What Is the Price of Open-Source Software?

The notion that all scientific software should be open-source and free has been actively promoted in recent years, mostly from the top down via mandates from funding agencies but occasionally from the bottom up, as exemplified by a recent Viewpoint in this journal. A commonly articulated rationale is that the results of scientific research funded by government grants should be free for society and that the scientific community benefits from free access. The purpose of this Viewpoint is to examine the consequences of these opinions.

What Is Scientific Software? Modern computational chemistry software is an extremely complex product based on advanced scientific ideas (models and theories) and sophisticated algorithms that transform these ideas from equations into useful tools. The development of practical software that can be used by nonexperts to solve contemporary research problems requires considerable technical effort to produce and maintain robust, efficient, and validated code. Unlike the development of, for example, a smartphone app, where the code base is small and a relatively large community can easily write extensions and add-ons, production of scientific software involves the curtailment of millions of lines of source code. The complexity of this code demands long-term user and developer support to maintain its integrity and performance while keeping up with new computer architectures, fixing bugs, and adding features. Recognizing the importance of these ideas, various funding agencies in the U.S. have made “sustainable software” a key priority in the distribution of research support. Sustainability is a critical goal, but one that can be realized in various ways.

Good Software Is Important to Science. Computational chemistry software is an essential scientific instrument that facilitates discovery and innovation far beyond the laboratories in which it is created, an achievement that was recognized by the 1998 and 2013 Nobel Prizes in Chemistry. Focusing on quantum chemistry software in this Viewpoint, we note that today any chemist can (with very little training) use numerous quantum chemistry programs as teaching and research tools that aid in the design and interpretation of experiments.

A software package should be more than just a tool for end users, however; it should also be a platform to develop and test new models and algorithms. Maintaining a code base requires extensive validation, and given the complexity of modern computational methods, even testing of "pilot code" or a "proof-of-principle" implementation requires access to basic software infrastructure, for example, an integrals library, a self-consistent field procedure, efficient I/O and memory management, tools for manipulating tensors, and so forth. Modularity is a laudable goal, but in reality, "interoperability" often comes at the expense of performance. In high-performance codes, the aforementioned components are tightly interwoven, to the extent that expert help is often required to modify key components or to develop nonstandard interfaces to them. As such, the ability to innovate along either applied or theoretical lines depends crucially on the quality of the software and the availability of documentation and expert support.

As examples, consider two widely used electronic structure programs, Q-CHEM and MOLPRO. These codes consist of ∼5.5 and ∼2.5 million lines of source code, respectively, written in multiple languages and each in continuous development over several decades. Q-CHEM incorporates scientific advances reported in more than 300 peer-reviewed scientific publications, whereas methods implemented for the first time in MOLPRO have led to 20 high-impact papers that have each been cited over 300 times. Neither code is static: more than 70 scientists are actively contributing to MOLPRO, and the Q-CHEM developer base numbers more than 100. Such agile innovation comes at a price, however. Significant effort is required to keep the code robust, efficient, and sound and to provide the documentation that ensures the usability of new methods and the extensibility of older ones.

Software from academia is often developed with an emphasis on ideas rather than implementation, fed by the need for timely peer-reviewed journal publications that provide ongoing grant support and future jobs for graduate students. To bring new ideas to the production level, with software that is accessible to (and useful for) the broader scientific community, contributions from expert programmers are required. These technical tasks usually cannot—and generally should not—be conducted by graduate students or postdocs, who should instead be focused on science and innovation. To this end, Q-CHEM employs four scientific programmers. Other quantum chemistry codes (e.g., MOLPRO, TURBOMOLE, JAGUAR, MOLCAS, PQS, and ONETEP) face the same challenges and adopt similar models to ensure sustainability.

It is important to distinguish these academically led software ventures from purely commercial endeavors. The large majority of the code in a package like Q-CHEM is funded by the government, either through grants to academic groups or, in some cases, through technology grants to the company itself. The role of the company programmers is to enable sustainability through bug fixes, user support, release management, and the addition of features that academic developers either cannot or will not add themselves. Programmers employed by the company place emphasis on functionality, robustness, and performance, more so than scientific innovation. They are directly addressing the "reproducibility problem." Sales revenue cannot support the entire development cost of an academic code, but it contributes critically to its sustainability. The cost that the customer pays for a code like Q-CHEM reflects this funding model: it is vastly lower than the development cost, particularly for academic customers but also for industry. It primarily reflects the sustainability cost.

Software Is Not Data. In his Viewpoint, Gezelter argues that both software and data should be open, yet it is important not to conflate the two. Software is not data, and simply because it is feasible to put software on the Internet does not imply that it should be posted. Software is a product that contains an intellectual component (models and algorithms) but owes its
existence to additional technical efforts. Such efforts include implementation of minor but useful (or requested) features that are not publishable in the peer-reviewed literature. This is not to say that details should be withheld as proprietary information. Just as models and algorithms should be described in full detail in scientific publications, so too should implementation details be specified, along with performance metrics (timings and scaling data) and benchmarks (energies and other computed properties). Nevertheless, the software itself is a product, not a scientific finding, more akin to, say, an NMR spectrometer—a sophisticated instrument—than to the spectra produced by that instrument.

Consider an analogy from the field of photovoltaics. Scientific findings concerning the mechanistic details of charge generation and exciton propagation in a given material are results that merit discussion in the peer-reviewed literature. However, creating a new solar panel based on this research requires significant additional engineering effort, which is most commonly conducted in an industrial setting. This is a common mechanism for technology transfer, by means of which society benefits from academic research. Likewise, new telecommunication technologies, information storage media, computer chips, and so forth are products that build upon these findings, more akin to, say, an NMR spectrometer—a sophisticated instrument—than to the spectra produced by that instrument.

**There Is No Free Software.** The creation of scientific software is a labor-intensive process, and its support and curation even more so. How do we pay for these labor costs? The answer is clear in the case of commercial software, where license fees are used to defray the costs of development and support. In this model, users buy the software that fits their research needs and afford them the highest productivity. The decision mechanism and price-versus-deliverables choices are similar to those faced when purchasing a computer or other lab equipment. Just like a new laser system may enable the pursuit of new science, software that offers a competitive advantage is a sensible investment of resources that is often impossible to achieve in the absence of a commercial platform. Software is not different.

Similarly, Gezelter specifically mentions the Quantum Chemistry Program Exchange (QCPE), an early repository for open-source software, as having contributed greatly to the growth of the field. It is therefore telling to note that QCPE was supported initially by the Air Force Office of Scientific Research and then by the National Science Foundation before subsequently becoming a fee-for-software service with paid employees to do the time-consuming work of testing, documenting, and distributing the contributed programs. Gezelter acknowledges the cost of maintaining scientific software and suggests alternative models to defray these costs including selling support, consulting, or an interface, all the while making the source code available for free. These suggestions strike us as naïve, something akin to giving away automobiles but charging for the mechanic who services them. Such a model creates a financial incentive to release a less-than-stellar product into the public domain, then charge to make it useful and usable. It is better to release a top-of-the-line product for a nominal fee.

Is “free” software genuinely free of charge to individual researchers? Consider software developed in the U.S. national laboratories. These ventures are supported by full-time scientific programmers employed specifically for the task, and the cost to support and develop these products is subtracted from the pool of research funding available to the rest of the community. The individual researcher pays for these codes, in a sense, with his rejected grant proposals in times of lean funding. In contrast to using one’s own performance metrics to guide software purchases, within this system, one has no choice in what one pays for. In other words, “free software” is not free for you; the only sense in which it is “free” is that you are freed from making a choice about how to spend your research money.

Computational chemistry software must balance the needs of two audiences: users, who gauge their productivity based on the speed, functionality, and user-friendliness of a given program; and developers, who may be more concerned with whether the structure “under the hood” provides an environment that fosters innovation and ease of implementation. As a quantitative example, consider that the cost of supporting a postdoctoral associate (salary plus benefits) is perhaps $4,800/month. If the use of well-supported commercial software can save 2 weeks of a postdoc’s time, then this would justify an expense of $2,000 to purchase a software license. This amount exceeds the cost of an academic license for many computational chemistry programs. Given the choice between a free product and a commercial one, a scientist should make a decision based on her own needs and her own criteria for doing innovative research.

**What Is “Open Source”?** The term “open source” is ubiquitous but its meaning is ambiguous. Some codes are “free” but are not open, whereas others make the source code available, albeit without binary executables, so that responsibility for compilation and installation is left to the user. Insofar as the use of commercial quantum chemistry software is a mainstay of modern chemical research and teaching, there exists a broad consensus that the commercial model offers the stability and user support that the community desires. Strict coding guidelines can be enforced within a model where source code access is limited to qualified developers, and this kind of stability offers one counterbalance to the “reproducibility crisis.” To the extent that such a crisis exists, it has occurred in spite of the existence of open-source electronic structure codes such as GAMESS, NWChem, and CP2k.

Occasionally the open-source model is touted on the grounds that one can use the source code to learn about the underlying algorithms, but this hardly seems relevant if the methods and algorithms are published in the scientific literature. Source code itself rarely constitutes enjoyable reading, and using source code to learn about an algorithm is a last resort forced by poorly written scientific papers. Better peer review is a more desirable solution.

A more practical use of openly available source code is to reuse parts of it in other programs, provided that the terms of the software license allow this. Often, they do not. Some ostensibly “open” chemistry codes forbid reuse, or even redistribution. Others, such as CP2k, use the restrictive General Public License that requires any derivative built on the original code to be open-source itself. Variation in design structure from one program to the next also severely hampers transferability, even if the license terms are amenable.

Access to source code allows developers to introduce their own innovations, but this is distinct from reuse. To facilitate innovation by developers, source code needs only to be available to people who intend to build upon it. This is commonly accomplished in the framework of “closed-source” software projects by granting academic groups access to the
source code for development purposes. Given the large number of
developers for many quantum chemistry codes, this
developer community should not be envisaged as some small,
secluded cabal but rather as a rich, diverse community of
academic scientists.

Let us analyze accessibility using Q-Chem as a specific
equation. This is commercial software whose source code is not
in the public domain because that would eliminate the product
that Q-Chem, Inc., is selling. Consider, however, how many
developers benefit from Q-Chem as a platform for their own
innovation: this community exceeds 100 scientists from at least
12 countries,7 for whom the code is open. We call this model
open teamware.5 Moreover, any licensed user can access to
the source code upon signing a nondisclosure agreement.
Given the size of the user base, this is likely a significantly larger
than the number of people who care to look at many
existing open-source packages. Does an “open-source” code
that serves just a few people offer more benefit to the scientific
community than a “closed-source” code that fosters a
community of 100+ active developers and thousands of
users? What would the impact be on computational chemistry
of destroying other teamware projects such as MOLPRO,6
TURBOMOLE,7 JAGUAR,8 MOLCAS,9 POS10 or onetep,11 in the
interest of satisfying some “open-source” mandate?
Practical Consequences of an Open-Source Mandate. One of the
pillars of science funding in the U.S. and elsewhere is a merit-
based funding model that distributes resources based on
intellectual merit, productivity, and impact. In the long run,
more-competitive ideas are selected over less-competitive ones,
and investigators are rewarded for a track record of
productivity. Research that is judged to have higher impact is
ranked as more meritorious and more deserving of support.
This model has proven successful in fostering innovation and
discovery, so it is worth considering what consequences a
blanket requirement that software be free and open-source
might engender.

Such a requirement would, in our view, detract from the
merit-based review process. When evaluating grant proposals
that involve software development, the questions to be asked should be:

1. What will be the quality of the software in terms of the
new science that it enables, either on the applications
side or on the development side?
2. How will the software foster productivity? For example,
how computationally efficient is it for a given task? How
usable will the software be, and how quickly will other
scientists be able to learn to use it for their own research?

A rigid, mindless focus on an open-source mantra is a
distraction from these more important criteria. It can even be
an excuse to ignore them, and creates an uneven playing field
in which developers who prefer to work with a commercial
platform are put at a disadvantage and potentially forced to
adopt less efficient practices.

Although Gezelter7 suggests that online code repositories
would help young researchers to showcase their work, an open-
source requirement actually puts young researchers at a
particular disadvantage. The demands of tenure and promotion
place special burdens to bring ideas before the community
rapidly and also to secure funding quickly. Open-source
requirements potentially force a scientist to choose between
pursuing a funding opportunity versus implementing an idea in
the quickest, most efficient, and highest-impact way. A strictly
open-source environment may furthermore disincentivize
young researchers to make new code available right away, lest
their ability to publish papers be short-circuited by a more
senior researcher with an army of postdocs poised to take
优势 of any new code. This would contribute directly to the
scenario that Gezelter wishes to avoid, namely, one where
students leave behind “orphaned” code that will never be
incorporated into mainstream, production-level software.

How should the impact of software be measured? Scientific
publications are a more sound metric than either the price of a
product or whether its source code is available in the public
domain. Software is meant to serve scientific research, in the
same way that any other scientific instrument is intended. As
such, the question should not be whether software is free or
open source, but rather, what new science can be accomplished
with it? Let us not allow political rhetoric to dictate how we are
to do science. Let different ideas and different models
(including open source!) compete freely and flourish, and let
the community focus instead on the most important metric of
all: what is good for scientific discovery.

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(13) See, for example, the Orca license. https://orcaforum.cec.mpg.de/license.html (accessed April 2015).


(17) See, for example, the GAMESS license: http://www.msg.ameslab.gov/gamess/License_Agreement.html (accessed April 2015).
