

1 What is ‘Natural’ about Naturalistic Neuroscience?

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4 SUMMARY:

5 A growing number of neuroscientific articles now discuss the revolutionary tools and techniques of
6 naturalistic behavioral studies. Falling under the umbrella of as ‘naturalistic neuroscience’, these studies
7 aim to impart the precision and control of traditional behavioral experiments while also documenting
8 ‘real-world’ animal behavior. The present study examines the tools and techniques used in these studies
9 from both a theoretical and modeling perspective. Results of this paper demonstrate the contradictions
10 generated by ‘naturalistic’ empirical manipulations, as well as cases where the process of experimental
11 comparability is hindered. I show that focusing on competing experimental aims more closely aligns
12 experimental outcomes compared to the coarse-grained use of the ‘naturalistic’ concept, which instead
13 oversimplifies the complexity of methodological drivers in behavioral neuroscience. From this analysis, I
14 recommend the use of a revised framework that imparts greater transparency in the experimental aims of
15 researchers studying animal behavior.

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17 Introduction

18 A central challenge for Neuroscience has been understanding how nervous systems flexibly and reliably
19 generate complex behaviors. How does an animal distinguish a benign encounter from a threat? How is
20 irrelevant information ignored to satisfy its needs? Since the days of Pavlov’s salivating dogs or Skinner’s
21 bar pressing rats, behavioral neuroscientists have constructed highly constrained lab paradigms to study
22 how experience modifies relatively simple behaviors. These behaviors give scientists the benefit of
23 precision and control: by manipulating the temporal relations between stimulus and response, neural
24 activity can be directly tied to the behavior. However, these behaviors are also seen as highly contrived in
25 the sense that there are no levers or bells in the habitats in which rats’ and dogs’ brains evolved, which
26 presumably shaped the neural circuits that generate most behaviors.

27 In parallel to simplified behaviors, traditional approaches in behavioral neuroscience have
28 focused on animals with simpler nervous systems. For example, the nematode, *Caenorhabditis elegans* with
29 a total of 302 neurons, and the sea slug, *Aplysia californica*, whose nervous system consists of a series of
30 neuronal clusters (ganglia) containing specific neurons identifiable from one slug to the next, were pressed
31 into service to dissect the complicated processes of learning and memory. Experiments using these
32 simplifications in behavior and in the nervous system – what allowed ready access to genetic manipulations
33 - have long been viewed as essential for meeting the challenge of explaining behavioral complexity.
34 The ruthless drive towards simplicity has ignored the many complex behaviors that animals exhibit in their
35 native environments – behaviors shaped evolutionarily by natural and sexual selection acting on
36 development and mature function of CNS neural circuits (Olton and Samuelson 1976; Morris 1981; Miller
37 et al., 2020). By restricting behavioral expression, simplified experiments can generate experimental
38 artifacts and also ignore key individual differences between animals. Neuronal responses can differ when
39 animals are studied outside of artificial lab conditions (Polley et al. 2004), overtraining animals in fixed

40 conditions affects plasticity (Jahangiri et al. 2019), animals respond differently to artificial and natural
41 stimuli (Cuthill et al. 2000; Fleishman et al. 1998), and stress can shape behavioral responses (Brandl et al.
42 2022, Gouveia and Hurst 2017; Deacon 2006).

43 Traditional, standardized paradigms of operant boxes or associative learning tasks imported into
44 behavioral neuroscience long ago have also been seen as inadequate for providing insights into how animals
45 variably navigate, choose, plan, persevere or give up in their complex natural worlds. Instead, to offer a
46 more ecologically valid study of behavior, scientists use novel methods that still offer precision and control
47 for measuring an animal’s activity, but now with the added sensitivity of study animal behavior without
48 traditional methods of restraint. This movement, known as ‘Naturalistic neuroscience’, has been described
49 as a ‘revolution’ in behavioral research (Anderson and Perona 2014). Naturalistic neuroscientific studies
50 characterize animal behavior as they would occur ‘in the wild’, albeit combined with the control and
51 precision of a standard conditioning experiment which rely on improved tools for measuring behavioral and
52 neural activity (Dennis et al. 2021, Matusz et al. 2019, Hoffmann et al. 2023; Ulanovsky, forthcoming).
53 In its promise to hit a Goldilocks balance between degrees of freedom and experimental control, naturalistic
54 neuroscience distinguishes itself from pre-existing frameworks for behavioral study. This balance promises
55 capturing an animal’s ‘true’ behavior by expanding the dimensionality of testing while maintaining rigorous
56 experimental standards. Of note within these efforts, ‘naturalistic neuroscientific’ studies embed the
57 assumption that the privileged way to study behavior is to get as close as possible to an animal’s true,
58 unaltered behavior.

59 However, the coarse-grained concept of ‘naturalistic’ oversimplifies the complexity of issues that
60 methodologically drive behavioral neuroscience. ‘Naturalistic’ acts as a placeholder for different conceptual
61 and technological experimental goals, as well as various experimental traditions that range from the use of
62 ethological to computational theories. By outlining these experimental aims, I provide evidence against the
63 view that there is an epistemically privileged way to study behavior, and I caution future empirical studies
64 from using the term ‘naturalistic’ to describe their behavioral studies. The present study provides evidence
65 of empirical harm when scientists use the underspecified description of ‘naturalistic’ behavior, focusing on
66 how experimental contradictions can get generated within behavioral research, as well as cases where the
67 process of experimental comparability is stifled.

68 **The Many Practices that Make Up Naturalistic Neuroscience**

69 Naturalistic neuroscientific studies are distinct for achieving the control and precision of traditional behavioral
70 experiments while showing sensitivity to animals’ complexity of responses. Methods cited in naturalistic
71 studies vary widely, ranging recording freely moving and uncaged animals (Mao et al. 2021, Voloh et al.
72 2023) to entirely substituting the lab with spaces that mimic natural environments (Yartsev and Ulanovsky
73 2013) or doing fieldwork studies where the animal is entirely ‘in the wild’ (Vallet and Wassenhove 2023).
74 Many of these efforts take inspiration from ethology, neuroethology, behavioral ecology, and related fields,

75 where there is a longstanding emphasis on environmental considerations and evolutionary perspectives
76 (Krakauer et al. 2017; Miller et al. 2023; Testard et al. 2021; Datta et al. 2019).
77 At the heart of naturalistic neuroscience are the novel technologies that can now capture individual and
78 collective behaviors of freely moving animals (Berdahl et al. 2013; Soria et al. 2021). These tools are thought
79 to confer advantages over both highly contrived research, as well as studies that might have studied animal
80 behavior in the wild, but lacked the tools that could still give scientists control.

81 For example, one important innovation in naturalistic neuroscience has been developing tools that no
82 longer confine animal movements when engaging in neural recording. Novel techniques, such as wireless
83 head-mounted optogenetic systems (Qazi et al. 2018, Hashimoto et al. 2014, Montgomery et al. 2015) and
84 ultrathin multifunctional optoelectronic devices (Kim et al. 2013) now allow animals to run around freely
85 without close proximity to the computer that is recording neural activity. Advances in materials science have
86 created more flexible electronics, allowing researchers to use optogenetic constructs with a single implanted
87 device without the need for batteries (McCall et al. 2013); developments in wirelessly rechargeable batteries
88 mean that this information can be recorded for days at a time (Kim et al. 2021). Wirelessly networked
89 microchips allow for neural recording and microstimulation (Lee et al. 2021), and the development of
90 miniature wireless fluorescence microscopes (miniScopes) has allowed researchers to record neural activities
91 of freely moving animals at the resolution of single cells (Zong et al. 2021; Dong et al. 2024). These
92 techniques have even expanded the scope of studies to include both the central and peripheral nervous
93 systems (Park et al. 2015).

94 Behavioral paradigms have evolved in parallel with neural approaches and now include virtual
95 reality environments that test animals' responses to rich sensory stimuli (Brown and de Bivort 2018; Naik et
96 al. 2020). Here too, naturalistic studies benefit from sophisticated technologies, such as improved projectors.
97 Virtual realities and avatars have allowed researchers to better test sophisticated social behaviors or
98 interactions between animals and others (Huang et al. 2020), as well as simple navigation (Jeung et al. 2023).
99 Technological advances also allow for a wider range of experimental subjects including, for examples, small
100 invertebrates, in sophisticated testing conditions (Peckmezian and Taylor 2015, Schultheiss et al. 2017).
101 These new tools also enable novel research using traditional model systems. For example, experimentalists
102 can now do whole brain imaging in freely moving zebrafish (Hasani et al. 2023) and conduct neural
103 recordings in freely moving bats (Ulanovsky and Moss 2007; Yartsev and Ulanovsky 2013). Moreover, new
104 tools have opened possibilities for studying novel and nontraditional animal models that were previously
105 inaccessible to neuroscientific query. Examples include miniature microdrives used to study food caching
106 behavior in freely moving tufted titmice (Payne et al. 2021) and remote monitoring techniques used to
107 document sleep in elephant seals (Kendall-Bar et al. 2023). Labs using these techniques also show signs of the
108 theoretical shifts that emerge from these new tools. For example, neuroscientists increasingly cite
109 incorporating more evolutionary and environmental outlooks in the study of behavior, such as examining

110 animals that are unique from a phylogenetic standpoint. Such studies include how animals transitioned from
111 aquatic mediums to land (salamanders), or those whose behaviors change dramatically depending on
112 environmental demands (shrew, squirrel).

113 **Framing Naturalistic Neuroscience: True Behavior Without Human Intervention**

114 The emphasis on objectivity currently frames the framework for naturalistic neuroscience. Naturalistic studies
115 are viewed as achieving “true” theories of the brain or helping to identify “real” or “real-life” behaviors in the
116 world (Miller et al. 2022, p. 13; Mobbs et al. 2021; Shamay-Tsoory and Mendelsohn, 2019). Studying
117 naturalistic behaviors includes studying species in their natural habitats (Vallet and Wassenhove 2023), as
118 well as modifying the lab environment to resemble the animal’s natural habitat.

119 The technological advances of naturalistic neuroscience – what sits at the heart of these efforts to capture the
120 truth about behavior – also rest on a promise to objectively measure behavior. One way to do this is to record
121 more information, as it is regularly assumed that greater amounts of data used to capture behavior yield
122 greater accuracy about animal activities. Within this framework, ‘naturalistic’ is a concept that linearly and
123 hierarchically ‘scales’ (Fan et al. 2021), with the practices of naturalistic behavior best adopted in “modest
124 steps” (Cisek 2024). Such language about naturalistic neuroscience assumes ‘improvement’ to pre-existing
125 behavioral experiments, once studied in impoverished ways and measured less accurately.

126 Relatedly, the technologies that drive naturalistic studies of behavior are also thought to improve
127 accuracy by *changing* and even *reconceptualizing* categories of behavior. Here, AI tools have seemingly
128 improved experimentation by refining behavioral categories previously recognized by humans. For example,
129 by using machine learning video-based tracking, researchers are able to rethink associations with behaviors,
130 such as the relationship between sleep and survival (Geissmann et al. 2019). In other cases, the use of these
131 techniques has identified novel behaviors (Hoyer et al. 2008). The objectivity of these tools is tied to the sheer
132 amount of data captured by them, which use sophisticated analyses to identify which functional categories are
133 significant.

134 Generally, the strategy of naturalistic neuroscience tools has consistently aimed to remove the
135 subjectivity and bias of human observers from behavioral experiments. For instance, better vision recognition
136 technologies allow tracking and uninterrupted recording of hours of animal behavior, and algorithms can now
137 parse those recordings to find behavioral patterns that may not be detected by humans manually tracking
138 behavior. Training on multiple animal model datasets and breaking the animal’s movements down into
139 behavioral components can enable deep learning algorithms to identify patterns of meaningful activity. These
140 tools – software such as Bonsai, SLEAP, DeepLabCut, Lightning Pose, and others – can even be tailored to
141 specific animals, such *Drosophila* (BonFly Neurogears 2023) and macaques (Bala et al. 2020, Labuguen et al.
142 2021; OpenMonkeyStudio and MacaquePose). By automating behavioral analysis (Datta et al. 2019), these
143 techniques can move past human constraints to test multiple animals at once and test more animals overall.

144 The position that *naturalistic studies are privileged views on behavior* sets up a tension between ‘wild’
145 behavior and laboratory ‘controlled’ behavior (see Figure 1). However, this tension alone fails to distinguish
146 the various experiments, as many studies may both impart control and capture wild behavior without there
147 being any meaningful way of distinguishing which one is more ‘naturalistic’ than the other (Fig. 1). A
148 framework that simply juxtaposes ‘wild’ and ‘controlled’ behavior also assumes the concept of naturalistic
149 behavior to be fixed, even in the face of the rapid technological progress that shapes what ‘naturalistic’ means.
150 For instance, what is ‘naturalistic’ today may no longer qualify as naturalistic once a new and more improved
151 technique for studying wild behavior in a controlled setting arrives (Fig. 1). This demonstrates the reliance of
152 ‘naturalistic’ on the different kinds of techniques that exist in relation to one another.

153 The current framework for naturalistic neuroscience relies on a concept of ‘natural’ that is both intuitive, and
154 yet vacuous upon investigation. Consider two widely used animals in neuroscientific research: the rat and the
155 fruit fly. One of the reasons the rat is used as a behavioral model is due to its high adaptability, but this fact
156 also blurs where to designate its natural habitat. Rats may have originated from specific parts of the world, but
157 they have long populated every continent thanks to human migration and now live among humans in urban
158 environments. Which of these histories count when one tries to measure rat behavior ‘in the wild’? As I will
159 discuss with the fruit fly, these histories become even more complicated as the domestication of animals
160 continue both inside and outside of the lab.

161 The overly broad category of ‘naturalistic’ precludes answering many crucial questions, including:
162 Does naturalistic behavior only apply to traditional animal models (*Aplysia*, flies, worms, rodents)? How does
163 one study naturalistic behavior in ‘wild types’ that no longer share behaviors with their own species? Is
164 viewing behavior from an evolutionary standpoint more naturalistic than from an ecological one? How much
165 is too much data for a behavioral study? What is the relationship between the ‘observer’ of behavior and the
166 behavioral subject? What defines behavior in experiments?

167 The intuition of ‘naturalistic’ rests on its use as a marker to distinguish what is ‘good’ from what is
168 artificial and ‘bad’. Moreover, the associations of ‘nature’ as pure and unbiased, and what keeps this category
169 distinctly separate from humans, is both socially and politically reinforced (Uggla 2010). This invokes the
170 core issue that ‘natural’ is a normative placeholder for different values and experimental aims. In the next
171 section, I shift the discussion to these aims in effort to elucidate some of the methodological complexities that
172 have been overlooked by mainstream discussions of naturalistic neuroscientific studies.

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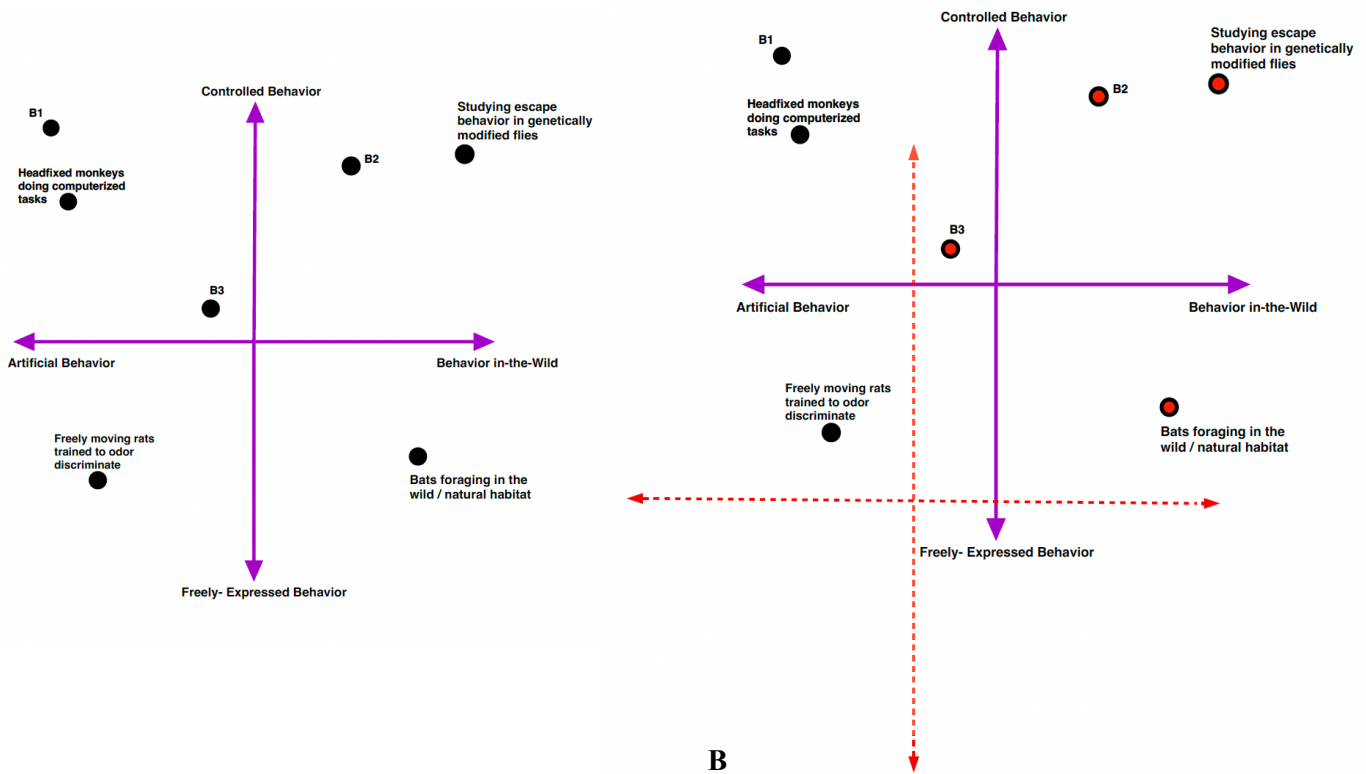


Figure 1: (a) Naturalistic is Too Broad Descriptions of naturalistic neuroscience practices are conceptually bound by attempts to get at a more objective characterization of behavior, but they also demonstrate the broadness of the term ‘naturalistic’. There may be two behaviors (B1 and B2) that appear separate on the naturalistic scale but are in fact conceptually similar to one another. **(b) Naturalistic is not a Fixed Concept** The conceptual landscape of naturalistic behavior can easily shift depending on technological innovation. Consider that two naturalistic experiments may, to varying degrees, capture ‘in the wild’ animal behavior while also using technology that offers those researchers fine grained control over their subject’s movements. For example, head fixed flies walking in 2D virtual landscapes vs. freely moving flies using virtual landscapes. The first experiment may be considered less naturalistic and more controlled than the second because, even if the fly is moving in a simulated reality, it seemingly has a more authentic experience compared to a head fixed fly. But now imagine that even a minor development in the technological capacities for measurement – such as more robust tools for measuring untethered flies in 3D environments – can change the understanding of ‘naturalistic’ between these two studies (represented by the red dotted line). What, then, becomes the dividing line on what ‘counts’ as naturalistic? If the tools and models were to change in the future, ‘naturalistic’ changes as well.

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193 **Experimental Aims and Epistemic Commitments**

194 To better explain why a study is considered naturalistic and how various naturalistic studies compare to one
195 another, I propose examining the various ways that scientists impart ‘control’ in the study of behavior. What
196 is ‘naturalistic’ changes depending on the various experimental aims of scientists, as these indicate differing
197 epistemic commitments. Epistemic commitments reveal what the concept of naturalistic is *relative to* by
198 clarifying one’s theoretical positioning, or what principles one is justifying in believing. For example, a
199 neuroscientist may want naturalistic studies because of their systems-level experimental aim of identifying an
200 important circuit, showing an epistemic commitment to a principle of reductionism. Another researcher,
201 having the experimental aim of using a nontraditional animal model for behavior, may latch onto an epistemic
202 commitment about development and the scope of behavioral flexibility.

203 Some clarity could be added here by identifying at least some of the differing epistemic commitments
204 imparted within naturalistic studies. Doing so can align researchers’ interests to make the ‘naturalistic’
205 concept more meaningful. It can also support non-intuitive positions, such as defending the use of traditional
206 models for many naturalistic behavioral experiments, as well as tempering the hype around new techniques
207 for naturalistic studies of behavior. In this respect, any discussion of naturalistic neuroscience demands a
208 follow up question: ‘naturalistic relative to what?’

209 Below, I illustrate how, in comparison to the conventional framework of naturalistic neuroscience that simply
210 juxtaposes ‘wild’ and ‘controlled’ behavior, a focus on epistemic aims provides a better way forward for
211 discussing naturalistic studies. Although it is beyond the scope of this review to discuss many naturalistic
212 neuroscience studies currently proliferating in the neurosciences, I show the complicated sides to the story of
213 studying naturalistic behavior by centering nonhuman animal modeling research. Moreover, this case alone,
214 exhibiting the range of epistemic issues that can emerge for those working in similar areas, reveals the
215 complexity of the problem as it scales to comparing research across domains.

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217 **Models are Mediated: Myth of the ‘Wild’ Model (overlooking the complexity of experimentation)**

218 A common point of discussion within the conventional characterization of naturalistic neuroscience,
219 particularly as it relates to nonhuman animal models, is a concern with repeatable behaviors and the use of
220 overtrained, traditional models. Traditionally, repeatable behaviors have been key to experimental research in
221 that they allow stable correlations between behavior and brain activity. Repeatability helps scientists compare
222 behavior between animals, control contextual variables that may influence behaviors, and even identify when
223 meaningful changes occur in the behavior of a single animal. Researchers have long identified ways to exploit
224 systems that exhibit repeatable behaviors (e.g. bar pressing or birdsong) or create conditions to make it more
225 likely that an animal will exhibit such behaviors.

226 Yet, those appealing to naturalistic neuroscience criticize the highly contrived situations that enable
227 repeatability. For example, ‘captivity effects’, or the behavioral and physiological changes generated by

228 housing animals in confined spaces, have been identified in numerous ways, ranging from genetic expression
229 across the brain (Bedoya Duque et al. 2023) to changes in hippocampal (LaDage et al. 2009) and cortical
230 volume (Bedoya Duque et al. 2023). Simply changing the environment that animals are normally housed in
231 leads to behavioral changes that relate directly to health and fitness of the animal (Vogt et al. 2024), including
232 relieving animals of stress that can affect experimental outcomes. These documented individual differences
233 between animals, such as in reward sensitivity, can also affect how one designs and uses behavioral assays. In
234 an extensive review on this matter, researchers document changes in brain morphology and function in many
235 animals – from chickadees and sparrows to mice and rats – being kept in laboratory conditions as opposed to
236 more enriched environments (Calisi and Bentley 2009). Even simple engagement with laboratory animals can
237 influence an animal’s behavior. Familiarity with a researcher can affect the performance of an animal in
238 particular cognitive tasks, as shown in ravens and crows (Cibulski et al. 2014).

239 Repeatable behaviors can be artifacts of overly controlled conditions and are often cited by
240 naturalistic neuroscientists as a reason for enriching an animals’ environment during behavioral testing
241 (Kentner et al. 2021). Under the conventional conceptual framework, housing in a more enriched condition
242 eliminates the distortions experimental control brings to bear on a study, again assuming that there is a
243 privileged view on behavior. For a model to achieve phenomenal access, or access to a behavioral
244 phenomenon of interest (Dietrich et al. 2020), it is assumed that the scientist simply ought to remove the
245 barriers of experimentation that can produce false behaviors, while still maintaining the standards of
246 laboratory control that can help measure them.

247 Unfortunately, such assumptions often overlook the complexities of experimental methods and the
248 different histories of lab animals. Specifically, standardized and repeatable behaviors are often generated from
249 research models that are engineered as laboratory *tools*, whose status as a tool enables phenomenal access. A
250 canonical example of this is studying naturalistic behavior in *Drosophila melanogaster*, or the common fruit
251 fly, whose transformation to a ‘standardized lab model’ was famously documented in research (Kohler 1994).
252 Wild-type controls were gradually and systematically modified over time, becoming a laboratory
253 domesticated wild-type. These changes are so well-known among fly researchers that those who use the
254 model admit to how little is known about variable behavior in the many strains of ‘wild type’ flies (Soto-
255 Yéber et al. 2018, Kaun Lab) and surprisingly little is even known about *Drosophila melanogaster’s* natural
256 environment at all (Asinof and Card 2024).

257 The current characterization of naturalistic neuroscience would perhaps warn us that the failure to
258 enact a model that accurately captures wild behavior is precisely the problem. And yet, as naturalistic
259 neuroscientists put emphasis on phenomenal access, they overlook the process of how phenomenal access is
260 achieved in the first place. Returning to artificially contrived behavioral studies may give us an answer, as we
261 attend to the various experimental aims for using the model, as well as researchers’ epistemic commitments.

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263 Consider the following quote from a group engaged in fruit fly research:

264 *Often the wild-type strain we use is dependent on the lab we were trained in.*
265 *My recommendation is to test a bunch of the common ones in your assay*
266 *and pick the one that produces the most consistent behavior. Use this for*
267 *your background strain (i.e. the strain to which you backcross / outcross all*
268 *of your transgenic / mutant lines). If you're not sure where to start, get the*
269 *most commonly used line from a lab that does the type of behavior you are*
270 *interested in (Kaun Lab)*

271 Through the conventional naturalistic framework, one might characterize this advice as a failure of naturalistic
272 experimentation since each ‘control’ animal seemingly distances the researcher from understanding ‘true’ or
273 ‘real’ *Drosophila* behavior. Note, however, that inconsistency in the model would be just as useless as an
274 animal that is *only* consistent, as the former would make a study intractable, and the latter would no longer
275 impart new information. All neuroscientists, naturalistic ones included, instead control their models in ways
276 that support their many experimental aims and epistemic commitments, even when their models are highly
277 mediated. Here, the ‘real’ behavior does not precede the methods used to control behavior, but rather comes
278 *through* it, as it enables researchers to observe variability in the first place. It is not only futile to attempt to
279 capture behavior ‘in the wild’ with a model organism, but the way observation of behavior is mediated also
280 makes it the case that any effort to capture ‘real’ behavior ignores the process of mediation.

281 **Why specific changes in a behavioral study qualify as ‘Naturalistic’**

282 Scientists constantly gain knowledge by using misbehaving models. Moreover, they continue to gain
283 meaningful information about naturalistic behavior through canonical models notoriously removed from the
284 ‘real’ world, such as the fruit fly, zebrafish, and rodents (Orger and Polavieja 2017; Dennis et al. 2021). Many
285 ‘naturalistic’ studies use *Drosophila melanogaster* (Vanin et al. 2012; Asinof and Card 2024) even though
286 there is little guarantee that the fly in the wild would even exhibit similar behavior. There are well-described
287 pattern generating circuits that elicit different behaviors across contexts, variable signaling processes that are
288 involved when flies engage in cooperative behavior, and distinct escape responses when stimuli presented to
289 flies vary in speed. What is notably prioritized in such cases is the stability of behavior over changing
290 contexts; thus, all of these discoveries, naturalistic in their own right, were made by making choices about
291 which variables could be ignored or focused on.

292 Here, the organism’s history, including the ways experimenters themselves made choices about
293 which wild type to choose, help render reliable results. Researchers simply use models and engage in
294 naturalistic studies to fit their aims in certain dimensions over others, such as prioritizing predicting behaviors
295 versus the discovery of new patterns associated with behaviors. One non-intuitive outcome of this observation
296 of experimentation is that greater experience with the animal and experimental conditions imparts knowledge
297 over time that is in fact relevant for modeling decisions in naturalistic neuroscientific studies.

298 In neuroscience, experimental aims can range from caring about naturalistic studies for the discovery of neural
299 circuits to furthering a technological end (see Box 1). Studying deviations from a standardized system – an
300 animal that is an engineered tool – can give generalizable information. This might be relevant to the organism,
301 but it might also go beyond knowing about the specific animal itself, since the experimental ends can vary
302 widely: testing the behavioral flexibility of a model, trying to further standardize the model, or knowing what
303 generalizable principles can be extracted. Such is often the case for naturalistic studies in *Drosophila*, where
304 the stated goal is better knowing what the animal would do ‘in the wild’ while the researchers test the
305 animal’s behavioral flexibility to get more generalizable principles. One recent example of this this strategy
306 has been coined ‘reverse neuroethology’ (Asinof and Card 2024). Here, researchers intentionally choose a
307 highly modeled and controlled system, as this grants them better ways of leveraging the study of its natural
308 behavior. In such cases, naturalistic behavior is paired with methodological development to become a vehicle
309 for capturing criteria that can be applied across species, as opposed to being of interest in an undirected
310 manner.

311 Two studies of social behavior can direct the idea of ‘control’ in very different ways, such that both
312 are labeled ‘naturalistic’ but are not considered naturalistic neuroscience to the different groups studying
313 them. Consider the simple change of moving rodents out of confined environments and into significantly
314 larger arenas. Placing rats in large arenas can still fail to meet the naturalistic ‘ideal’ because these animals are
315 behaviorally modified due to their selective breeding (Kondrakiewicz et al. 2018). Even so, not all naturalistic
316 experiments are directed similarly. While a larger arena certainly matters for a range of experimental goals –
317 for example, mechanistically examining system-wide brain activity and knowing how different brain regions
318 may interact or behaviorally understanding the social transmission of fear – expanding the parameters of
319 activity too much can generate worse results if one wants to know how odor cues are socially relayed (Datta et
320 al. 2019) or whether or not exposure to predator odors in early life can affect later behaviors (St-Cyr et al.
321 2018). Having more data in the latter cases will not improving behavioral knowledge. Thus, naturalistic
322 neuroscientists consistently make choices on what kinds of control is appropriate to leverage for one’s
323 experimental goals, given they are always navigating the limitations of their models.

324 To put it simply, there can be different experimental goals under the same heading of naturalistic
325 studies. Is the researcher trying to better understand the general mechanisms of a behavior (i.e., systems
326 involved in survival-critical decisions or what kinds of neurons generate courtship behaviors) or is she trying
327 to understand the ability for the animal to adapt in various conditions (to better understand sensory cue
328 integration or changes in neuronal sensitization)? Is she trying to model the animal into a tool that is
329 comparable to another for future neuroscientific study? Is the researcher trying to study repeatable behavior to
330 link their results to another kind of experiment or another species? Although all of these aims might be
331 considered ‘naturalistic’ relative to a previous practice, they share very little in common beyond the label.

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333 **Replacing Traditional Models with Krogh Animals (Negative Models)**

334 In discussions of naturalistic neuroscience, one can improve a behavioral assay for a more canonical model or
335 re- theorize an experimental approach to modeling altogether by shifting away from the use of canonical
336 animal *models* – such as flies, rats, and mice – more generally. In this vein, researchers have warned against
337 conflating model organisms with natural, unmodified organisms (Katz 2016), and many doing naturalistic
338 studies have rethought model choice, encouraging the use of nontraditional animals instead of lab-mediated
339 ones (Mathuru 2020; Stevenson et al. 2017; Testard et al. 2021; Yartsev 2017; Jourjine and Hokstra 2021).
340 This point of view recapitulates conventional framework which assumes that attempts to control and model
341 animals pushes in the opposite direction of what is ‘wild’ behavior, since it remains to be seen that behaviors
342 from generations of genetically engineered animals would replicate in the ‘real world’ (Vanin 2012, Crabbe et
343 al. 1999).

344 In such discussions, Krogh’s principle is often popularly invoked by neuroscientists seeking more
345 naturalistic behaviors (Stevenson 2018; Katz 2016). This principle states that for any biological question,
346 there is an organism whose biology is uniquely suited to answering that question.

347 Examples of Krogh’s principle include using a squid to study potentiation because they have a giant axon
348 (Yartsev 2017) or using mice to study olfaction because that is the animals’ preferred means of sensory
349 navigation, as they learn with smell better than auditory and visual stimuli (Nigrosh et al. 1975). It can also
350 involve using other species of traditional models, such as fish and mice that are more specialized to certain
351 tasks – such as aggression in Siamese fighting fish, parental care in deer mice (Bendesky et al. 2017; Jourjine
352 and Hoekstra 2021) – to leveraging the loss of an ability, such as sine song, in *Drosophila yakuba* (Kelley
353 2024; Ye et al. 2024). In all these cases, a distinct feature of the animal makes it experimentally useful in a
354 way that advantages it in comparison to others.

355 In contrast to Krogh’s principle, animals may be chosen for study on the basis of ‘practical’ reasons,
356 such as the availability of experimental tools for that model, as well as the logistical conveniences specific
357 models themselves afford. For example, there are many species of fruit flies, but a vast majority of
358 neuroscience research focuses on *melanogaster* because of the availability of lines and reporters. Scientists
359 may also choose this model because of low costs, ease of supply, husbandry, established communities
360 (conferences centered on specific models), databases (FlyBase, Xenbase, WormBook), ease of replicability,
361 and more (Leonelli and Ankeny, Dietrich et al. 2020; Ding et al. 2024; Zilkha et al. 2016). One cause for
362 concern is that these reasons can often trump others when deciding which model system to use for an
363 experiment. For example, there is currently an overwhelming use of mouse and rat models in neuroscientific
364 research, which has limited the range of the kinds of nervous systems studied (Yartsev 2017, Juntti 2019).
365 However, in its criticism of practical models, the traditional conceptual framework for naturalistic behavior
366 continues to fall short by suggesting that the use of nonconventional models somehow leads to a more

367 accurate behavioral readout. Even in the case of Krogh organisms, different epistemic aims can arise that need
368 to be specified.

369 To help with categorization and comparison, neuroscientists may appeal to naturalistic studies of
370 behavior to study biological diversity, *splitting* organisms into their differences, or they can identify the
371 common mechanisms and patterns by lumping animals together. In a historical review of Krogh's principle,
372 researchers instead show that Krogh organisms do not depend on their generalizability (Green et al. 2020, 4).
373 Unlike standard models whose similarities and differences to other systems are known, the representational
374 scope of the Krogh animal is itself an empirical question. Because of this, it is difficult to know if the
375 identified traits are generalizable or even relevant to other species.

376 When using Krogh systems, researchers may be less concerned with control for generalization and
377 more interested in control with respect to behavioral flexibility. For example, by having extreme adaptations,
378 Krogh animals often serve other experimental goals, such as helping scientists explore variation over
379 identified physiological features (Green et al. 2020, 8), or by as serving as 'negative' models where animals
380 *lack* the specific features or behaviors that scientists are interested in studying (Green et al. 2020). An
381 octopus, an asocial creature, could be used to study sociality (Edsinger and Dölen 2018), or a naked mole rat,
382 that is cancer resistant, to study anticancer mechanisms (Tian et al. 2013). Serving as comparison cases with
383 positive models, such animals can impart invaluable information about the scope of behavioral flexibility, the
384 importance of environment, energy expenditure, and more. Knowing why certain physiological limitations are
385 *not* observed in selected species can offer invaluable insights.

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388 **Box 1: Epistemic Commitments underlying Various Experimental Aims**

389

Various Experimental Goals with Respect to Behavior
<p>Engineering aims: Knowing how context plays a role with respect to behavior and using that to elicit certain responses. Examples of this include leveraging the methodological development and manipulability of model organisms (genetic or otherwise).</p> <p>Example: This approach includes strategies for ‘reverse neuro-ethology’ (Asinof and Card 2024).</p>
<p>Technological aims: may not be directly tied to understanding the organism itself, but rather finding common motifs that can be exploited in other contexts (such as computational principles); here, naturalistic behaviors are leveraged as reasons for creating models in the first place so they can be meaningfully compared to other models.</p> <p>Example: Recognizing animal models as tools and as systems that are represented, some scientists interested in naturalistic behaviors insist on developing nontraditional animals into animal <i>models</i>. Examples of this can include taking individualized approaches to increase genetic accessibility in specific animals, such as using adeno-associated viral tools in amphibians (Jaeger et al. 2024), to large-scale, community building efforts, such as “cephalopod-omics,” which tries to apply a conglomerate of techniques normally used on invertebrates, such as sequencing, imaging, and genetic manipulations, to invertebrates (Baden et al. 2024). These studies fall under the scope of naturalistic in their commitment to expanding the use of animals for research, even though the idea is to use these tools to make behaviors more meaningfully comparable, such studies introduce an intermediary for comparison as opposed to directly observing behavioral similarities or differences. It is thought that by doing so, experimentalists open the range of models that can be studied in neuroscience as well (Juntti 2019). Doing so may generate models of abstraction that make the models comparable in the first place.</p>
<p>Aim of exploiting variability: Finding out about behavior in order to derive principles that can be applied to a different system; this is not the same as finding generalizable principles, but instead exploiting knowledge about adaptation, aspects of variability, or extreme behavior.</p> <p>Example: Krogh’s principle; researchers can look to naturalistic behaviors that are both extreme behaviors, or even the lack of behaviors (what is not there) to better understand the conceptual potentials and limits of behavior.</p>

390

391 **Consequences of Adopting a Conventional Framework:**

392 Finally, failing to characterize what is ‘naturalistic’ without attention to experimental aims is a disregard to an
 393 experimenter’s conceptual commitments. These can range from methodological commitments - such as
 394 determining if one should be in a field or the lab – to decisions about what scientific tradition one should draw
 395 from.

396 For instance, naturalistic neuroscience appeals to a large range of fields for inspiration, including
 397 neuroethology, ethology, behavioral ecology, comparative neuroscience, evolutionary neuroscience, and
 398 evolutionary biology. Here, naturalistic neuroscientists claim it is important to recognize a range of views:
 399 recognizing behavior as evolved (Krakauer et al. 2017; Miller et al. 2023; Testard et al. 2021; Datta et al.
 400 2019), acknowledging the role of non-neuronal processes in supporting the complexity of behavior (MacIver
 401 2009), the importance of fewer constraints on experiments, such as letting an animal run around freely
 402 (Gomez-Marin et al. 2014; Brown and de Bivort 2018; Parker et al. 2022), identifying innate behaviors

403 (Gomez-Marin et al. 2014), and identifying adaptation and selection pressures (Mobbs et al. 2018; Miller et
404 al. 2023; Testard et al. 2021; Mobbs et al. 2018).

405 The traditions that inspire these activities tend to be cited interchangeably when naturalistic behavior
406 is invoked. Some appeal to evolutionary biology to think about evolved behavior; others look to ethology for
407 methodological inspiration, such as conducting experiments within an open field and uncontrolled
408 environment. However, one challenge to understanding naturalistic behavior from various traditions is that
409 they crucially differ in conceptual commitments, research cultures, and topics of investigation, varying also in
410 their sensitivity to both context and evolutionary and developmental concerns. From traditions like ethology,
411 evolutionary biology, and behavioral ecology alone, a spectrum of views arises that are bookended by two
412 extremes. On one end are propositions to ignore all preceding paradigms used to study behavior; for example,
413 those keen on an ecological tradition may not see value in studying nonhuman systems. Here, they can
414 recommend abolishing all nonhuman animal studies when making attempts to study human behavior. On the
415 other end, scientists see naturalistic studies of behavior as having natural continuity with traditional behavioral
416 paradigms. Although they draw inspiration from ecological experiments, they can propose the status quo with
417 minor adjustments here and there to make a study appear more naturalistic.

418 This picture gets even more complicated given that, even within a *single* tradition, there have been major
419 historical disagreements about what behaviors are relevant to investigate and complicated discussions about
420 how one should investigate them. This has certainly been the case in ethology and neuroethology (Dhein
421 2022). A second challenge, then, involves the difficulty of knowing which practices one should prioritize if
422 the experimental aims differ by discipline or research traditions.

423 Ethologists, for example, have traditionally taken interest in the behaviors they observe in animals'
424 respective ecologies and habitats. This means studying escape, food seeking, recognizing prey, and other
425 practices that are relevant to the day-to-day challenges and survival of the animal. This contrasts with the vast
426 repertoire of behaviors studied in neuroscience, such as drug-seeking behavior, play behavior, sociality,
427 novelty-seeking, binge-eating, compulsive-like, anxiety-like or depression-like behaviors, and so on, that are
428 set by the interests of humans and therefore been imposed on animals that have been highly manipulated and
429 controlled. How can one meaningfully draw from ethology in neuroscience when their goals for studying
430 behavior differ?

431 Similar to the challenges associated with animal models and behavioral testing, there will be
432 disagreement depending on which tradition one draws from. However, in addition to having various
433 experimental aims, there can be epistemic differences in theoretical commitments as well. This generates
434 conflict with the concept of 'naturalistic' insofar as two groups could be opposing each other's understanding
435 of naturalistic.

436 Another salient example of when differences in 'tradition' matter is with respect to
437 representationalist commitments. There are dramatic consequences if one ascribes to a literal interpretation of

438 brain patterns representing the ‘real’ behavior or if one is simply using such language heuristically. These
 439 differences separate how research can link across different practices with researchers not even recognizing
 440 when this is the case. Consider a naturalistic researcher who uses context to re-conceptualize olfactory
 441 behavior as extended and environmentally embedded (Jacobs 2023), whereas another sees olfaction as
 442 represented or mapped ‘in the brain’ (Brann and Datta 2023). While both may be committed to more
 443 naturalistic studies, their attitudes about what is in fact going on neuronally may be radically different, with
 444 one putting more of an emphasis on environment and embodied behavioral approaches, while the other tries to
 445 articulate the ‘olfactory code’ from a purely computational point-of-view.

446 **Box 2: Animal Behavior Under Various ‘Traditions’ and Explanations**
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<p>Ethological: A behavior-based science that is in the business of observing animal behavior and explaining what the animal is doing for the animal itself</p> <ul style="list-style-type: none"> - Preferred Explanations: ethologists had strong commitments to ‘innate’ mechanisms, acquired releasing mechanisms, imprinting, drive intensity, fixed action patterns (escape response from Mollusk <i>Tritonia</i>), and more. - Ethological explanations made use of physiological information but did not reduce behavior to them, nor localize function to specific areas of the body - Especially interested in reproductive behavior - Self-described “animal watchers” (Tinbergen) - Field-based research
<p>Neuroethology: Study of how animal behavior is realized by the central nervous system.</p> <ul style="list-style-type: none"> - Preferred Explanations: Delineating what capacities a brain should have to realize the complexity of animal behaviors. Some neuroethologists claim that the nervous system evolved to produce behavior (Camhi 1984). - Interested in the comparative physiology of behavior (principles of neural function are studied in various animals) - How does the nervous system solve specific problems; these can include mating (such as a female sparrow or frog detecting, discriminating, and orienting toward a male call), escape responses, how animals use light to seek food, shelter, detect predators, or orient for navigation, and prey catching (frogs). - Going between ‘field’ and ‘lab’ sites
<p>Behavioral Ecology: Studies that try to delineate the ecological factors that can drive behavioral adaptations.</p> <ul style="list-style-type: none"> - Preferred explanations are rooted in evolutionary principles; seen as a ‘successor’ to ethology with less demand for theoretical coherence - Population dynamics and models; examining the genetic basis of behavior, behavioral syndromes, - Topics can include parental care - Field-based research
<p>Evolutionary Biology (of Behavior):</p> <ul style="list-style-type: none"> - Preferred explanation: Ultimate source of explanation is natural selection - Interested in fitness, selection (sexual selection), variation, and retention - Looking at behavior in groups, not just individual behavior - Interested in genomic changes associated with behavioral differences - Field or lab-based
<p>Computational analysis of behavior (Computational ethology/ computational ecology):</p> <ul style="list-style-type: none"> - Preferred explanation is in the language of computation and conceiving of behavior as information processing - Engaged in simulations of behavior; prioritizes prediction - Interested in substituting human decisions on behavioral motifs with computerized detections - Lab-based research; drawing from datasets

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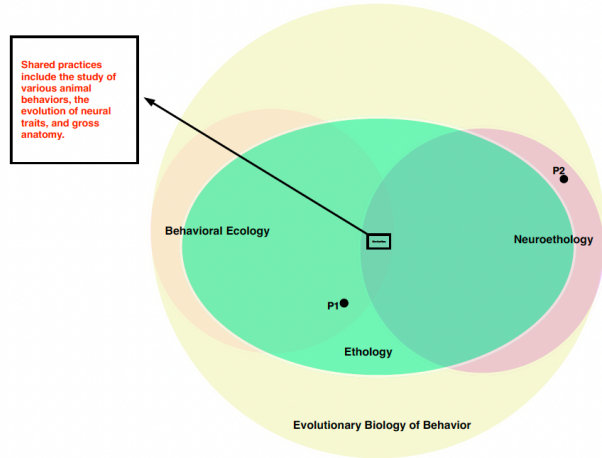


Figure 2: Although many of these traditions share overlap, their explanatory aims can differ in important respects. For example, while all these traditions notably study natural behaviors such as threat detection or foraging, the representationalist commitments of neuroethologists who subscribe to innate behaviors are going to look radically different from those of neuroethologists. The scale of explanation will also differ between them, such as an explanation that meets the criteria for overlapping traditions (P1) but fails to satisfy assumptions within another (P2).

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467

Keeping Mediated Models

468 The animal models, tools, technologies, and traditions that make up the landscape of naturalistic neuroscience
469 have generated conversations that overshadow the theoretical commitments and goals that inform how those
470 technologies are supposed to meet one’s modeling aims.

471 This piece tempers enthusiasm for the conventional framework of naturalistic studies by challenging the idea
472 that capturing ‘real’ behavior is hindered by the traditional strategies for experimental control. The
473 conventional framework that describes naturalistic behavior reduces away the many theoretical or epistemic
474 commitments that underlie behavioral studies. One might try to understand how an animal’s behavior changes
475 in new contexts and meet other demands, such as knowing how animals compare to one another, or how to
476 make animals comparable in the first place. Many of these strategies leverage control to fit specific goals,
477 undermining the idea that there is any single *privileged* approach to study behavior.

478 However, recognizing how ‘natural’ is used as a stand-in term for various experimental aims corrects
479 perspectives on experimentation in more ways than one. For example, neglecting assumptions of objectivity
480 have historically led researchers to harmful outcomes. The term ‘natural’ has uniquely been a social signifier
481 of exclusion with roots in scientific studies, whether in antiquated discussions about assertive, ambitious
482 women, or more recent history of homosexual behavior. These studies may seem unrelated to the current topic
483 because naturalistic studies are trying to dispense with bias to reveal ‘true’ behavioral patterns. And yet, the
484 patterned use of ‘un/natural’ is the same: In the same way that labeling homosexuality as ‘unnatural’ because
485 of biological differences was about hidden values, labeling a bar pressing rats ‘unnatural’ because of its
486 differences to wild type rats is about hidden epistemic aims. Failing to acknowledge these aims and the role of
487 the experimenters behind them irresponsibly masks the differences with the word ‘natural’.

488 However, even if there *were* an option to do naturalistic neuroscience in a way that abolished the scientist as a
489 mediator of observation, no one should want that. For one, it would be experimentally intractable and generate

490 unintelligible results. Although it is true that humans do not use sonar or electroreception to locate their food,
491 use smell or magnetoreception to navigate, or see with polarization or infrared, we came to know most of
492 these differences in other animals from previous empirical research. This demonstrates an irony about the
493 study of naturalistic behavior, where much of the reason for knowing that studying ‘naturalistic’ behavior is
494 preferable to behavior confined and controlled environments precisely emerges from the fact that those
495 behaviors were first studied in controlled settings (Clarkson et al. 2018). However, a second reason we should
496 not eliminate the human observer is that it makes empirical observations of the world intelligible *to us*. It is
497 our experiences that give us the capacities for modeling decisions in neuroscientific experiments (Nemati
498 2024). Part of the reason for this predicament is that neuroscience is a science that inherently relies on
499 modeling and abstracting from complexity to proffer appropriate explanations (Chirimuuta 2024). Unlike
500 some classic ethological practices that were simply in the business of documenting animal activities, there are
501 different requirements for the kinds of mechanistic and causal explanations neuroscience should be giving us.
502 Moving forward, it would benefit us to take a historical lens to the technological improvements that have
503 made naturalistic studies of behavior possible, as well as the theoretical assumptions embedded in them. We
504 now raise many questions about behavior because it is now possible to capture the dynamic and multi-
505 dimensional features of an environment and of brain activity. Shifts from traditional views of brain modularity
506 (Anderson 2021) have allowed scientists to favor probabilistic distributions of neural and behavioral activity
507 that rely on more neural data and population-level activity over linear statistical models of discrete variables
508 (Brown et al. 2004; Cunningham and Yu 2014; Pang et al. 2016). Dynamical and adaptive thinking (Fairhall
509 et al. 2001) emphasizes the changing brain, encouraging the study of the brain’s robust plasticity (Gomez-
510 Marin 2014). Neuroscientists can also now simultaneously record the activity of very large numbers of
511 neurons, from many different brain regions, as the animal engages in a specific task (Neuropixels), enabling
512 analyses of neural networks (Bassett and Sporns 2017; Bassett et al. 2018) while powering dynamical
513 explanations and use of dynamical systems theory itself (Izhikevich 2007; McClelland et al. 2010; Ross
514 2022). Finally, improved devices, such as better GPS technology, microphone arrays, motion sensors, and
515 sophisticated cameras, now capture complex behavior with more storage space to do it.

516 Twenty years in the making, these shifts share a heightened regard for behaviorally relevant naturalistic and
517 environmental factors. Yet, understanding how these tools embed their own assumptions is important for
518 knowing how certain studies of behavior are privileged over others.

519 While much work remains to be done to show how modeling aims link to experimental outcomes, talking past
520 one another can have more serious epistemic consequences for experimentation, such as when experimenters
521 put focus on experimental choices when their underlying assumptions do in fact differ. Not recognizing that
522 we are asking different questions may generate miscommunication and the illusion of reproducibility errors,
523 as was the case in two similar odor studies that got varied mechanistic explanations on the basis of different
524 tasks (Federick et al. 2017). It also bears consequences by distracting researchers when there are quick

525 explanations for why an experiment fails. Rather than adopt a seemingly better model or task, or using a more
526 advanced tool to measure behavior, as has often been suggested, it may do behavioral neuroscience good to
527 accept that not all experimental goals require such approaches.

References

- Anderson, D. J., & Perona, P. (2014). Toward a Science of Computational Ethology. *Neuron*, 84(1), 18–31. <https://doi.org/10.1016/j.neuron.2014.09.005>
- Anderson, M. L. (2021). *After phrenology: Neural reuse and the interactive brain*. MIT Press.
- Asinof, S. K., & Card, G. M. (2024). *Neural Control of Naturalistic Behavior Choices*. <https://doi.org/10.1146/annurev-neuro-111020-094019>
- Bala, P. C., Eisenreich, B. R., Yoo, S. B. M., Hayden, B. Y., Park, H. S., & Zimmermann, J. (2020). Automated markerless pose estimation in freely moving macaques with OpenMonkeyStudio. *Nature Communications*, 11(1), 4560. <https://doi.org/10.1038/s41467-020-18441-5>
- Bedoya Duque, M. A., Thomas, W. R., Dechmann, D. K., Nieland, J., Baldoni, C., von Elverfeldt, D., ... & Dávalos, L. M. (2023). Large captivity effect based on gene expression comparisons between captive and wild shrew brains. *bioRxiv*, 2023-10.
- Bassett, D. S., & Sporns, O. (2017). Network neuroscience. *Nature Neuroscience*, 20(3), 353–364. <https://doi.org/10.1038/nn.4502>
- Bassett, D. S., Zurn, P., & Gold, J. I. (2018). On the nature and use of models in network neuroscience. *Nature Reviews Neuroscience*, 19(9), 566–578. <https://doi.org/10.1038/s41583-018-0038-8>
- Bendesky, A., Kwon, Y. M., Lassance, J. M., Lewarch, C. L., Yao, S., Peterson, B. K., ... & Hoekstra, H. E. (2017). The genetic basis of parental care evolution in monogamous mice. *Nature*, 544(7651), 434-439. <https://doi.org/10.1038/nature22074>
- Berdahl, A., Torney, C. J., Ioannou, C. C., Faria, J. J., & Couzin, I. D. (2013). Emergent sensing of complex environments by mobile animal groups. *Science (New York, N.Y.)*, 339(6119), 574–576. <https://doi.org/10.1126/science.1225883>
- Brandl, H. B., Pruessner, J. C., & Farine, D. R. (2022). The social transmission of stress in animal collectives. *Proceedings of the Royal Society B: Biological Sciences*, 289(1974), 20212158. <https://doi.org/10.1098/rspb.2021.2158>
- Brown, E. N., Kass, R. E., & Mitra, P. P. (2004a). Multiple neural spike train data analysis: State-of-the-art and future challenges. *Nature Neuroscience*, 7(5), 456–461. <https://doi.org/10.1038/nn1228>
- Brann, D. H., & Datta, S. R. (2020). Finding the Brain in the Nose. *Annual Review of Neuroscience*, 43(Volume 43, 2020), 277–295. <https://doi.org/10.1146/annurev-neuro-102119-103452>

Brown, A. E. X., & de Bivort, B. (2018). Ethology as a physical science. *Nature Physics*, *14*(7), 653–657. <https://doi.org/10.1038/s41567-018-0093-0>

Calisi, R. M., & Bentley, G. E. (2009). Lab and field experiments: Are they the same animal? *Hormones and Behavior*, *56*(1), 1–10. <https://doi.org/10.1016/j.yhbeh.2009.02.010>

Camhi, J. M. (1984). Neuroethology: Nerve cells and the natural behavior of animals. *Sinauer Associates*.

Chirimuuta, M. (2022). Artifacts and levels of abstraction. *Frontiers in Ecology and Evolution*, *10*. <https://www.frontiersin.org/articles/10.3389/fevo.2022.952992>

Chirimuuta, M. (2024). *The Brain Abstracted: Simplification in the history and philosophy of neuroscience*. MIT Press.

Cibulski, L., Wascher, C. A. F., Weiß, B. M., & Kotrschal, K. (2014). Familiarity with the experimenter influences the performance of Common ravens (*Corvus corax*) and Carrion crows (*Corvus corone corone*) in cognitive tasks. *Behavioural Processes*, *103*, 129–137. <https://doi.org/10.1016/j.beproc.2013.11.013>

Cisek, P., & Green, A. M. (2024). Toward a neuroscience of natural behavior. *Current Opinion in Neurobiology*, *86*, 102859. <https://doi.org/10.1016/j.conb.2024.102859>

Clarkson, J. M., Dwyer, D. M., Flecknell, P. A., Leach, M. C., & Rowe, C. (2018). Handling method alters the hedonic value of reward in laboratory mice. *Scientific Reports*, *8*(1), 2448. <https://doi.org/10.1038/s41598-018-20716-3>

Crabbe, J. C., Wahlsten, D., & Dudek, B. C. (1999). Genetics of mouse behavior: Interactions with laboratory environment. *Science (New York, N.Y.)*, *284*(5420), 1670–1672. <https://doi.org/10.1126/science.284.5420.1670>

Cunningham, J. P., & Yu, B. M. (2014). Dimensionality reduction for large-scale neural recordings. *Nature Neuroscience*, *17*(11), 1500–1509. <https://doi.org/10.1038/nn.3776>

Cuthill, I. C., Hart, N. S., Partridge, J. C., Bennett, A. T. D., Hunt, S., & Church, S. C. (2000). Avian colour vision and avian video playback experiments. *Acta Ethologica*, *3*(1), 29–37. <https://doi.org/10.1007/s102110000027>

Datta, S. R., Anderson, D. J., Branson, K., Perona, P., & Leifer, A. (2019). Computational Neuroethology: A Call to Action. *Neuron*, *104*(1), 11–24. <https://doi.org/10.1016/j.neuron.2019.09.038>

Deacon, R. M. J. (2006). Housing, husbandry and handling of rodents for behavioral experiments. *Nature Protocols*, *1*(2), 936–946. <https://doi.org/10.1038/nprot.2006.120>

- Dennis, E. J., Hady, A. E., Michael, A., Clemens, A., Tervo, D. R. G., Voigts, J., & Datta, S. R. (2021a). Systems Neuroscience of Natural Behaviors in Rodents. *Journal of Neuroscience*, *41*(5), 911–919. <https://doi.org/10.1523/JNEUROSCI.1877-20.2020>
- Dhein, K. (2022). From Karl von Frisch to neuroethology: A methodological perspective on the Friscean tradition's expansion into neuroethology. *Berichte zur Wissenschaftsgeschichte*, *45*(1-2), 30-54.
- Dietrich, M. R., Ankeny, R. A., Crowe, N., Green, S., & Leonelli, S. (2020). How to choose your research organism. *Studies in History and Philosophy of Science Part C: Studies in History and Philosophy of Biological and Biomedical Sciences*, *80*, 101227.
- Dong, Z., Feng, Y., Diego, K., Baggetta, A. M., Sweis, B. M., Pennington, Z. T., Lamsifer, S. I., Zaki, Y., Sanguiliano, F., Philipsberg, P. A., Morales-Rodriguez, D., Kircher, D., Slesinger, P., Shuman, T., Aharoni, D., & Cai, D. J. (2024). *Simultaneous dual-color calcium imaging in freely-behaving mice* (p. 2024.07.03.601770). bioRxiv. <https://doi.org/10.1101/2024.07.03.601770>
- Edsinger, E., & Dölen, G. (2018). A conserved role for serotonergic neurotransmission in mediating social behavior in octopus. *Current Biology*, *28*(19), 3136-3142. doi: [10.1016/j.cub.2018.07.061](https://doi.org/10.1016/j.cub.2018.07.061).
- Fairhall, A. L., Lewen, G. D., Bialek, W., & de Ruyter van Steveninck, R. R. (2001). Efficiency and ambiguity in an adaptive neural code. *Nature*, *412*(6849), Article 6849. <https://doi.org/10.1038/35090500>
- Fan, S., Dal Monte, O., & Chang, S. W. C. (2021). Levels of naturalism in social neuroscience research. *iScience*, *24*(7), 102702. <https://doi.org/10.1016/j.isci.2021.102702>
- Frederick, D. E., Brown, A., Tacopina, S., Mehta, N., Vujovic, M., Brim, E., ... & Kay, L. M. (2017). Task-dependent behavioral dynamics make the case for temporal integration in multiple strategies during odor processing. *Journal of Neuroscience*, *37*(16), 4416-4426. <https://doi.org/10.1523/JNEUROSCI.1797-16.2017>
- Fleishman, L. J., McClintock, W. J., D'eath, R. B., Brainard, D. H., & Endler, J. A. (1998). Colour perception and the use of video playback experiments in animal behaviour. *Animal Behaviour*, *56*(4), 1035–1040. <https://doi.org/10.1006/anbe.1998.0894>
- Geissmann, Q., Beckwith, E. J., & Gilestro, G. F. (2019). Most sleep does not serve a vital function: Evidence from *Drosophila melanogaster*. *Science Advances*, *5*(2), eaau9253. <https://doi.org/10.1126/sciadv.aau9253>
- Gomez-Marin, A., & Ghazanfar, A. A. (2019). The Life of Behavior. *Neuron*, *104*(1), 25–36. <https://doi.org/10.1016/j.neuron.2019.09.017>

Gouveia, K., & Hurst, J. L. (2017). Optimising reliability of mouse performance in behavioural testing: The major role of non-aversive handling. *Scientific Reports*, 7, 44999. <https://doi.org/10.1038/srep44999>

Green, S., Dietrich, M. R., Leonelli, S., & Ankeny, R. A. (2018). 'Extreme' organisms and the problem of generalization: Interpreting the Krogh principle. *History and Philosophy of the Life Sciences*, 40(4), 65. <https://doi.org/10.1007/s40656-018-0231-0>

Hasani, H., Sun, J., Zhu, S. I., Rong, Q., Willomitzer, F., Amor, R., ... & Goodhill, G. J. (2023). Whole-brain imaging of freely-moving zebrafish. *Frontiers in Neuroscience*, 17, 1127574. <https://doi.org/10.3389/fnins.2023.1127574>

Hashimoto, M., Hata, A., Miyata, T., & Hirase, H. (2014). Programmable wireless light-emitting diode stimulator for chronic stimulation of optogenetic molecules in freely moving mice. *Neurophotonics*, 1(1), 011002. <https://doi.org/10.1117/1.NPh.1.1.011002>

Hoffmann, S., Beetz, M. J., Stöckl, A., & Mesce, K. A. (2023). Editorial: Naturalistic neuroscience — Towards a full cycle from lab to field. *Frontiers in Neural Circuits*, 17. <https://www.frontiersin.org/articles/10.3389/fncir.2023.1251771>

Hoyer, S. C., Eckart, A., Herrel, A., Zars, T., Fischer, S. A., Hardie, S. L., & Heisenberg, M. (2008). Octopamine in male aggression of *Drosophila*. *Current Biology: CB*, 18(3), 159–167. <https://doi.org/10.1016/j.cub.2007.12.052>

Huang, K.-H., Rupprecht, P., Frank, T., Kawakami, K., Bouwmeester, T., & Friedrich, R. W. (2020). A virtual reality system to analyze neural activity and behavior in adult zebrafish. *Nature Methods*, 17(3), 343–351. <https://doi.org/10.1038/s41592-020-0759-2>

Izhikevich, E. M. (2007). *Dynamical systems in neuroscience*. MIT press.

Jacobs, L. F. (2021). How the evolution of air breathing shaped hippocampal function. *Philosophical Transactions of the Royal Society B: Biological Sciences*, 377(1844), 20200532. <https://doi.org/10.1098/rstb.2020.0532>

Jaeger, E. C., Vijatovic, D., Deryckere, A., Zorin, N., Nguyen, A. L., Ivanian, G., ... & Sweeney, L. B. (2024). Adeno-Associated Viral Tools to Trace Neural Development and Connectivity Across Amphibians. *bioRxiv*, 2024-02.

Jahangiri, Z., Gholamnezhad, Z., Hosseini, M., Beheshti, F., & Kasraie, N. (2019). The effects of moderate exercise and overtraining on learning and memory, hippocampal inflammatory cytokine levels, and brain oxidative stress markers in rats. *The Journal of Physiological Sciences*, 69(6), 993–1004. <https://doi.org/10.1007/s12576-019-00719-z>

Jeung, S., Hilton, C., Berg, T., Gehrke, L., & Gramann, K. (2023). Virtual Reality for Spatial Navigation. In C. Maymon, G. Grimshaw, & Y. C. Wu (Eds.), *Virtual Reality in Behavioral*

Neuroscience: New Insights and Methods (pp. 103–129). Springer International Publishing.
https://doi.org/10.1007/7854_2022_403

Jourjine, N., & Hoekstra, H. E. (2021). Expanding evolutionary neuroscience: Insights from comparing variation in behavior. *Neuron*, *109*(7), 1084–1099.
<https://doi.org/10.1016/j.neuron.2021.02.002>

Juntti, S. (2019). The Future of Gene-Guided Neuroscience Research in Non-Traditional Model Organisms. *Brain Behavior and Evolution*, *93*(2–3), 108–121. <https://doi.org/10.1159/000500072>

Kaun Lab. (2016, November 16). *Which wild-type should I use?* Kaunlab.
<https://www.kaunlab.com/post/2016/11/16/which-wild-type-should-i-use>

Katz, P. S. (2016). ‘Model organisms’ in the light of evolution. *Current Biology*, *26*(14), R649–R650.

Kelley, D. B. (2024). Courtship behavior: Resurrecting an undead song. *Current Biology*, *34*(13), R623–R625.

Kendall-Bar, J. M., Williams, T. M., Mukherji, R., Lozano, D. A., Pitman, J. K., Holser, R. R., Keates, T., Beltran, R. S., Robinson, P. W., Crocker, D. E., Adachi, T., Lyamin, O. I., Vyssotski, A. L., & Costa, D. P. (2023). Brain activity of diving seals reveals short sleep cycles at depth. *Science*, *380*(6642), 260–265. <https://doi.org/10.1126/science.adf0566>

Kentner, A. C., Speno, A. V., Doucette, J., & Roderick, R. C. (2021). The Contribution of Environmental Enrichment to Phenotypic Variation in Mice and Rats. *eNeuro*, *8*(2).
<https://doi.org/10.1523/ENEURO.0539-20.2021>

Kim, T. I., McCall, J. G., Jung, Y. H., Huang, X., Siuda, E. R., Li, Y., ... & Bruchas, M. R. (2013). Injectable, cellular-scale optoelectronics with applications for wireless optogenetics. *Science*, *340*(6129), 211–216. DOI: [10.1126/science.1232437](https://doi.org/10.1126/science.1232437)

Kim, C. Y., Ku, M. J., Qazi, R., Nam, H. J., Park, J. W., Nam, K. S., Oh, S., Kang, I., Jang, J.-H., Kim, W. Y., Kim, J.-H., & Jeong, J.-W. (2021). Soft subdermal implant capable of wireless battery charging and programmable controls for applications in optogenetics. *Nature Communications*, *12*(1), 535. <https://doi.org/10.1038/s41467-020-20803-y>

Kohler, R. E. (1994). *Lords of the fly: Drosophila genetics and the experimental life*. University of Chicago Press.

Kondrakiewicz, K., Kostecki, M., Szadzińska, W., & Knapska, E. (2019). Ecological validity of social interaction tests in rats and mice. *Genes, Brain and Behavior*, *18*(1), e12525.
<https://doi.org/10.1111/gbb.12525>

- Krakauer, J. W., Ghazanfar, A. A., Gomez-Marin, A., MacIver, M. A., & Poeppel, D. (2017). Neuroscience Needs Behavior: Correcting a Reductionist Bias. *Neuron*, *93*(3), 480–490. <https://doi.org/10.1016/j.neuron.2016.12.041>
- Labuguen, R., Matsumoto, J., Negrete, S. B., Nishimaru, H., Nishijo, H., Takada, M., Go, Y., Inoue, K., & Shibata, T. (2021). MacaquePose: A Novel “In the Wild” Macaque Monkey Pose Dataset for Markerless Motion Capture. *Frontiers in Behavioral Neuroscience*, *14*. <https://doi.org/10.3389/fnbeh.2020.581154>
- LaDage, L. D., Roth II, T. C., Fox, R. A., & Pravosudov, V. V. (2009). Effects of captivity and memory-based experiences on the hippocampus in mountain chickadees. *Behavioral Neuroscience*, *123*(2), 284–291. <https://doi.org/10.1037/a0014817>
- Lee, J., Leung, V., Lee, A.-H., Huang, J., Asbeck, P., Mercier, P. P., Shellhammer, S., Larson, L., Laiwalla, F., & Nurmikko, A. (2021). Neural recording and stimulation using wireless networks of microimplants. *Nature Electronics*, *4*(8), 604–614. <https://doi.org/10.1038/s41928-021-00631-8>
- MacIver, M. A. (2008). Neuroethology: From Morphological Computation to Planning. In M. Aydede & P. Robbins (Eds.), *The Cambridge Handbook of Situated Cognition* (pp. 480–504). Cambridge University Press. <https://doi.org/10.1017/CBO9780511816826.026>
- Mao, D., Avila, E., Caziot, B., Laurens, J., Dickman, J. D., & Angelaki, D. E. (2021). Spatial modulation of hippocampal activity in freely moving macaques. *Neuron*, *109*(21), 3521–3534.e6. <https://doi.org/10.1016/j.neuron.2021.09.032>
- Mathuru, A. S., Libersat, F., Vyas, A., & Teseo, S. (2020). Why behavioral neuroscience still needs diversity?: A curious case of a persistent need. *Neuroscience & Biobehavioral Reviews*, *116*, 130–141. <https://doi.org/10.1016/j.neubiorev.2020.06.021>
- Matusz, P. J., Dikker, S., Huth, A. G., & Perrodin, C. (2019). Are We Ready for Real-world Neuroscience? *Journal of Cognitive Neuroscience*, *31*(3), 327–338. https://doi.org/10.1162/jocn_e_01276
- McCall, J. G., Kim, T., Shin, G., Huang, X., Jung, Y. H., Al-Hasani, R., Omenetto, F. G., Bruchas, M. R., & Rogers, J. A. (2013). Fabrication of flexible, multimodal light-emitting devices for wireless optogenetics. *Nature Protocols*, *8*(12), 2413–2428. <https://doi.org/10.1038/nprot.2013.158>
- McClelland, J. L., Botvinick, M. M., Noelle, D. C., Plaut, D. C., Rogers, T. T., Seidenberg, M. S., & Smith, L. B. (2010). Letting structure emerge: connectionist and dynamical systems approaches to cognition. *Trends in cognitive sciences*, *14*(8), 348–356.
- Miller, C. T., Gire, D., Hoke, K., Huk, A. C., Kelley, D., Leopold, D. A., Smear, M. C., Theunissen, F., Yartsev, M., & Niell, C. M. (2022). Natural behavior is the language of the brain. *Current Biology*, *32*(10), R482–R493. <https://doi.org/10.1016/j.cub.2022.03.031>

Mobbs, D., Trimmer, P. C., Blumstein, D. T., & Dayan, P. (2018). Foraging for foundations in decision neuroscience: Insights from ethology. *Nature Reviews Neuroscience*, *19*(7), 419–427. <https://doi.org/10.1038/s41583-018-0010-7>

Montgomery, K. L., Yeh, A. J., Ho, J. S., Tsao, V., Mohan Iyer, S., Grosenick, L., Ferenczi, E. A., Tanabe, Y., Deisseroth, K., Delp, S. L., & Poon, A. S. Y. (2015). Wirelessly powered, fully internal optogenetics for brain, spinal and peripheral circuits in mice. *Nature Methods*, *12*(10), 969–974. <https://doi.org/10.1038/nmeth.3536>

Morris, R. G. M. (1981). Spatial localization does not require the presence of local cues. *Learning and Motivation*, *12*(2), 239–260. [https://doi.org/10.1016/0023-9690\(81\)90020-5](https://doi.org/10.1016/0023-9690(81)90020-5)

Naik, H., Bastien, R., Navab, N., & Couzin, I. D. (2020). Animals in Virtual Environments. *IEEE Transactions on Visualization and Computer Graphics*, *26*(5), 2073–2083. *IEEE Transactions on Visualization and Computer Graphics*. <https://doi.org/10.1109/TVCG.2020.2973063>

Nemati, N. (2024). Rethinking Neuroscientific Methodology: Lived Experience in Behavioral Studies. *Biological Theory*, *19*: 184-197. <https://doi.org/10.1007/s13752-024-00460-w>

Nigrosh, B. J., Slotnick, B. M., & Nevin, J. A. (1975). Olfactory discrimination, reversal learning, and stimulus control in rats. *Journal of comparative and physiological psychology*, *89*(4), 285. <https://doi.org/10.1037/h0076821>

Olton, D. S., & Samuelson, R. J. (1976). Remembrance of places passed: Spatial memory in rats. *Journal of Experimental Psychology: Animal Behavior Processes*, *2*(2), 97–116. <https://doi.org/10.1037/0097-7403.2.2.97>

Orger, M. B., & Polavieja, G. G. de. (2017). Zebrafish Behavior: Opportunities and Challenges. *Annual Review of Neuroscience*, *40*(Volume 40, 2017), 125–147. <https://doi.org/10.1146/annurev-neuro-071714-033857>

Pang, R., Lansdell, B. J., & Fairhall, A. L. (2016). Dimensionality reduction in neuroscience. *Current Biology*, *26*(14), R656–R660. <https://doi.org/10.1016/j.cub.2016.05.029>

Park, S. I., Brenner, D. S., Shin, G., Morgan, C. D., Copits, B. A., Chung, H. U., Pullen, M. Y., Noh, K. N., Davidson, S., Oh, S. J., Yoon, J., Jang, K.-I., Samineni, V. K., Norman, M., Grajales-Reyes, J. G., Vogt, S. K., Sundaram, S. S., Wilson, K. M., Ha, J. S., ... Rogers, J. A. (2015). Soft, stretchable, fully implantable miniaturized optoelectronic systems for wireless optogenetics. *Nature Biotechnology*, *33*(12), 1280–1286. <https://doi.org/10.1038/nbt.3415>

Parker, P. R., Abe, E. T., Beatie, N. T., Leonard, E. S., Martins, D. M., Sharp, S. L., Wyrick, D. G., Mazzucato, L., & Niell, C. M. (2022). Distance estimation from monocular cues in an ethological visuomotor task. *eLife*, *11*, e74708. <https://doi.org/10.7554/eLife.74708>

- Payne, H. L., Lynch, G. F., & Aronov, D. (2021). Neural representations of space in the hippocampus of a food-caching bird. *Science (New York, N.Y.)*, *373*(6552), 343–348. <https://doi.org/10.1126/science.abg2009>
- Peckmezian, T., & Taylor, P. W. (2015). A virtual reality paradigm for the study of visually mediated behaviour and cognition in spiders. *Animal Behaviour*, *107*, 87-95. <https://doi.org/10.1016/j.anbehav.2015.06.018>
- Polley, D. B., Kvašňák, E., & Frostig, R. D. (2004). Naturalistic experience transforms sensory maps in the adult cortex of caged animals. *Nature*, *429*(6987), 67–71. <https://doi.org/10.1038/nature02469>
- Qazi, R., Kim, C. Y., Byun, S.-H., & Jeong, J.-W. (2018). Microscale Inorganic LED Based Wireless Neural Systems for Chronic in vivo Optogenetics. *Frontiers in Neuroscience*, *12*, 764. <https://doi.org/10.3389/fnins.2018.00764>
- Schultheiss, P., Buatois, A., Avarguès-Weber, A., & Giurfa, M. (2017). Using virtual reality to study visual performances of honeybees. *Current Opinion in Insect Science*, *24*, 43–50. <https://doi.org/10.1016/j.cois.2017.08.003>
- Ross, L. N., Jirsa, V., & McIntosh, A. (2024). The Possibility Space Concept in Neuroscience: Possibilities, Constraints, and Explanation. <https://philsci-archive.pitt.edu/23682/>
- Shamay-Tsoory, S. G., & Mendelsohn, A. (2019). Real-Life Neuroscience: An Ecological Approach to Brain and Behavior Research. *Perspectives on Psychological Science*, *14*(5), 841–859. <https://doi.org/10.1177/1745691619856350>
- Soria, E., Schiano, F., & Floreano, D. (2021). Predictive control of aerial swarms in cluttered environments. *Nature Machine Intelligence*, *3*(6), 545-554. <https://doi.org/10.1038/s42256-021-00341-y>
- Soto-Yéber, L., Soto-Ortiz, J., Godoy, P., & Godoy-Herrera, R. (2018). The behavior of adult *Drosophila* in the wild. *PLoS ONE*, *13*(12), e0209917. <https://doi.org/10.1371/journal.pone.0209917>
- St-Cyr, S., Abuaiash, S., Spinieli, R. L., & McGowan, P. O. (2018). Maternal predator odor exposure in mice programs adult offspring social behavior and increases stress-induced behaviors in semi-naturalistic and commonly-used laboratory tasks. *Frontiers in Behavioral Neuroscience*, *12*, 136. <https://doi.org/10.3389/fnbeh.2018.00136>
- Stevenson, T. J., Alward, B. A., Ebling, F. J. P., Fernald, R. D., Kelly, A., & Ophir, A. G. (2018). The Value of Comparative Animal Research: Krogh's Principle Facilitates Scientific Discoveries. *Policy Insights from the Behavioral and Brain Sciences*, *5*(1), 118–125. <https://doi.org/10.1177/2372732217745097>

Testard, C., Tremblay, S., & Platt, M. (2021). From the field to the lab and back: Neuroethology of primate social behavior. *Current Opinion in Neurobiology*, 68, 76–83.
<https://doi.org/10.1016/j.conb.2021.01.005>

Tian, X., Azpurua, J., Hine, C. *et al.* (2013). High-molecular-mass hyaluronan mediates the cancer resistance of the naked mole rat. *Nature* 499, 346–349.
<https://doi.org/10.1038/nature12234>

Uggla, Y. (2010). What is this thing called “natural”? The nature-culture divide in climate change and biodiversity policy. *Journal of Political Ecology*, 17(1), Article 1.
<https://doi.org/10.2458/v17i1.21701>

Ulanovsky, N., & Moss, C. F. (2007). Hippocampal cellular and network activity in freely moving echolocating bats. *Nature Neuroscience*, 10(2), 224–233. <https://doi.org/10.1038/nn1829>

Vallet, W., & van Wassenhove, V. (2023). Can cognitive neuroscience solve the lab-dilemma by going wild? *Neuroscience & Biobehavioral Reviews*, 155, 105463.
<https://doi.org/10.1016/j.neubiorev.2023.105463>

Vanin, S., Bhutani, S., Montelli, S., Menegazzi, P., Green, E. W., Pegoraro, M., Sandrelli, F., Costa, R., & Kyriacou, C. P. (2012). Unexpected features of *Drosophila* circadian behavioural rhythms under natural conditions. *Nature*, 484(7394), 371–375.
<https://doi.org/10.1038/nature10991>

Vogt, C. C., Zippel, M. N., Sprockett, D. D., Miller, C. H., Hardy, S. X., Arthur, M. K., Greenstein, A. M., Colvin, M. S., Michel, L. M., Moeller, A. H., & Sheehan, M. J. (2024). Female behavior drives the formation of distinct social structures in C57BL/6J versus wild-derived outbred mice in field enclosures. *BMC Biology*, 22(1), 35.
<https://doi.org/10.1186/s12915-024-01809-0>

Voloh, B., Maisson, D. J.-N., Cervera, R. L., Conover, I., Zambre, M., Hayden, B., & Zimmermann, J. (2023). Hierarchical action encoding in prefrontal cortex of freely moving macaques. *Cell Reports*, 42(9), 113091. <https://doi.org/10.1016/j.celrep.2023.113091>

Yartsev, M. M., & Ulanovsky, N. (2013). Representation of Three-Dimensional Space in the Hippocampus of Flying Bats. *Science*, 340(6130), 367–372.
<https://doi.org/10.1126/science.1235338>

Yartsev, M. M. (2017). The emperor’s new wardrobe: rebalancing diversity of animal models in neuroscience research. *Science*, 358(6362), 466–469. <https://doi.org/10.1126/science.aan8865>

Ye, D., Walsh, J. T., Junker, I. P., & Ding, Y. (2024). Changes in the cellular makeup of motor patterning circuits drive courtship song evolution in *Drosophila*. *Current Biology*, 34(11), 2319–2329.

Zong, W., Obenaus, H. A., Skytøen, E. R., Eneqvist, H., de Jong, N. L., Vale, R., ... & Moser, E. I. (2022). Large-scale two-photon calcium imaging in freely moving mice. *Cell*, 185(7), 1240-1256. [10.1016/j.cell.2022.02.017](https://doi.org/10.1016/j.cell.2022.02.017)