-established in the opensource software community for discussing issues and potential solutions, building code together, and for peer-reviewing contents. As it was shown before that LLMs can solve realworld GitHub issues (8), developing an AI-assistant that interacts with humans directly in the well-established Github platform is the obvious next step. I am presenting git-bob, a functional proof-of-concept implementation of an LLMbased AI-assistant that can respond to GitHub issues, discuss potential solutions with humans iteratively, write code for them, and submit it as pull-request to be reviewed by humans. It is technically similar to various online services for data analysis such as the OpenAI ChatGPT Data Analyst or Github Copilot Workflows, with three major differences: 1) Multiple humans can interact with git-bob in one communication thread. This allows bringing together domain specialists, e.g. a life scientist, data-analyst and the AI-assistant in one discussion, stimulating knowledge exchange on how to interact with the AI-assistant. 2) Discussions with git-bob and resulting code-modifications are conserved in an online-platform that others can read and follow, making the interaction with the AI-assistant fully transparent, and 3) git-bob is open-source. Other developers can read its built-in system prompts and modify them to their needs. Git-bobs source code is available online: https: //github.com/haesleinhuepf/git-bob.

Features and limitations

A common workflow involving git-bob is demonstrated in Figure 1: a user opens an issue, a kind of discussion thread, on a repository on github.com, where git-bob is installed. A repository member can then trigger git-bob to answer by writing a command such as "git-bob comment on this". If externals try so, an automatic response will inform them that only repository members are allowed to trigger git-bob because running git-bob may cause costs for repository owners. Once triggered, git-bob will respond to the question, potentially including a code snippet and resulting plots or images. Users and the AI-assistant can then discuss back and forth until some potential solution is reached. Optionally, git-bob can then be asked to implement the solution and send a GitHub pull-request, another kind of discussion thread, but accompanied by file modifications to the repository, e.g. including a Jupyter Notebook containing the previously discussed code solution to a given issue. A human would need to review this pull-request and merge it into the code base of the repository. Git-bob also has the capability to review pull-requests, e.g. originating from humans, but it is not allowed to merge them. This reflects established practices in science, where eventually a scientist is responsible for data analysis code that becomes part of the project. Additional tasks git-bob is capable of are: 1) The assistant can support users of open source libraries by providing advice and code examples, as shown

in Supplementary Figure S1. In case the assistant is not sure about the answer, it is capable of forwarding the question to a human (Supplementary Figure S2). 2) It can be used to document code (Supplementary Figure S4). Such a task can be time-consuming when performed without AI-assistance, which can generate documentation for multiple Python functions in seconds to minutes. 3) it can analysis data in the repository directly, e.g. summarize and plot data in CSV files (Supplementary Figure S3). 4) If manuscript files are stored in a github repository, e.g. in latex format, git-bob can assist in writing. For example, the abstract for this manuscript was written by the AI-assistant and this is documented transparently as shown in Supplementary Figure S5.

When writing data analysis code, git-bob is intrinsically limited by the capabilities of the used LLM. For example, it has been shown that state-of-the-art (SOTA) LLMs can solve bioimage analysis questions by generating functionally correct code just above 50% of tested cases (5). This fundamental limitation may disappear when improved LLMs are published. For now, it can be evaded by multi-turn interactions between humans and AI-assistant. Further technical limitations are file size of files to be read and written, as prompt input and output lengths are limited to multiple thousand words for SOTA LLMs. Also when processing data, technical limitations of the GitHub IT infrastructure have to be considered. git-bob executed in public repositories is per default executed on virtual machines with 4 CPU cores, 16 GB of RAM and 14 GB of SSD storage. In private repositories, only 2 CPU cores and 7 GB RAM are available (9). More capable systems are available on a paid basis.

A highlight of git-bob is that a local installation are not required. Git-bob is implemented as GitHub workflow, which can be installed by uploading a configuration file to a repository and setting access rights. It is compatible and was tested with the commercial models OpenAI's GPT4-omni, Anthropic's Claude, Google Gemini and freely available models hosted on Github Models Marketplace. Git-bob reports which model was used in all of its messages, as good scientific practice suggests. Obviously, the communication with the selected LLM is transmitted to the service provider, including source code files from the repository and images provided with the github issue. Hence, users are recommended to not submit any personal or sensitive information.

Conclusion

LLMs are being integrated in scientific workflows unavoidably, but commonly it is not documented how they were employed, also because of lack of tools conserving this information conveniently. If the scientific community documented how the prompted LLMs like they document how they used data analysis software, we could learn from each other how to prompt efficiently and responsibly. To overcome current limitations, I propose git-bob, a functional, LLM-based proof-of-concept AI-assistant embedded in the GitGub platform. It enables scientists to interact with an LLM via Github Issues and Pull-Requests offering new ways for implementing good scientific practice for the conservation of discussions

between humans and AI when they are working on projects together.

ACKNOWLEDGEMENTS

I would like to thank Elena Katharina Nicolay (UFZ Leipzig) for testing git-bob in its early days and for providing constructive feedback on the manuscript. I acknowledge the financial support by the Federal Ministry of Education and Research of Germany and by Sächsische Staatsministerium für Wissenschaft, Kultur und Tourismus in the programme Center of Excellence for Al-research "Center for Scalable Data Analytics and Artificial Intelligence Dresden/Leipzig", project identification number: ScaDS.Al.

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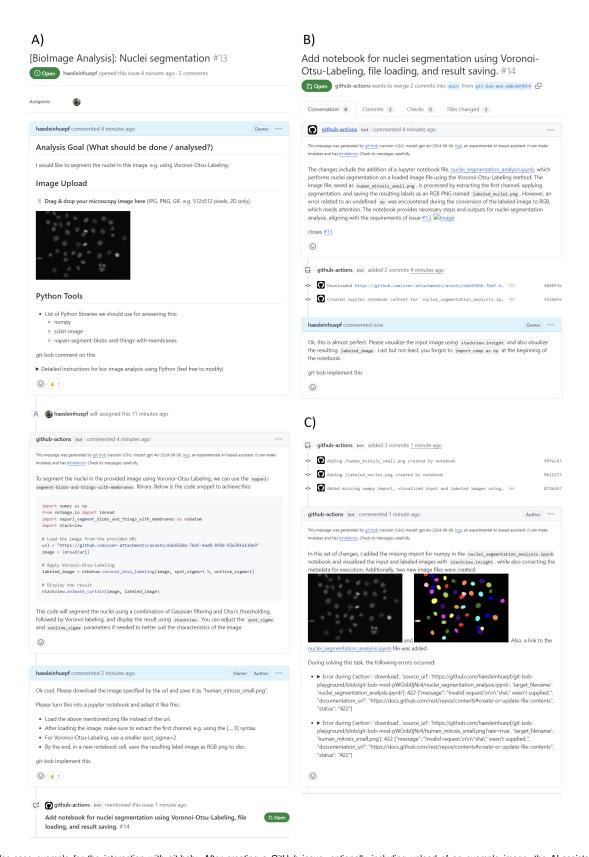


Fig. 1. Use-case example for the interaction with git-bob: After creating a GitHub issue, optionally including upload of an example image, the Al-assistant and the human exchange about how to process the data (A). Once they concluded to implement the solution, a GitHub pull-request is sent (B) containing the programmed solution as files in the repository. Also here, human and Al-assistant can continue discussing and modify the solution until the example is satisfactory (C). The discussion also contains error messages observed while the Al-assistant was executing code or creating files, allowing to modify prompts or improve git-bob itself. The entire discussion and corresponding code can be read online: https://github.com/haesleinhuepf/git-bob-playground/issues/13 and https://github.com/haesleinhuepf/git-bob-playground/pull/14

Supplementary material

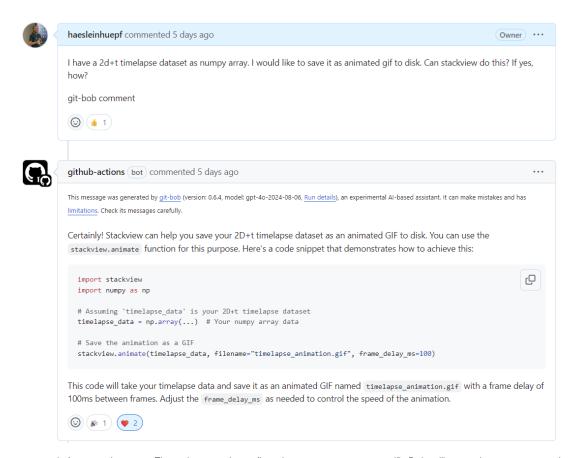


Fig. S1. Use-case example for supporting users: The assistant can be configured to act as expert on a specific Python library and answer user questions. The entire discussion and corresponding code can be read online: https://github.com/haesleinhuepf/stackview/issues/79

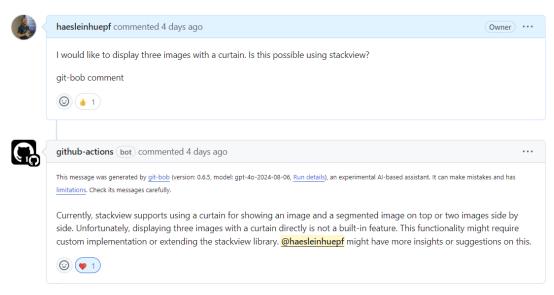


Fig. S2. Use-case example for asking an expert: The answer to the question shown here is "No", but this is nowhere written in the documentation or the configuration of the assistant. In this case the assistant is not sure, and it can be configured to forward a question to a maintainer of the library where the question arrived. The entire discussion and corresponding code can be read online: https://github.com/haesleinhuepf/stackview/issues/80

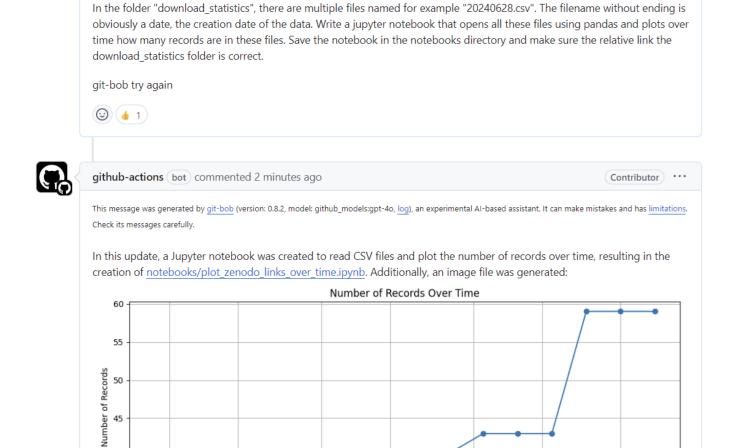


Fig. S3. Use-case example for plotting data: after explaining the assistant the folder structure of the project, it generates code for parsing a folder of CSV files and plotting $results. \ The \ entire \ discussion \ and \ corresponding \ code \ can \ be \ read \ online: \ https://github.com/NFDI4BIOIMAGE/training/issues/250$

Date

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haesleinhuepf commented 3 minutes ago

Member

A)

Add numpy-style docstrings to _github_utilities.py #29

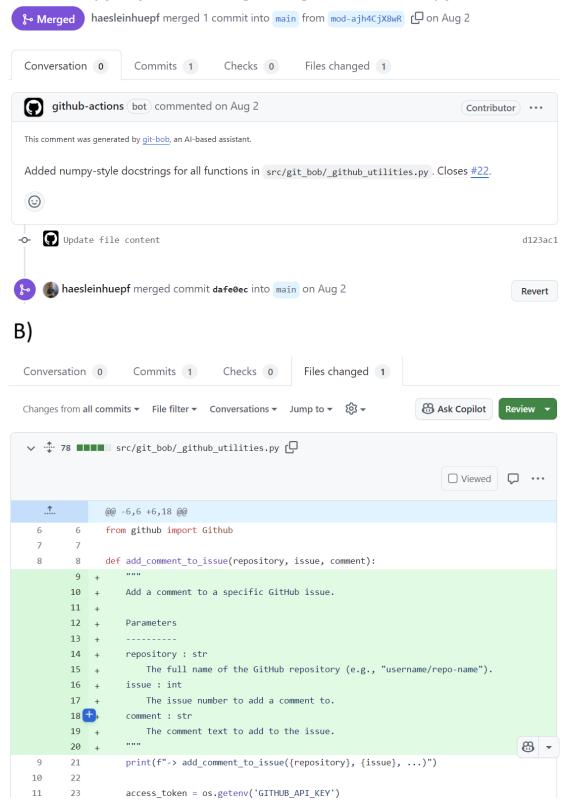
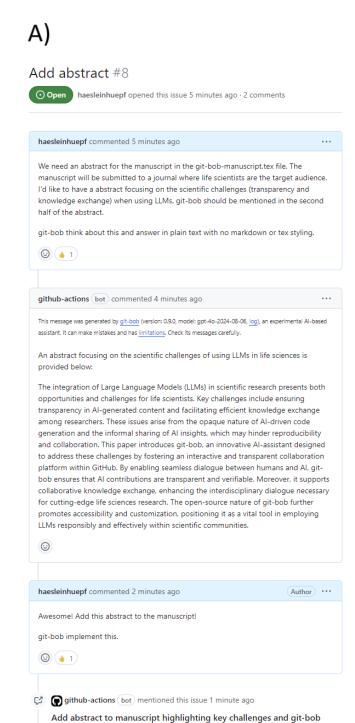


Fig. S4. Use-case example for adding and revising documentation in code: git-bob was used to partially write the code documentation of its own code. When asked to add documentation in a specific format, it sent a pull-request (A) and the human could inspect the code modifications (B, excerpt) before mergin the code into the project's code base. The entire discussion and corresponding code can be read online: https://github.com/haesleinhuepf/git-bob/pull/29



B)

Add abstract to manuscript highlighting key challenges and git-bob solution. #9



Fig. S5. Use-case example for working on scientific manuscripts: after a first draft of the manuscript was written, git-bob was asked to formulate an abstract (A). The abstract was then submitted as pull-request with a short description (B). The human can also review and potentially modify the proposed text in this online interface (C). The entire discussion can be read online: https://github.com/haesleinhuepf/git-bob-manuscript/issues/8 and https://github.com/haesleinhuepf/git-bob-manuscript/pull/9.

solution, #9

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