

Code Review For and By Scientists

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Abstract

We present results from two pilot studies of code review by and for scientists. Our principal findings are that scientists are enthusiastic, but need to be shown the practice in action, and that just-in-time review of small code changes is more likely to succeed than large-scale end-of-work reviews.

1 Introduction

Since Fagan’s work in the 1970s [1, 2], dozens of studies have shown that code review is the most effective way to find bugs [3, 4]. Ironically, given that Fagan and others were inspired by academic peer review, code review is still rare in scientific software development. This is partly because most authors don’t publish their code, and hence have little incentive to improve its quality, but also a case of the blind (not) leading the blind.

In 2013–14, we ran two pilot studies to explore the benefits of code review for typical scientist-coders, and how to transfer the skill to them. Our focus was not the small minority of scientists who already use good software engineering practices [5], but on the large majority who have never encountered them.

2 Post-Hoc Reviews

Our first study, done in conjunction with the Public Library of Science (PLOS), ran from August to October 2013. Professional developers working at Mozilla reviewed samples of code taken from papers published in *PLOS Computational Biology* in the preceding 18 months, and their reviews were shared with the scientists. We then conducted semi-structured interviews [6, 7] with 11 developers in late August and early September, and with 4 authors whose code had been reviewed in late September and early October, to determine whether non-scientists could usefully review

typical scientific software, whether those reviews were intelligible and useful to the scientists, and whether the participants felt the reviews were valuable. Four significant findings emerged.

1) *Scientists’ starting points varied widely.* Scientists’ prior knowledge and practices varied widely. Some used engineering practices and tools such as version control some of the time, but documentation of the form developers expected was rare, and practices like continuous integration were unknown or considered “very unusual”.

In particular, most scientists had never taken part in code review. Some of the code authors we interviewed worked in (or lead) teams in which they read each other’s code, but most participants had never, or only rarely, discussed their code with others. Instead, they discussed the results the code produced.

2) *Developers felt limited.* Most developers repeatedly noted the lack of documentation, commenting, tests, and example data sets, e.g., “Not having the project build is a big problem; I can’t verify that the code is correct.” Equally, most reviewers were frustrated at not being able to run the code as the first step in review, e.g., “That’s the easiest way to see if it works at all,” and “It’s a way to validate the intentions of the author.” Several commented that, as a result, they had to make some assumptions in the review.

Few reviewers read in detail the paper with which the code was associated, usually because they lacked the domain expertise to do so. They could deduce what the code did and assess it in terms of its ability to perform that operation efficiently, but they could not tell if the code fulfilled its intended scientific role. That, combined with their isolation from the scientists, left developers feeling that they were doing “drive-by” reviews.

3) *Standards and expectations were very different.* Many reviewers were struck by how scientists’ code differed from theirs, particularly in lacking commenting and explanation in the code. Several suggested

that scientists appear to have “less concern for maintainability and readability” and that the code was “not written for others to use”. Some also pointed out naive lack of complexity or abstraction, redundancy in the code, inconsistencies in or ignorance of standards in formatting, and unhelpful naming.

On their side, the scientists all aspired to create readable and re-usable code, but many noted that their code doesn’t respect the common etiquette of open source: for example, they often don’t make an effort to package their code for re-use by others. One scientist commented that: “[It’s] not important to have something that’s exact—only when we publish” and reiterated the low status of code in their science: “In the business of science, all that matters is the figures. The quality of the code is just not on the critical path.” This echoes findings from other empirical research [8, 9].

4) *Both sides wanted more contact.* All but one of the reviewers remarked that they miss the social context they normally associate with code reviews: working relationships, understanding goals and priorities, and trust. They would have preferred some form of dialog, not least to provide them with the code context, and to establish appropriate expectations for both reviewers and authors (e.g., “knowing what kind of feedback the author wants”). A typical remark was, “It would have been easier if we were allowed to contact the scientist just to get a feel for his mindset.” Most referred to dialogs that are normally part of their code review experience, whether spoken or online: “Discussion catches details that get lost otherwise” such as “tiny changes in numerical interpretation that are important”. The reviewers also pointed out that the dialog has value for the reviewer, as well as for the author: “When there is a dialog, you end up learning a lot yourself.”

3 Transferring the Skill

Despite these frustrations, participants in this study were positive about it. We therefore set up a second study that paired experienced scientific programmers with small groups of less-experienced ones to explore ways of transferring the practice itself. This second study also aimed to address the fact that scientists and developers alike wanted reviews during development, when they could act on the feedback, rather than a “drive-by” review after submission.

Ten groups ranging in size from a couple of people to half a dozen initially signed up to take part in this study. In interviews and early discussions, they identified four main reasons why they think they ought

to do code review:

1) *Rigor:* Scientists want to get the right answer and be able to reproduce their work later. (Most recognized that correctness and reproducibility aren’t the same thing.)

2) *Reusability:* Scientists are very aware that their understanding of code dissipates over time and that this is a large hidden cost. Equally, they suspect that they spend a lot of time reinventing wheels. They may not know how code review will help with that, but they hope that it will.

3) *Collaboration:* Many scientists hoped that they could use code review as an excuse for conversations about code with their colleagues that simply don’t happen right now. Some believed that review would foster better testing, encourage scientists to produce code that is easier to understand and share, and make team members more aware of each other’s research.

4) *Knowledge transfer:* All participants felt that there must be a better way to build programs than what they’re doing right now. Taking part in this study was, for them, a way to get mentoring about programming in general from someone who could answer questions they didn’t know to ask.

One thing we *didn’t* hear was people saying that they wanted to learn about the science embodied in the code being reviewed. Rightly or wrongly, scientists seem to feel that “what it does” can be learned in others ways.

Each team was paired with a mentor: another scientist-developer with experience of code review, from a different institution but usually in the same or a cognate discipline, who would introduce the team to code review processes and good practices. Teams and mentors collaborated during March–June 2014. We followed teams’ interactions on email (via shared mailing lists) and within their repositories, attended meetings when possible (using video or telephone conferencing), and interviewed participants periodically through the study about their goals, experiences, and impressions.

Our biggest finding is how much *perspective matters to collaboration*. Over time, professional software developers (and scientists whose primary role has become programming) integrate skills and knowledge, building up a standard vocabulary of terms, tools, and conventions, as well as a practical repertoire that it’s easy to take for granted.

Most scientists don’t (yet) have such an overview: they are focused almost entirely on the purpose the code serves rather than the code itself. They often lack the integrated understanding and skill that comes with experience, which meant it was often very useful for mentors to “state the obvious”.

How to make the transition from the first perspective to the second is usually not clear to working scientists. “Stuff that seemed like overkill makes sense now... A lightbulb finally went off: I understand how other people look at my code and how to make it work.” And what basics to articulate—where to look and how to look—is often not clear to experienced developers. If the two are going to collaborate, then at some point the different perspectives must be acknowledged and bridged. Social mechanisms like courtesy, deference, and avoiding embarrassment tend to obscure this, so it’s crucial to articulate collaborators’ perspectives explicitly for both sides, and equally to help establish appropriate expectations explicitly.

This observation isn’t new: as Segal discussed in her study of industrial software engineers collaborating with scientists [8], “[Programmers] demand an upfront articulation of requirements, whereas the scientists had experience, and hence expectations, of emergent requirements. The project documentation does not suffice to construct a shared understanding.”

For every mentor who says, “I wish I knew more clearly what the scientists are expecting,” there’s a scientist who says, “I wish I knew what it’s reasonable to expect.” One concrete step we could take is to show scientists what it actually looks like to do a code review: not just the end result, but the process itself.

The most successful groups both saw code review done and did it. Their mentors walked through an initial code review; articulated what they look for in the code, what they see, and how they interpret it; and made concrete suggestions about first steps that clarified scientists’ expectations. But scientists perceived the real benefits only after they undertook the process themselves, engaging in code review for each other, and discussing those reviews with others.

Another observation was that *there are several models of peer review*, each with its own emphasis. In *asynchronous review*, which is the predominant model in open-source development, comments and revisions are mediated by the version control system. *Offline review with synchronous presentation* uses the tools available in the version control system to present the code and supporting material for review, and to capture the reviewers comments, but the reviewer and development team meet (usually virtually) to talk through the code and the review in order to facilitate a freer interaction, especially in terms of questions and clarifications. And finally, *synchronous review* involves walking and talking through the code together, whether one-to-one, as a group, or with one person leading.

There are other axes of variation: is review conducted one-to-one, one-to-many, or as a group? Does it use advance preparation or is done on the day? Is there a leader or hierarchy, or just peers? The choices about which options to mix depends on whether the goal is fixing code, learning from code, or learning from other developers. Although all code review offers opportunities to learn about code and about best practices, activity focused on learning, group development, and knowledge transfer tends to include synchronous elements in order to develop rapport and trust. Preparation offline gives developers time to focus on the code, engage with it personally, and form independent opinions.

The groups that were most successful in the second study took the time for a synchronous conversation to introduce themselves and to discuss their goals and expectations (often in addition to introductory emails). Their mentors set out a review process and timetable at the outset, based on his or her experience. This usually involved a synchronous conversation of fixed length at regular intervals (3-4 weeks) with interim goals and tasks agreed during each of those conversations.

Finally, *do not underestimate returns from small investments*. None of the mentors expected scientists to overhaul complete code bases. The advice from one mentor was cogent: if you check the docstring and write a test every time you touch a method, the work will accumulate. This insight was echoed in different ways throughout the study. Adopting good practices and applying them incrementally, doing one thing at a time as it is needed, will achieve changes faster than you might think.

4 Conclusions

All of the mentors who engaged in the pilot study did so because, in their experience, the benefits of code review to science are profound in terms of improving reproducibility, promoting re-usability, disseminating best practices, and thereby improving efficiency and avoiding error. Moreover, they believe that the cost of introducing code review is quickly recovered, and that good practices can be introduced incrementally in ways that are not disruptive or expensive.

The scientists engaged in the pilot study because they wanted to write more accurate, efficient, and reusable code, and wanted support in learning how to do so. They too perceived that improvements to code development pay off in terms of their science, and code review provides an otherwise scarce opportunity to “talk code” and learn better practices. Many of

the scientists reported benefits even during the study: code improvements, better use of facilities provided in development tools, and better documentation practices spreading in the lab.

Nevertheless, many teams found it difficult to reserve the time to work with the mentor, given lab priorities such as bidding for further funding and meeting publication deadlines. And priorities were not the only obstacles: identifying specific code for review (from a larger code base) was not always obvious, the code was not always well-modularized and was rarely well-documented, and teams often had to put code into a repository for the very first to make it available for review. Some groups also needed to negotiate issues of confidentiality (for datasets as well as code).

Based on our study, we believe the following should be priorities when scientists start doing code review:

- Have a goal, a benefit in mind.
- Start with a conversation: articulate your goals and expectations, build rapport.
- Make your code available (in a repository). Include a typical data set, an overview of how the code works (e.g., the relevant scientific paper, a diagram of the structure of the software, an overview of the goals and context in which it was developed).
- Set up a schedule and commit a little time on a regular basis.
- Choose small enough chunks of code for discussion (typically 3–4 pages rather than 30–40).
- Understand that it’s the code that’s being reviewed (not you). Tailor your comments to others on the same terms.
- Make the process reciprocal: be prepared to make as well as receive comments.

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