
Commentary | Novel Tools and Methods

The Future is Open: Open-Source Tools for Behavioral Neuroscience Research

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**The Future is Open: Open-Source Tools for Behavioral
Neuroscience Research**

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Abbreviated title: Open Source Tools for Behavioral Neuroscience

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51 **Significance Statement**

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53 There has been a recent and substantial increase in the use of open-source tools for conducting
54 research studies in neuroscience. The OpenBehavior Project was created to disseminate open-
55 source projects specific to the study of behavior. In this commentary, we emphasize the benefits
56 of adopting an open-source mindset and highlight current methods and projects that give
57 promise for open-source tools to drive advancement of behavioral measurement and ultimately
58 understanding the neural basis of behavior.
59

60 Over the past decade, there has been an explosion in the use of new neurobiological
61 tools for measuring and controlling brain cell activity. Recent developments in optogenetics,
62 chemogenetics, cellular imaging, and fiber photometry have spiked publications across cellular,
63 systems, and behavioral neuroscience. Researchers with expertise in molecular biology or
64 cellular physiology are now carrying out behavioral studies, and often bring a fresh approach to
65 the fine-grained study of behavior that has led to the development of many new assays for
66 measuring behavior and cognition in animal models (mice, flies, worms, etc).

67 Thanks to a revolution in low-cost methods for 3D printing and off-the-shelf
68 microcontrollers (e.g. Arduino, Teensy, microPython) and single-board computers (Raspberry
69 Pi), many of these research groups are able to create complex behavioral tasks quite easily.
70 The R and Python languages, specialized computing libraries (e.g. numpy, OpenCV,
71 TensorFlow), and the Anaconda Python distribution have been crucial for the development of
72 open source analysis software for neuroscience projects. In parallel, these developments in
73 neuroscience research have occurred during a time when there is a simultaneous movement
74 towards sharing computer code (Gleeson et al., 2017; Eglén et al., 2017), through websites like
75 GitHub, and opening up the process of software and hardware design to non-experts through
76 hackerspaces and makerspaces.

77 Despite these developments, there is still room for growth with regards to sharing.
78 Designs for some new tools have been posted on websites created by individual researchers or
79 shared via public repositories such as GitHub. In other cases, designs and protocols have been
80 published and several new journals and tracks in existing journals are emerging for reports on
81 open source hardware and software. In this commentary, we aim to emphasize the benefits of
82 adopting an open source mindset for the behavioral neuroscience field, and we highlight current
83 methods and projects that give promise for open source tools to drive advancement of
84 behavioral measurement and ultimately understanding the neural basis of these behaviors.

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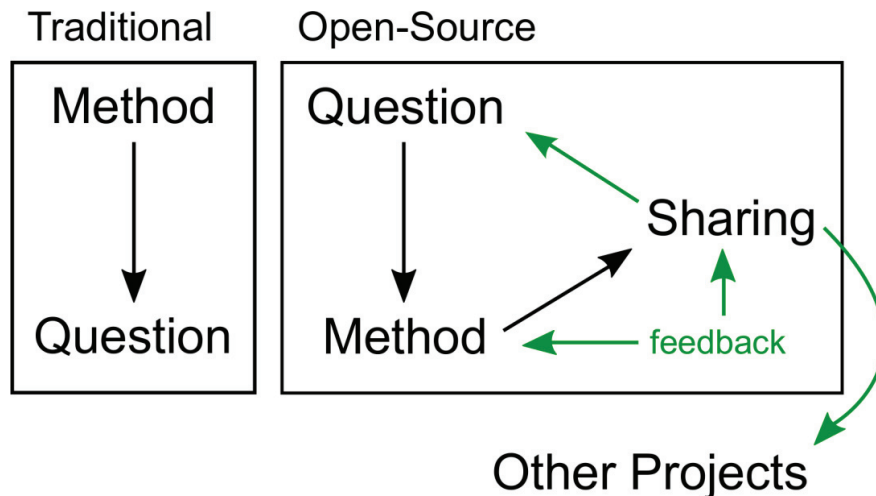
86 **Why open source?**

87 The main idea behind an open source project is that the creator or developer provides
88 open access to the source code and design files, whether that be for software or hardware.
89 Open source projects typically provide a license for others to use and modify the design,
90 although many licenses require that any modifications remain open source. Under such
91 licenses, it is not permissible to take an open source design, modify a few things, and claim it is
92 a new closed design. Releasing a project with an open source license provides transparency for
93 others to view, modify, and improve the project. open source can be relevant for many levels of
94 scientific research; open-access journals, code and data repositories, and sharing methods,
95 protocols, or files are all examples of how one can contribute to open source science.

96 The term “open source” is also often synonymous with being cost-effective. Many
97 commercial products used in neuroscience can be replicated in an open source manner at a
98 fraction of the initial cost. However, there are additional advantages to incorporating open
99 source science in a research lab. With a recent increase in microcontrollers, microprocessors,
100 3D printing and laser-cutting technologies, most people now have access to create devices or
101 products in a way that was previously unavailable to researchers. Additionally, a major benefit to

102 open source science is the ability for customization and flexibility. Instead of being restricted to
 103 studying only what a commercial part is capable of doing or measuring, it is now possible to
 104 study a level deeper through developing a device or software that will help answer the research
 105 question, instead of letting the technology drive the research question (Figure 1). In behavioral
 106 neuroscience, this allows researchers to enter uncharted territory of analyzing previously
 107 unmeasured or fine-grained aspects of behavior (Krakauer et al., 2017).

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Figure 1: Open Source Creative Process: Methods and Questions

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Traditional methods in neuroscience are purchased commercially, and are used

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to answer a specific research question. Due to the need to maximize use based

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on the cost of the tool, the method often *drives* subsequent research questions.

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However, in an open source model, the research question drives the

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development of a method or tool. A major advantage of this in behavioral

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neuroscience is that previously unmeasured aspects of behavior now have the

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potential to be measured, leading to a new frontier of behavioral measurement

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and analysis. The tool is subsequently shared to the community, and the user

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seeks feedback from the community to refine the method. Sharing of an open

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source tool leads to the development of new projects across multiple research

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labs, leading researchers to, quite literally, think “outside the box.”

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Several extremely successful projects have come from this open source movement

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(Maia Chagas, 2018), including neuroscience projects such as the Open Electrophysiology

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project (Siegle et al., 2017), the UCLA miniaturized microscope (Aharoni et al., 2019), and

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software such as Bonsai (Lopes et al., 2015) and DeepLabCut (Mathis et al., 2018) for video

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recordings and analysis. However, the field of open source neuroscience is expanding at a rapid

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pace and it is becoming hard to keep up with all the latest advances in research tools and the

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hardware and software that has enabled them.

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131 The OpenBehavior Project

132 In 2016, it became clear that there were many projects reporting on new tools for the
133 study of behavior, and thus we launched the OpenBehavior project. Access to design files and
134 build instructions relied on word of mouth and isolated blogs and posts on social media. We
135 made it our goal to disseminate information about tools as soon as they emerge as preprints on
136 bioRxiv or PsyArXiv, peer-reviewed manuscripts, or independent posts by developers on
137 hackaday, GitHub, lab websites, or social media. The project is based around a website
138 covering bleeding-edge open source tools and a related Twitter account that keeps followers
139 up-to-date with new projects relevant to behavioral neuroscience in species from flies and fish,
140 to rodent and, more recently, humans. Through these efforts, we hope to contribute to the rapid
141 replication and adoption of new tools into ongoing research and trigger modifications of existing
142 tools for novel research applications.

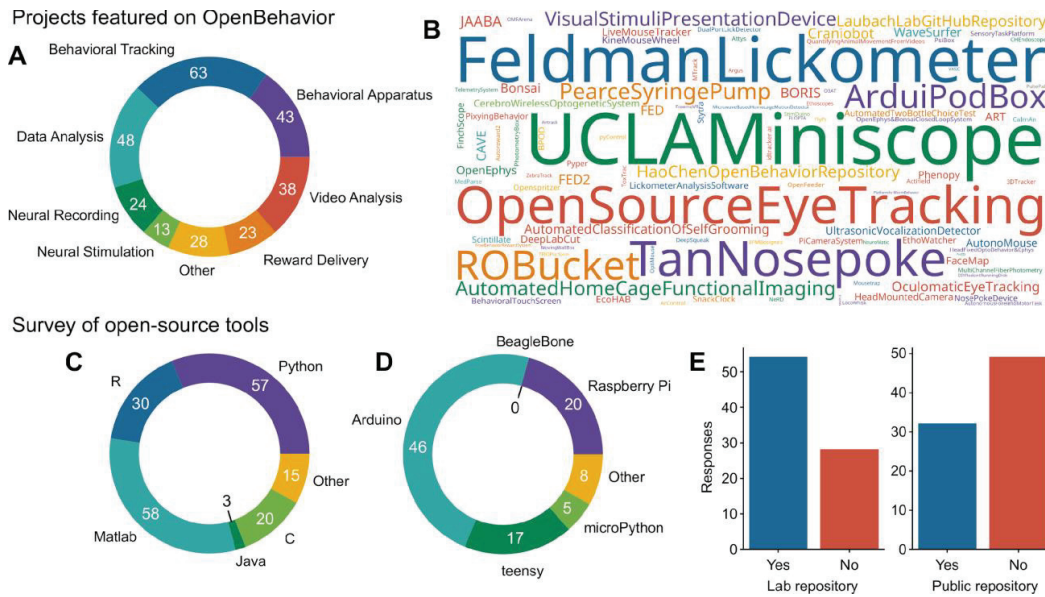
143 To date, dozens of projects have been shared through OpenBehavior.com, with even
144 more shared through active Twitter engagement. In May 2019, we celebrated our 100th open
145 source project post, which have covered devices for delivering rewarding foods and fluids,
146 measuring home cage activity, video tracking and analysis, and physiological methods used in
147 behavioral experiments such as miniaturized microscopes and fiber photometry (Figure 2A-B).
148 While video analysis is a prominent focus of many projects, several other types of projects have
149 been popular on the site, including devices for tracking patterns of feeding behavior in the home
150 cage (FED – Nguyen et al., 2016), a system for multi-channel electrophysiology (OpenEphys –
151 Siegle et al., 2017), systems for fiber photometry (PhotometryBox - Owens & Kreitzer, 2019),
152 stimulators for optogenetic stimulation (Stimduino – Sheinin et al., 2015), supervised (JAABA –
153 Kabra et al., 2013; DeepLabCut – Mathis et al., 2018) and unsupervised (FaceMap – Stringer et
154 al., 2019) machine learning algorithms for analyzing behaviors from video, and integrated
155 systems for behavioral control (Bpod – RRID:SCR_015943) including video recording and real-
156 time analysis (Bonsai – Lopes et al., 2015). Recently, we have begun to track and share tools
157 for research in human behavioral neuroscience, computational models, and relevant data
158 analysis methods.

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160 Sharing and dissemination of open source tools

161 Thanks to the sharing of proper documentation and an understanding of open source
162 methods, researchers were able to modify some of the projects to better fit their experiments'
163 needs. One example of how open source tools can lead to new research projects is found in
164 some of the earliest posts on OpenBehavior. We featured a number of devices for delivering
165 rewarding fluids to rodents. One project, the Automated Mouse Homecage Two-Bottle Choice
166 Test by Dr. Meaghan Creed, was developed to allow for automated taste preference tests and
167 oral self-administration studies in mice. The project was posted to a website for sharing the
168 designs of open source hardware (hackaday.io) and the device was quickly used by a number
169 of labs. One of these labs, with knowledge of open source methods and insightful
170 documentation from Creed, was able to modify the device using a more advanced
171 microcontroller which allowed them to measure fluid consumption over 16 reward tubes

172 simultaneously in rats (Frie & Khokhar, 2019). The experiences of these users of our website
 173 and followers of our Twitter feed indicate that we have had strong initial success in our overall
 174 mission to accelerate research through promotion of collaboration and sharing.
 175



176 **Figure 2: Projects featured on OpenBehavior and a survey of our followers.**

177 A - Types of projects featured on OpenBehavior. The most common type of
 178 project allows for tracking behaviors in video recordings. Most projects have
 179 multiple functions. For example, Bonsai can be used for video recording, tracking
 180 behaviors, and controlling behavioral equipment. B - Based on web hits from
 181 unique URLs, we depict the overall interest of our followers. C - A survey on use
 182 of open source tools revealed that most labs use more than one programming
 183 language, with Matlab/Octave and Python most commonly used. D - The survey
 184 also found that the majority of respondents reported using Arduinos
 185 microcontrollers, and less common tools included Raspberry Pi single board
 186 computers and Teensy microcontrollers. E - The majority of respondents
 187 reported having repositories for code and designs in their labs. However, most of
 188 these researchers did not report use of public repositories.
 189

190 To assess how OpenBehavior might further improve sharing and dissemination, in the
 191 spring of 2019 we conducted an online survey. While not a scientific poll, the results are
 192 informative about the views and needs of the open source community of behavioral
 193 neuroscientists. Fifty percent of respondents (48 out of 70) indicated that they follow the site
 194 with the intention to incorporate some of the devices and software that we have profiled into
 195 their research programs in the future. Another thirty percent of survey respondents (22 out of

196 70) indicated they have used tools featured on the site that were not developed by their own
197 labs either straight from the project documentation (16 out of 22) or with some modifications of
198 their own (6 out of 22). Many participants who reported integration of open source tools into
199 their research programs have often incorporated more than one, which has generated their own
200 documented method for recording and analyzing behavior (van de Boom et al., 2017) or
201 generated full closed-loop systems for behavioral experiments (Solari et al., 2018; Buccino et
202 al., 2018).

203 Further efforts on dissemination and training are needed to further the impact of
204 OpenBehavior and related projects within the research community. We are exploring adding a
205 forum to the website to encourage interactions between developers and users, which was
206 suggested by several participants of our survey. Furthermore, we would like to inspire DIY
207 hackers and open source engineers to think about projects that could be useful for behavioral
208 neuroscience, just as we've begun to seek hackers to make sense of large datasets in
209 neuroscience (Goodwin, 2018). To this end, we have initiated efforts through a partnership with
210 Hackaday.io, a website that is popular in the DIY community.

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212 **Expanding adoption of open source tools**

213 Despite these advantages of open source tools, incentives to sharing and the ability to
214 categorize and disseminate developments remains a challenge. Worse, there are major
215 technical barriers that hold many researchers back from diving headfirst into a newly released
216 research tool. Not everyone has the incentive, skills, or time needed to incorporate new tools
217 into ongoing research projects. It takes time to learn the skills required to build new devices and
218 programs from source. Clear instructions from developers are further needed to recreate and
219 use new devices and programs. Concerns persist about the reliability of self-made devices or
220 undiscovered bugs in programs written for relatively small user bases. The lack of immediately
221 available technical support and extensive validation of new tools does not add positively to
222 confidence in using new open source tools.

223 Notwithstanding these concerns, there has been movement towards to the use of open
224 source software and hardware in neuroscience as well as evidence for sharing new tools by
225 neuroscience labs. To assess how followers of OpenBehavior make use of software and
226 hardware in their research, we ran a second on-line survey in late May 2019 that queried
227 respondents about the programming languages used in their labs, their use of microcontrollers,
228 3D printers, and printed circuit boards, and also whether they used in-lab and/or public
229 repositories for their code and designs. Findings from the survey are described in Figure 2C-E.
230 Remarkably with regards to sharing, while most (65%) respondents reported having repositories
231 for their labs (54 of 82), less than 40% of respondents (32 of 81) reported sharing their code and
232 designs on public repositories.

233 These findings are relevant in the light of ongoing discussions about the availability of
234 neural data and analysis code (Halchenko & Hanke, 2012; Ascoli et al., 2017; Eglén et al.,
235 2017; Gleeson et al., 2017), and open sharing of new methods for data collection (OpenEphys –
236 Siegle et al., 2017; UCLA miniscope – Aharoni et al., 2019). We hope that this will lead to new
237 conversations about sharing behavioral data, analysis code, and hardware. It seems

238 straightforward to encourage an open source mindset, which can be done across several levels.
239 Anyone should be able to replicate an open source project, given they are provided with
240 detailed documentation and dissemination of software or hardware devices. It is necessary to
241 encourage a set of standards to make reproducibility possible, such as in the methods for two-
242 bottle preference testing described above. See Box 1 for our recommendations for best
243 practices in developing open source tools.

244 Additional efforts are needed to offer and maintain productivity using open source tools.
245 There is a need for forums for public discussions on the tools, perhaps through the Neuronline
246 forums managed by SfN. There will always be some troubleshooting, which is why an open
247 forum for sharing feedback on already developed tools is necessary. To further drive innovation
248 and development, we suggest implementing webinars, online tutorials, and workshops to allow
249 people all over the world to have access to the development of open source tools. Hands-on
250 workshops have been successful for several open source tools, such as optogenetics,
251 CLARITY, Miniscope, and DeepLabCut. These activities will require financial support to enable
252 storing data, designs, and protocols, maintaining a well-documented website and source code,
253 and offering training workshops. We hope that major funders (e.g. NIH, NSF, EU) consider
254 providing special opportunities for supporting development and training for open source
255 research tools.

256 Finally, there is a need for tracking the use of open source tools, by creating and utilizing
257 RRIIDs (SciCrunch) in publications. To our knowledge, RRIIDs have not been commonly created
258 for hardware. Having a system for tracking usage has three potential impacts. First, tool usage
259 can be tracked beyond citations of methods papers. Second, revisions and spin-offs can be
260 noted and also tracked. Third, developers might have increased incentives to share designs
261 early in the process, especially if an index, similar to the h factor, was developed for RRIIDs
262 Inevitably, creating new platforms and incentives for sharing the development, use, and
263 replication of open source behavioral tools is crucial for bringing open source science to the
264 forefront.

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267 **Box 1. Recommendations for best practices in developing open source tools**

268 1. **Clear documentation of the project:** Provide all design files, as well as a Bill of Materials,
269 Build Instructions, graphical (video/photo/3D renderings) descriptions or tutorials for the project.

270 2. **Central repository for files:** Provide all files and documentation of the project on a site like
271 GitHub, Hackaday.io, OSF.io, or on the research group's website.

272 3. **Experimental validation:** Show an example of the device being used in a behavioral
273 experiment.

274 4. **Make the project easy to find:** Create a Research Resource Identifier (RRID), using the
275 SciCrunch project, for the device so that others can track the project across publications.

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