

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

<https://doi.org/10.1523/ENEURO.0223-19.2019>

**Cite as:** eNeuro 2019; 10.1523/ENEURO.0223-19.2019

Received: 11 June 2019

Revised: 12 July 2019

Accepted: 21 July 2019

---

*This Early Release article has been peer-reviewed and accepted, but has not been through the composition and copyediting processes. The final version may differ slightly in style or formatting and will contain links to any extended data.*

**Alerts:** Sign up at [www.eneuro.org/alerts](http://www.eneuro.org/alerts) to receive customized email alerts when the fully formatted version of this article is published.

Copyright © 2019 White et al.

This is an open-access article distributed under the terms of the Creative Commons Attribution 4.0 International license, which permits unrestricted use, distribution and reproduction in any medium provided that the original work is properly attributed.



51 **Significance Statement**

52

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.

85

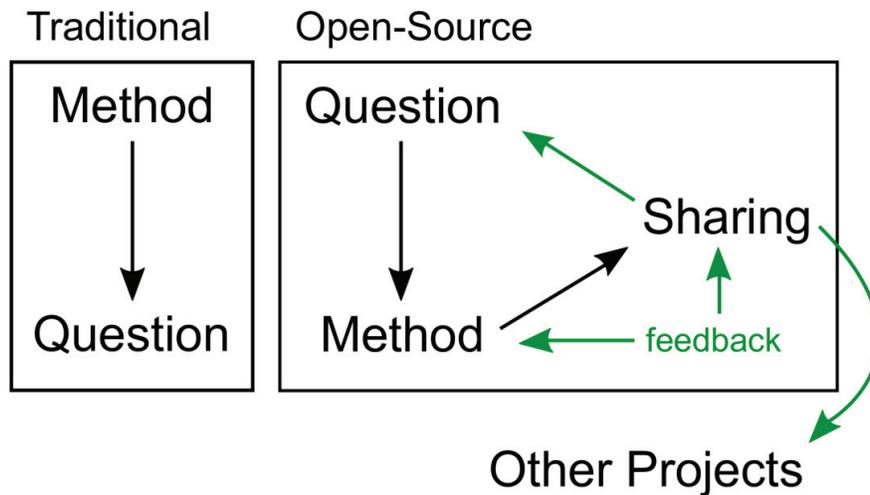
### 86 **Why open source?**

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).

108  
 109



110

**Figure 1: Open Source Creative Process: Methods and Questions**

111

Traditional methods in neuroscience are purchased commercially, and are used

112

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

113

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

114

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

115

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

116

neuroscience is that previously unmeasured aspects of behavior now have the

117

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

118

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

119

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

120

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

121

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

122

123

Several extremely successful projects have come from this open source movement

124

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

125

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

126

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

127

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

128

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

129

hardware and software that has enabled them.

130

**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.

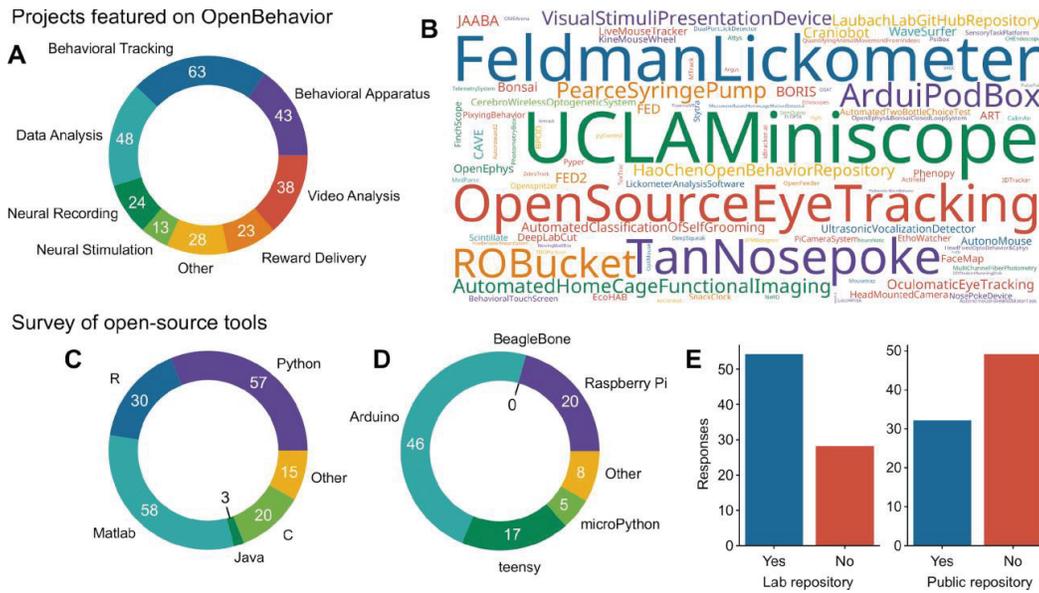
143 To date, dozens of projects have been shared through OpenBehavior.com, with even  
144 more shared through active Twitter engagement. In May 2019, we celebrated our 100th open  
145 source project post, which have covered devices for delivering rewarding foods and fluids,  
146 measuring home cage 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.

159

**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.

211

#### 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 RRIDs (SciCrunch) in publications. To our knowledge, RRIDs 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 RRIDs  
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.

265  
266

- 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.
- 276
- 277

278 **Citations**

- 279 Aharoni D, Khakh BS, Silva AJ, Golshani P (2019) All the light that we can see: a new era in  
280 miniaturized microscopy. *Nature Methods* 16:11–13.
- 281 Ascoli GA, Maraver P, Nanda S, Polavaram S, Armañanzas R (2017) Win–win data sharing in  
282 neuroscience. *Nature Methods* 14:112–116.
- 283 Buccino AP, Lepperød ME, Dragly S-A, Häfliger P, Fyhn M, Hafting T (2018) Open source  
284 modules for tracking animal behavior and closed-loop stimulation based on Open Ephys  
285 and Bonsai. *Journal of Neural Engineering* 15:055002.
- 286 Eglen SJ, Marwick B, Halchenko YO, Hanke M, Sufi S, Gleeson P, Silver RA, Davison AP,  
287 Lanyon L, Abrams M, Wachtler T, Willshaw DJ, Pouzat C, Poline J-B (2017) Toward  
288 standard practices for sharing computer code and programs in neuroscience. *Nature*  
289 *Neuroscience* 20:770–773.
- 290 Frie JA, Khokhar JY (2019) An open source automated two-bottle choice test apparatus for rats.  
291 *HardwareX* 5:e00061.
- 292 Gleeson P, Davison AP, Silver RA, Ascoli GA (2017) A Commitment to Open Source in  
293 Neuroscience. *Neuron* 96:964–965.
- 294 Goodwin D (2015) Neuroscience Needs Hackers. *Scientific American* 313:14–14.
- 295 Halchenko YO, Hanke M (2012) Open is Not Enough. Let’s Take the Next Step: An Integrated,  
296 Community-Driven Computing Platform for Neuroscience. *Frontiers in Neuroinformatics*  
297 6 Available at: <http://journal.frontiersin.org/article/10.3389/fninf.2012.00022/abstract>  
298 [Accessed June 11, 2019].
- 299 Kabra M, Robie AA, Rivera-Alba M, Branson S, Branson K (2013) JAABA: interactive machine  
300 learning for automatic annotation of animal behavior. *Nature Methods* 10:64–67.
- 301 Krakauer JW, Ghazanfar AA, Gomez-Marin A, Maclver MA, Poeppel D (2017) Neuroscience  
302 Needs Behavior: Correcting a Reductionist Bias. *Neuron* 93:480–490.
- 303 Lopes G, Bonacchi N, Frazão J, Neto JP, Atallah BV, Soares S, Moreira L, Matias S, Itskov  
304 PM, Correia PA, Medina RE, Calcaterra L, Dreosti E, Paton JJ, Kampff AR (2015)  
305 Bonsai: an event-based framework for processing and controlling data streams.  
306 *Frontiers in Neuroinformatics* 9 Available at:  
307 <http://journal.frontiersin.org/article/10.3389/fninf.2015.00007/abstract> [Accessed June  
308 11, 2019].
- 309 Maia Chagas A (2018) Haves and have nots must find a better way: The case for open scientific  
310 hardware. *PLOS Biology* 16:e3000014.
- 311 Mathis A, Mamidanna P, Cury KM, Abe T, Murthy VN, Mathis MW, Bethge M (2018)  
312 DeepLabCut: markerless pose estimation of user-defined body parts with deep learning.  
313 *Nature Neuroscience* 21:1281–1289.
- 314 Nguyen KP, O’Neal TJ, Bolonduro OA, White E, Kravitz AV (2016) Feeding Experimentation  
315 Device (FED): A flexible open source device for measuring feeding behavior. *Journal of*  
316 *Neuroscience Methods* 267:108–114.
- 317 Owen SF, Kreitzer AC (2019) An open source control system for in vivo fluorescence  
318 measurements from deep-brain structures. *Journal of Neuroscience Methods* 311:170–  
319 177.

- 320 Sheinin A, Lavi A, Michaelovski I (2015) StimDuino: An Arduino-based electrophysiological  
321 stimulus isolator. *Journal of Neuroscience Methods* 243:8–17.
- 322 Siegle JH, López AC, Patel YA, Abramov K, Ohayon S, Voigts J (2017) Open Ephys: an open  
323 source, plugin-based platform for multichannel electrophysiology. *Journal of Neural*  
324 *Engineering* 14:045003.
- 325 Solari N, Sviatkó K, Laszlovszky T, Hegedüs P, Hangya B (2018) Open Source Tools for  
326 Temporally Controlled Rodent Behavior Suitable for Electrophysiology and Optogenetic  
327 Manipulations. *Frontiers in Systems Neuroscience* 12 Available at:  
328 <http://journal.frontiersin.org/article/10.3389/fnsys.2018.00018/full> [Accessed June 11,  
329 2019].
- 330 Stringer C, Pachitariu M, Steinmetz N, Reddy CB, Carandini M, Harris KD (2019) Spontaneous  
331 behaviors drive multidimensional, brainwide activity. *Science* 364:eaav7893.