1 Biologging for the future: how biologgers can help solve fundamental questions, from 2 individuals to ecosystems

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32 1. Abstract

33 Archival instruments attached to animals (biologgers) have enabled exciting discoveries and have promoted effective conservation and management for decades. Recent research 34 35 indicates that the field of biologging is poised to shift from pattern description to process 36 explanation. Here we describe how biologgers have been - and can be - used to test 37 hypotheses and challenge theory in behavior and ecology through three case studies and many 38 short examples. These examples, spanning predator-prey interactions, state-dependent risk-39 taking, resource tracking, and collective movement decisions, show how biologging can resolve 40 long-standing mysteries if research is designed with a solid conceptual foundation. The next 41 phase of biologging science will require scaling studies from individuals to populations and 42 possibly to ecosystems. It will also benefit from building equitable international and interdisciplinary partnerships, bridging the terrestrial-marine divide, and addressing ethical 43 44 conundrums including animal handling and open science practices. Doing so will help cement 45 biologging as an indispensable tool for producing generalizable knowledge about how 46 organisms and ecosystems function.

47 **2. Background**

A holistic understanding of complex ecological processes requires creative solutions for measuring broad-scale, high-resolution data across sites, species, and systems. Archival instruments attached to animals (hereafter, "**biologgers**") have become routine tools for recording physiological, behavioral, and demographic characteristics of individuals, and their interactions with environmental and ecological features ¹ (Figure 1).

53 The inception of biologging enabled novel discoveries such as understanding how 54 animals migrate across the globe relative to their prey and predators, including questions at 55 temporal and spatial scales that far exceed those from traditional approaches. Like other 56 technologically-driven disciplines including astronomy and genetics, biologging is poised to shift 57 from pattern discovery to designing studies with strong conceptual underpinnings². Biologging 58 studies have transitioned from providing unique observations to adopting a rigorous scientific 59 approach in recent years. Researchers now formulate defined questions, ensure adequate 60 sample sizes, and conduct in-depth analyses, shifting from simple movement descriptions to hypothesis-driven investigations. For example, biologgers have provided data on foraging 61 62 success, which can be used to test hypotheses about functional response types, risk-reward 63 trade-offs, and resource tracking (Figure 1).

Biologging studies have also applied "big-data" approaches to tackle behavioral and ecological mechanisms underlying movements and the proximate internal and external factors that constrain them (Figure 1)³. In turn, the discoveries and natural history descriptions enabled by new technology are primed for the testing and refinement of long-standing theories ⁴, with implications for broad generalizations ⁵ and evidence-based conservation solutions ⁶. The next step is to answer questions that span multiple taxa, such as whether generalizable "rules" underpin complex movements and species interactions ⁷.

71 For many of these outstanding questions, ecological theory exists to generate 72 hypotheses about how animals behave and how behavior scales to ecosystem processes such 73 as distribution and abundance across trophic levels ⁸. Previous empirical tests of many theories 74 have been limited to controlled laboratory experiments. In contrast, biologging in natural settings 75 enables precise measurements in ecologically relevant contexts where competing selective 76 pressures such as predation, starvation, and infection exist. Thus, biologging is perfectly 77 positioned to test many hypotheses from long-standing ecological theories that are underpinned 78 by animal behavior.

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0 3. Objectives

81 The rapid development of biologging has motivated numerous proposals for future 82 directions. Previous researchers have urged the biologging community to integrate with other disciplines ⁹, use mathematics and optimality ^{9,10}, re-unite big data approaches with field-based 83 ecological processes ¹¹, target key knowledge gaps ¹², and reconnect tools and questions ^{13,14}. 84 85 We believe a key next step is to test long-standing ecological theories through biologging 86 studies, and to identify areas of intensive data collection that are well-positioned to bridge this 87 gap. This review aims to illustrate how biologging is uniquely suited to test hypotheses and 88 refine theory because it can integrate complexity into mechanistic tests. First, we describe three 89 case studies (representing state-dependent behavior, learning, and memory) and many 90 example publications in which observational data led to discoveries that were then leveraged to 91 test hypotheses and develop robust conceptual frameworks. Second, we identify outstanding 92 questions in ecological theory that would be best addressed with large-scale integration of 93 biologging studies across systems. Finally, we discuss practical steps toward theory-grounded 94 biologging through data synthesis across the terrestrial-marine divide, concept-driven motivation 95 for technology development, open data practices, and intentional interdisciplinary partnerships.

96 4. Examples of development from discovery to theory

97 We provide three case-studies below that range from single-species behavioral research 98 to multispecies interactions and consequences for population dynamics (Figure 2). These case 99 studies show how discoveries and theory testing can comprise a positive feedback loop, where 100 observations lead to insight that inform hypotheses, and testing those hypotheses can lead to 101 theory development, which can then be refined through continued observations and discoveries. 102 In addition to summarizing key findings, we propose directions for future research questions that 103 could be addressed with targeted biologging studies or comparative studies using existing data 104 grounded in conceptually solid foundations.

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106 **4.1. State-dependent risk-taking throughout migration in elephant seals**

107 Body condition has been hypothesized to mediate how animals navigate risk-reward 108 trade-offs; however, the difficulty of measuring body condition, predation risk, and food rewards 109 in the wild has limited our understanding of how intrinsic states affect risk-taking behavior ¹⁵. 110 Biologgers have begun to address this gap by collecting data on animal movement behavior: pioneering time-depth recorders facilitated the discovery that elephant seals cease swimming 111 112 and passively drift during some dives ¹⁶. These so-called "drift dives" often occur after foraging 113 bouts, and seals with less body fat at the start of the foraging trip sink faster ¹⁷. Later, drift rates were validated against longitudinal energy gain rates ¹⁸ as a valuable metric for estimating body 114 115 composition ¹⁹ at fine spatio-temporal scales throughout migrations. At the same time, inertial measurements of drifting seals provided further support for hypotheses about the functions of 116 117 drift dives for resting and food processing ²⁰. Finally, biologgers were used to confirm that seals 118 sleep during drift dives ²¹ and derive body composition estimates and ethograms at a daily scale 119 throughout months-long migrations to test predictions from state-dependent risk aversion theory about when animals should rest and forage ²². This three-decade research arc examining 120 121 whether behavior is state-dependent throughout oceanic migrations suggests that seals in 122 superior body condition sacrifice more profitable nocturnal foraging hours to sleep in the safety 123 of darkness (Figure 2) ²².

124 When grounded in a strong conceptual framework, the development of new technologies 125 and analytical approaches holds great promise for studying state-dependent behavior in free-126 ranging animals. New on-board processing algorithms have been developed to estimate and 127 transmit real-time body composition data to test state-dependent behavior in species that do not perform drift dives ²³. Likewise, other components of an animal's internal state (e.g., hunger, 128 129 heat, stress, exhaustion ²⁴) may influence behavior and could be measured with new sensors. Could new physiological biologgers that measure brain activity ²⁵ and heart rate ²⁶ be used to 130 131 test theories of physiological recharge, rebound, and replenishment (e.g., sleep quotas) in other species ²⁷? Can we simultaneously instrument predators and prey to parse spatial and temporal 132 133 dimensions of risk and reward in "dynamic landscapes of fear" or "dynamic energy landscapes" theoretical frameworks 28,29? 134

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136 **4.2. Social and experiential learning in whooping cranes**

137 How animals learn long-distance migrations is an outstanding question that has 138 implications for wildlife conservation and management including the designation of protected 139 areas and strategic planning of species re-introductions. The role of social versus experiential 140 learning in migratory behavior has been tested in studies of re-introduced whooping cranes 141 (Grus americana) (Figure 2). Using satellite transmitters and VHF tags, movements and survival 142 of re-introduced whooping cranes demonstrated successful migration and dispersal patterns ³⁰. A decade later, a reanalysis of these data tested the hypothesis that re-introduced cranes 143 144 migrate more efficiently when flying with experienced birds ³¹. More recently, researchers used 145 a long-term satellite tracking dataset to discover an ontogenetic switch from social to 146 experiential learning as birds age, allowing them to track resources throughout their migration at

147 all ages ³². This work transitioned from initial descriptive discoveries (examining whether re-148 introduction was a successful management strategy) to theory-testing (disentangling the roles of 149 social versus experiential learning) and laid the groundwork for testing many long-standing 150 ecological questions. For example, if social learning is a primary mechanism by which 151 movement strategies evolve, will interrupted information transfer have population-level 152 consequences ³³? How do implicit or explicit social cues influence the timing of animal 153 migrations ³⁴? How do migratory animals gain information from other species ³⁵? To what extent 154 does social learning shape the outcomes of predator-prey interactions and demographic 155 processes ³⁶?

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4.3. Migration based on past and current resource availability across species

158 Strong foundational biologging discoveries about memory and **behavioral** 159 compensation have set the stage for rich theory and emerging tools to address long-standing 160 questions about migration ³⁷. One area that has seen significant theoretical progress across 161 species and systems is the influence of variation in resource availability on migratory behavior 162 (Figure 2). In the late 1970's, researchers observed simultaneous changes in plant phenology and migratory movements in waterfowl ³⁸. Later, biologgers were used to examine the link 163 between **vegetation green-up** and ungulate migration decisions ³⁹. Follow-up studies tested 164 165 hypotheses that mule deer migrate in concert with green-up, maximizing energy intake rather than speed ⁴⁰, and that the rate and order of green-up influences the ability of animals to green-166 167 wave surf ⁴¹. Simulation of zebras and mule deer and empirical validation with GPS tracks 168 showed that previous experience (memory) with green-up patterns also plays an essential role in migration ^{42,43}. More recently, data analysis from biologgers suggested revising the theory 169 170 about the relative roles of memory ⁴⁴ and proximate cues ³⁴ in the movement of terrestrial and 171 marine animals.

172 Biologging can help us address the interplay of animal migrations in response to 173 environmental drivers. Questions that can be addressed include: What movement patterns 174 emerge in response to stochastic resource patterns (i.e., neither pulses nor waves) ^{45,46}? How 175 can in situ resource availability be measured (instead of inferred from physical environmental 176 metrics) at a sufficiently fine resolution to inform theory about animal movement, especially in 3-177 dimensional ocean environments ^{47,48}? Can we unite theory examining predation risk and 178 resource tracking to integrate top-down and bottom-up processes in estimating risk-reward 179 trade-offs and optimal decision-making during migration ⁴⁹? In turn, can we develop and revise 180 theories about how animals contribute to resource pulses and environmental predictability through environmental feedback ⁵⁰? For example, one could determine if migratory marine 181 182 megafauna that serve as ecosystem engineers promote a green wave of productivity that sustains themselves and their ecosystems, like bison in terrestrial systems ⁵¹. Biologging data 183 184 are uniquely suited to understand how animals respond to and influence resource pulses. 185 particularly in a changing world.

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187 **4.4. Examples in other sub-disciplines.**

188 There are many additional examples of biologging studies in both the marine and 189 terrestrial realms that successfully test theory and/or have strong conceptual foundations. In 190 Figure 3, we provide short summaries of studies spanning a broad set of sub-disciplines 191 including Morphology, Physiology, Behavior, Demography, Sociality, Ecology and Cross-taxa 192 studies and crossing levels of biological organization from individuals to populations and to 193 ecosystems (Figure 1). These examples demonstrate how biologging scientists can maximize 194 the impact of their work by ensuring that theoretical concepts ground our research (e.g., Figure 195 3), regardless of whether the work is more motivated by pattern (inductive) or by theory 196 (deductive) ⁵².

197 **5. Outstanding Questions**

198 Emerging questions that can be answered with biologgers, ranging across temporal and 199 spatial scales, are provided in Figure 4. Many of the highlighted questions are moderately well 200 understood from a theoretical perspective but are primed for empirical testing through finer-201 resolution or larger-scale biologging data. For other questions cases, extensive empirical data 202 could challenge or refine theory by identifying deviations from optimality or highlighting 203 discrepancies between theory and empirical data. Many cases represent significant 204 opportunities for developing both theory and empirical evidence, especially across disciplinary boundaries. All can be used as an agenda for further research and a benchmark for future 205 206 evaluations of progress in biologging science.

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208 6. Roadmap for the Future

209 6.1. Big data syntheses

210 Synthesis of multiple biologging datasets with extensive breadth and depth can produce 211 generalizable knowledge about patterns and processes in the natural world across vast spatial 212 and temporal scales (Figure 4). Recent efforts include the Tagging Of Pacific Predators ⁵³, 213 Retroactive Analysis of Antarctic Tracking Data, ⁵⁴, Arctic Animal Movement Archive ⁵⁵, COVID-214 19 Bio-Logging Initiative ⁵⁶, and Wyoming Migration Initiative ⁵⁷. Many of these efforts were 215 necessarily discovery-based but laid a foundation for future concept-driven research, such as 216 answering questions like: Can we identify multispecies hotspots in three-dimensional space and 217 time? What role do these hotspots play in ecosystem function (Figure 4)? Re-analyzing and 218 synthesizing datasets can maximize insights and minimize research impacts including animal 219 handling. A promising area for future conceptual research is bridging terrestrial and marine 220 systems. For example, terrestrial studies often include in situ focal follows to interpret biologging 221 data ⁵⁸. In contrast, marine studies commonly bridge the spatio-temporal scales between 222 coarser environmental and finer movement parameters by integrating environmental sensors 223 into biologgers ^{14,59}. The few studies that span the marine-terrestrial interface have produced 224 valuable comparative insights ⁶⁰ and deepened our understanding of relationships between 225 land and sea ⁶¹. 226

227 6.2. Interdisciplinary integration

228 Integrating biologging with complementary disciplines can also generate new theory and hypotheses at the interfaces between the two ⁶². For example, including genomic and 229 230 demographic covariates could help disentangle the contributions of genetic mechanisms. ontogeny, and senescence to animal movement patterns ^{63,64}. Similarly, combining mark-231 232 recapture studies with biologging studies could help elucidate the fitness consequences of 233 movement ^{65,66}. Merging trait databases (e.g., inter-specific variation in body size, brain size, sensory organs, reproductive characteristics, and diet ⁶⁷) with tracking data ⁶⁸ or vulnerability 234 235 risk assessments ⁶⁹ could facilitate large-scale ecological-evolutionary insights (Figure 4). 236 Alternatively, using biologgers to measure traits and putting those traits into an eco-evo 237 framework can provide insightful phylogenetic contexts for behaviors and species interactions 70–72 238

239 Biologging data can also facilitate field-based experimental biology by allowing for the 240 quantification of movement responses to experimental treatments such as simulated human 241 activity ⁷³. Long-duration interdisciplinary studies that span a range of environmental conditions 242 and extreme events are necessary to predict future changes and answer a key outstanding 243 question: How sensitive are behavior-performance relationships to environmental context 244 (Figure 4) 62,74? Understanding and predicting the spatio-temporal dynamics of populations is a central goal in ecology and conservation, and requires understanding variation in, and genetic 245 246 and phenotypic drivers of, demographic rates 75-77.

247 Another emerging interdisciplinary direction is field sensory biology, which focuses on 248 the detection of prey by predators ⁴⁷ and less commonly the detection of predators by prey ⁷⁸. 249 Future biologging studies of feeding behavior and prey detection can help address questions 250 including, including how animals perceive and respond to multimodal sensory information (vision, olfaction, tactile, acoustic) in natural systems ⁷⁹? How can biologgers measure how 251 252 animals produce and perceive auditory cues to respond to conspecifics, heterospecifics, and 253 their environment ⁸⁰? How is sensory perception built into an animal's cognitive map of its environment ^{81,82}? 254

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256 **6.3. Novel tools, approaches, and theories**

257 Using existing biologging data to address new questions outside those of the initial study 258 is sometimes problematic. New tools, approaches, or theories may be needed to answer a 259 given conceptual question in these cases. Here, concept-driven motivation for improved tagging 260 technologies can drive the collaborative development of smaller biologgers and more reliable 261 sensors that overcome current limitations in measuring the covariates that matter to animals 262 (e.g., *in situ* resources, predator abundance, real-time measurements of endocrine markers, 263 infection status, or heat stress). For example, the recent use of real-time monitoring 83, backmounted accelerometers⁸⁴, and underwater cameras triggered by accelerometry during feeding 264 265 attempts ⁸⁵ have facilitated conceptual advances in movement decisions, interspecific 266 communication, and resource acquisition. In some cases, this may require the integration of 267 several theories; for example, can we unite the Landscape of Fear with the Landscape of 268 Disgust theories to empirically test how animals navigate the risks of mobile parasites and 269 predators ^{86–88}? 270

271 6.4. Inclusive, equitable biologging

272 More collaborative and diverse teams facilitate innovation and transformative scientific 273 insights ^{89,90}. More equitable approaches to biologging science, including data collection, processing, archiving, and reporting, will provide the greatest impact and the broadest 274 275 participation by diverse individuals, institutions, and nations. For example, the research community can seek to understand and preserve the socio-ecological systems within which the 276 277 animals operate through global and interdisciplinary collaborators ¹⁴, including Indigenous 278 communities who can be compensated for their time and meaningfully involved in 279 instrumentation decisions. Likewise, a global tag registry could help the biologging community 280 know who to contact about sampling efforts and datasets ⁹¹. Standardized and reproducible 281 data can fast-track comparative biological analyses and interdisciplinary research using 282 biologgers ⁹² by ensuring that data are both available and usable. This is especially needed for 283 non-spatial data such as physiologging, accelerometry, and video data that do not fit the standards developed for spatial data ⁹³. Open-access publication of those datasets could credit 284 285 those who collect and share data ⁶⁸.

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287 **7. Concluding Remarks**

288 The key to moving from anecdote to generalizable theory is to examine interdisciplinary 289 patterns and processes across species and habitats. For example, habitat selection can be 290 influenced by the population context (e.g., density and social dynamics), by the community 291 context (e.g., predators and competitor density), and by the environmental context (e.g., 292 drought, fire) (Figure 1). Biologging enables efficient, fine-scale data collection that can provide 293 the breadth and depth needed to develop, test, and refine our understanding of ecological 294 processes. Research with a conceptual focus is often more impactful and can provide insights across various disciplines and systems (Figure 3) ⁶⁶. Moreover, predictive models used in 295 296 management and conservation will be most accurate when they are informed by process-based 297 understandings of animal movements and their roles in ecological processes 94.

298 Similarly, theory can be applied to humans in their roles as shields (creating a spatial 299 refuge 'shielding' prey from predators) ⁹⁵, predators, resource drivers, disease sources, and 300 tested with biologgers to inform conservation and management actions ⁹⁶. Of course, theory 301 testing and refinement are not always possible. Research is still in the description and discovery phase for species that have never been instrumented. In these cases, a closer connection 302 303 between conceptual questions and biologging technology can expedite the development of new 304 theories and contribute to the iterative process of testing and refinement ¹³. We believe biologging can have "unimaginably important applications" ⁶² and address new ideas by 305 306 focusing on theory-driven hypothesis testing. Through these avenues, biologging can illuminate 307 how nature works and thus provide a roadmap for better protection of species, ecosystems, and 308 the services they provide 68. 309

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circles) that can be used to address key research topics (white squares) in behavior and ecology across scales from individuals to populations to communities and ecosytems.





Figure 2. Illustrated examples of the development from discovery to hypothesis testing to theory
refinement for each of three case studies: (a) state-dependent risk taking throughout migration
in elephant seals, (b) social and experiential learning in whooping cranes, and (c) migration
based on past and current resource availability across species.



Figure 3. Examples of biologging studies with strong conceptual/theoretical foundations.
 Examples are organized by discipline. See references for full citations ^{39,55,58,65,70,72,95,98–111}.



575 576

Temporal Scales

Figure 4. Some promising areas of research that could be addressed using biologgers, across

temporal and spatial scales. Colored by discipline (see Fig 4 for key). 577