<u>Minimum Information for Reusable</u> <u>Arthropod Abundance Data (MIReAAD)</u>

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4 Myriad: a countless or extremely great number

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72 Abstract

73 Arthropods play a dominant role in natural and human-modified terrestrial ecosystem dynamics.

74 Spatially-explicit population time-series are crucial for statistical or mathematical models of

these dynamics and assessment of their veterinary, medical, agricultural, and ecological impacts.

76 Arthropod data have been collected world-wide for over a century, but remain scattered and

77 largely inaccessible. With the ever-present and growing threat of arthropod vectors of infectious

diseases and pest species, there are enormous amounts of historical and ongoing surveillance. 78 79 These data are currently reported in a wide variety of formats, typically lacking sufficient 80 metadata to make reuse and re-analysis possible. We present the first minimum information 81 standard for arthropod abundance. Developed with broad stakeholder collaboration, it balances 82 sufficiency for reuse with the practicality of preparing the data for submission. It is designed to optimize data (re-)usability from the "FAIR," (Findable, Accessible, Interoperable, and 83 84 Reusable) principles of public data archiving (PDA). This standard will facilitate data unification 85 across research initiatives and communities dedicated to surveillance for detection and control of 86 vector-borne diseases and pests.

87 Introduction

88 Arthropods play a dominant role in the dynamics of practically all natural and human-modified terrestrial ecosystems $^{1-3}$, and have significant economic and health effects. For example, certain 89 90 insects provide significant economic benefits (e.g. pollination) exceeding \$57 billion a year to the United States alone⁴. Meanwhile, invasive insects cost an estimated \$70 billion dollars per 91 year globally⁵ and insect pests may reduce agricultural harvests by up to 16%, with an equal 92 amount of further losses of harvested goods⁶. Particularly noteworthy is a subset of arthropods 93 94 that are disease vectors, transmitting pathogens to and between animals as well as plants. Vectorborne diseases cause billions of dollars in crop and livestock losses, every year⁷⁻⁹. In humans, 95 96 vector borne diseases account for more than 17% of all infectious diseases (e.g. malaria, Chagas, 97 dengue, and leishmaniasis, Zika, West Nile, Lyme disease, and sleeping sickness), with hundreds of thousands of deaths, hundreds of millions of cases, and billions of people at risk, annually^{10,11}. 98

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100 The current economic and health burden of arthropod pests, exacerbated by invasive species, and uncertain effects of climate change^{12,13}, has driven significant research programs and data 101 collection efforts. These include crop pest, mosquito, and tick survey and reporting initiatives^{14–} 102 ¹⁸, citizen science projects^{19–21}, and digitization of museum specimen data^{22,23}, all yielding a rich 103 104 and growing trove of field-based data spanning multiple spatial and temporal scales. Monitoring 105 arthropod abundance (e.g. Figure 1) in different disciplines (e.g., biodiversity research, pest-106 control assessment, vector-borne disease monitoring, or pollination research) uses similar 107 techniques, with similar objectives: to quantify abundance, phenology and geographical ranges 108 of target arthropod species. Despite a growing number of data collections, they are often not 109 reusable, or comparable to similar data, due to a lack of standardization and metadata. In 110 contrast, the advent of the deposition of data from high-throughput technologies (e.g. NCBI and 111 GenBank), data and code sharing, and other practices to improve transparency and reusability of research results are increasing rapidly across the sciences $^{24-29}$. Furthering these advances through 112 113 standardization and public archiving of arthropod abundance data can bring significant benefits, 114 including (1) supporting empirical parameterization and validation of mathematical models (e.g. 115 of pest or disease emergence and spread), (2) validation of model predictions, (3) reduction in 116 the duplication of expensive empirical research, and (4) revealing new patterns and questions through meta-analyses^{30–33}. This will also lead to substantial public benefit through improved 117 118 human, animal, plant, and ecosystem health, and reduced economic costs.

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A key impediment to the re-use of these data is the lack of adequate metadata or data descriptors
 (*i.e.* data about the data)³⁴⁻³⁷. In general, for data to be most valuable to the scientific community, 4

122 they should meet the FAIR Principles – they should be Findable, Accessible, Interoperable and 123 Reusable - and delineate the key components of good data management and stewardship practices^{38,39}. Data are Findable and Accessible when they are archived and freely downloadable 124 125 from an online public data repository that is indexed and easily searchable. Interoperability and 126 reusability describe the ease with which humans or computer programs can understand the data 127 (e.g. via metadata) and explore/re-use them across a variety of non-proprietary platforms. Even 128 when data are available, metadata for arthropod abundance data are often absent or not readily 129 interpretable, limiting their reusability at a fundamental level.

130 **Results**

131 A minimum information standard for arthropod abundance data

132 Here, we present a Minimum Information for Reusable Arthropod Abundance Data (MIReAAD) 133 standard for reporting primarily longitudinal (repeated, temporally explicit) field-based 134 collections of arthropods. In the same manner as has been developed in other biological disciplines⁴⁰⁻⁴⁵, this standard is "minimum" because it defines the necessary minimal information 135 136 required to understand and reuse a dataset without consulting any further text, materials, or methods⁴⁶. MIReAAD is designed to facilitate data archiving efforts of publishers and field 137 138 researchers. It is not a data model and therefore does not define controlled vocabularies, or 139 specific field titles, but should be easy to understand, and interpret by the wider scientific community⁴⁶. 140

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142 The minimal standards are separated into two components, metadata and data. For each 143 component, we provide a description of the information that should be included,

144 recommendations for how to make that information as useful as possible, and examples. The 145 metadata component (Table 1) includes information for the origin of the data set (e.g. study 146 information and licensing for usage). The second component (Table 2) lists and describes 147 specific data fields that should be included in data collection sheets. We also provide 148 recommendations and examples to demonstrate how these recommendations can be 149 implemented. MIReAAD was designed to match the data that are generally collected by 150 academic researchers and surveillance initiatives, and can serve as a checklist for important 151 information that needs to be recorded but is often unintentionally omitted (e.g. Figure 2A). By 152 adhering to MIReAAD standards, omissions and ambiguity can be avoided even if the data are 153 shared in different formats (Figure 2B and C). Finally, we identify common problems likely to 154 be encountered across all the MIReAAD metadata and data fields, and data quality standards that 155 can be employed to avoid confusion (Box 1).

Box 1. Data quality standards

No abbreviations. Abbreviations (including in columns names) are ambiguous, with the exception of measurement units (*e.g.* centigrade and meters).

No external legend/key files. While repetitive, all data should be explicitly given within the data table. Separate files mapping ID numbers to GPS locations, full species names, etc., should be avoided. In addition, rich metadata is essential for good data discovery and reuse.

Unambiguous dates. Because of country-level differences in date formats, data should be reported with 4 digit years, and months provided alphabetically and not numerically (*e.g.* 4-Jun-2017 or Nov 12, 2015).

Machine-readable file formats. Data should be provided in non-proprietary machine readable formats such as comma-separated text files. PDFs and multiple spreadsheets in the same document should be avoided.

No font styling or subsection headings. Formatting (color, bold, italics, subscripts, sheet tab names, *etc.*) should not be required for understanding the data. Subsection headings should not be required to understand data; every line of data should be interpretable in isolation from any other line of data.

Highest precision possible. Data should be provided at the highest temporal, spatial, numerical, and taxonomic resolution available. If location (*e.g.*, geographical coordinate) data need to be presented at a lower resolution than available for privacy reasons, this should be made clear in the submission in Study Information (Resource Metadata; Table

1).

Language. Once data are ready to be deposited/submitted, all fields and data are preferably written in English. This will allow researchers and data curators worldwide to understand and reuse the data. Use of other languages is better than not publishing data. Please avoid introducing data reuse barriers through incomplete translation. For example, non-English field names in an English-language submission.

158 Examples

Below we provide three examples to illustrate MIReAAD compliant data (linked to Supplemental Data Files 1-4, respectively). Researchers can use these data sheets as a basis for formatting their own data. In these examples, note that all data meet the data quality standards of Box 1; are adequately described, have columns labeled, *etc.* to eliminate ambiguity (even if the data appear repetitive; for example, the sex and life stage are repeated in every row). Examples 1 and 2 should be sufficient for most data generators. Example 3 (Data Files 3-4) demonstrates a more complex data collection scenario.

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167 1. *Long-format trapping data*. Each row captures count data for a single species' occurrence in a
168 given sampling event. This illustrates an example of the most common mosquito collection

169 protocol. [<u>Sup Datasheet 1</u>]. Also see Figure 2B. 170

171 2. Wide format trapping data. Each row captures count data from a given sampling event. Each 172 identified taxonomic group is identified in a separate column. An 'additional sample 173 information' field, 'sub-location,' has been added to describe the various locations around the 174 village where collections were made. [Sup Datasheet 2]. This illustrates an example of adult 175 mosquito populations that have been tracked over time and in specific locations. Also see Figure 176 2C.

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178 3. Complex trapping data scenario. Tick surveillance performed using tick drags and flags and 179 collections of ectoparasites on trapped mice. The tick drags/flags report three life stages 180 independently (adult, larvae, and nymph) [Sup Datasheet 3]. Larvae are only identified to the 181 genus, while adults and nymphs are identified to the species. Observations of different life 182 stages and sexes are preferably documented in separate records. A Sample Name is used to help 183 link these records (but would not be necessary.) The mouse survey uses an additional sample 184 information field to record the sex of the trapped mouse from which the parasites were collected 185 [Sup Datasheet 4].

186 Discussion

187 MIReAAD as the path to FAIR data principles

We designed MIReAAD to achieve a balance between standards that are too onerous for data generators and standards that are sufficient to ensure at least minimal reusability^{31,40}. Like all

190 minimum standards, MIReAAD only aims at ensuring data 'Reusability'. However, ultimately 191 this will promote the implementation of data models — the explicit definition of data field 192 names, data formats (e.g., for dates and GPS locations), and controlled vocabularies (e.g., the Darwin Core⁴⁷). Data models enable 'Interoperability', and in turn facilitate structured databases, 193 public repositories, and development of data analysis tools^{46,48}. Deposition in open databases 194 make data 'Findable' and 'Accessible'⁴⁹⁻⁵¹. MIReAAD compliant data contain sufficient 195 196 information for established aggregators/databases such as VectorBase and SCAN (Symbiota Collections of Arthropods Network⁵²) to process and store the data in a standardized data model 197 198 [e.g., Darwin Core, a widely used universal data standard that supports opportunistic observation 199 and collection data (occurrence core) as well as presence/absence and abundance data collected using strict and documented methodology (event core)⁴⁷], and ultimately facilitate data transfer 200 201 to even more comprehensive biodiversity databases [e.g. GBIF, which contains over one billion 202 species occurrence records, from thousands of environmental, ecological, and natural resource 203 investigations, including research on Arthropoda in numerous ecological and monitoring projects, allowing for study of changes and trends in populations.⁵¹]. Indeed, in Supplemental 204 205 File 5, we provide an example of the mapping of data fields from this minimum information 206 standard, to DarwinCore and GBIF. In this way, MIReAAD opens the door to FAIR data and 207 more sophisticated methods to integrate data across many scales.

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209 Benefits to field researchers

It is essential that the benefits of a minimal data standard extend not just to data re-users, but also
to the researchers who collect and generate data in the first place. MIReAAD provides a
framework for data preparation that can help scientists achieve recognized professional merit for 10

sharing data such as increased citation rates, academic recognition, opportunities for co-213 214 authorship, and new collaborations [sensu Roche et al. 2014³¹]. Large, deposited data sets can now themselves be standalone, citable "data papers" (e.g. 53-55) or even depositions without any 215 216 traditional manuscript (but as an authored 'digital product,' with persistent identifiers, such as a 217 DOI number), if desired. Data sets are increasingly recognized as valuable research outputs that 218 count towards academic recognition and professional advancement (e.g. grants, interviews, and 219 tenure). For example, several funders (e.g. United States National Science Foundation and Swiss 220 National Science Foundation) have adopted or are in the process of adopting the Declaration on Research Assessments (DORA)⁵⁶, offering further opportunities for data generators to gain 221 recognition and publication credit for their work⁵⁷. Also, an increasing number of funders are 222 223 mandating public data access, and detailed data management plans are often required even at the 224 grant proposal stage. Therefore, reporting data according to MIReAAD will provide a 225 foundational pipeline for stipulating archival formats.

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Furthermore, many data generators are also data users. Developing analyses that rely on standardized fields can facilitate the development of generalized analytical tools that can be easily extended to datasets beyond those that were collected by a single individual or lab. In this way, they can enable extensions of work that would otherwise not happen, such as comparisons of population dynamics in different locations or assessments of interspecies interactions. Adopting MIReAAD therefore can both help data generators reap the benefits of sharing data they have collected and enable them to more readily leverage data collected by others.

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235 Further MIReAAD applications and extensions

236 The creation of minimum information standards for these types of databases facilitates analyses 237 of data at the scales that cannot be attained by a single individual or lab group. Linking records 238 to additional information also extends the utility of these data to address population level 239 questions. For example, a well-populated database presents opportunities to investigate 240 interactions between populations of different species of arthropod that overlap in geography, but 241 may be of interest individually to different realms of research. As a case in point, in the 242 northeastern USA, Agrilus plannipennis, the Emerald Ash Borer (EAB), is a highly destructive 243 invasive insect, monitored closely by both state and federal agencies for management⁵⁸. 244 Interestingly, EAB are creating lots of new habitat for carpenter bees, a species interaction that 245 can be tracked and anticipated using large scale arthropod data.

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247 Another example of the utility of linked data is for disease vectors. Data on insecticide resistance 248 linked with time and place would be valuable for coordinating control strategies within and 249 between nations and communities. Presence/absence data on infection levels would be helpful 250 for tracking and investigating disease outbreaks, and dynamics. Standardization of these data 251 would be particularly useful for pathogens that infect multiple vectors and hosts and would 252 facilitate a "One Health" approach. Other important vector phenotypes that contribute to control 253 and transmission such as pathogen susceptibility, biting preferences, and breeding behaviours 254 could be measured over time and space.

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We note that MIRreAAD is applicable not only to abundance measurements, but could be easily extended to any other kind of routinely sampled time-series field data. For example, in addition to aphid abundance, plant pathogen (such as mosaic virus) infection and insecticide resistance
statuses of the aphids could be reported in MIRreAAD format.

260 Conclusion

261 We present MIReAAD, a minimum information standard for representing arthropod 262 abundance data. MIReAAD will facilitate collation and analyses of data at scales that cannot be 263 attained by a single individual or lab, to address key questions across temporal and spatial scales, 264 such as within and across-year phenology of abundance of target arthropod taxa over large 265 geographical areas. This is particularly important given the pressing need to understand and 266 predict the population dynamics of harmful (e.g., disease vectors and pests) as well as beneficial 267 (e.g., pollinators, bio-control agents) arthropods in natural and human modified landscapes. This 268 is the first step for achieving the broad benefits of FAIR data for arthropod abundance. We call 269 on data generators, authors, reviewers, editors, journals, research infrastructures (e.g. data 270 repositories) and funders to embrace MIReAAD as a standard to facilitate FAIR data use and 271 compliance for arthropod abundance data.

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Table 1. The MIReAAD Study Information (Resource metadata) fields. The information in this
table should be included with every data submission, for example by including data in the file
header as demonstrated in Data Files 1-4.

Field	Details	Recommendations	Examples
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Contact details	A name, person, authority, etc. that may be contacted with enquiries about the data.	Include investigator ORCID(s), email address, website (if institutional) if possible.	Kurt Vandegrift orcid.org/0000-0002-5690-3300 kurtvandegrift@gmail.com State University Agricultural Extension John Smith (jsmith@StateU.edu) www.StateU.edu/AgriculturalExtension/
General description of the experiment/ collection set	A short description of the study objectives, sampling design, and hypotheses. Used to aid in browsing multiple studies. A short title and long form name might be helpful.	Useful things to indicate are: Random sampling or continuous monitoring in fixed locations General time frames and location. General description of where data is from.	"Long term, fixed trapped, municipal surveillance of west Nile vector population in Colorado from 2000-2010" "Pennsylvania <i>Ixodes scapularis</i> weekly abundance" Continuous (weekly) monitoring of tick numbers attached to White-footed mice in fixed locations in Pennsylvania, USA (12 sites). 2003-present." "Long term aphid emergence monitoring using continuous suction traps"
Citations	Reference to related publications, digital if possible (e.g. DOI(s) or PMID(s)).		"A web-based relational database for monitoring and analyzing mosquito population dynamics Sucaet Y, Van Hemert J, Tucker B, Bartholomay L." "PMID: 18714883" Horiuchi, Kaho, Kosei Hashimoto, and Fumio Hayashi. "Cantharidin world in air: Spatiotemporal distributions of flying canthariphilous insects in the forest interior." Entomological Science (2018).
Species Identification Method	A description of method of species identification. Particularly important for cryptic species complexes.		"Morphological" "Genotyped, using method of Smith et al 2014, PMID: 18714883"
Not present vs zero information	Indication of what gaps, zeros, NA, etc mean.	It is imperative, especially for population surveys, to understand the difference between a species was not found when the collection method would be expected to find the given species (confirmed absence) or a species was not	"Zero indicates was looked for and not found. NA represents a trap failure etc"

		looked for (<i>e.g.</i> a trap failure) Preferably, a zero indicates was looked for and not found, and a NA represents was not looked for/trap failure/ etc. Blank values are discouraged	
GPS obfuscation information	If GPS data obfuscation (<i>e.g.</i> GPS points are intentionally offset from their actual locations) or de- resolution occurs (<i>e.g.</i> GPS precision is intentionally reduced), a statement on the manner by which this occurred.	The highest resolution data (<i>e.g.</i> trap-level, specific GPS location) are the most useful. It is hoped that no data obfuscation / de- resolution occurs	"GPS locations have been truncated to 3 decimals" "GPS locations obfuscated using N- Dispersion" "No GPS deresolution was performed"
Data usage information	The data reuse policy for your data. Please provide a creative commons license identification. See https://creativecommons.org for more information.	For data to be F.A.I.R., it must be Reusable. We therefore recommend data be provided as "CC0" or "CC BY 4.0". "CC0", under which data are made available for any use without restriction or particular requirements on the part of users "CC BY 4.0", under which data are made available for any use provided that attribution is appropriately given for the sources of data used, in the manner specified by the owner (<i>e.g.</i> citation).	"CCO" or "CC BY 4.0"

Table 2. The MIReAAD data fields. Fig 1B provides an annotated example.

Field(s)	Details	Recommendations	Examples
Start Time (for collection)	Start time of the data sample collection. <i>e.g.</i> The trap was set	Be as specific as <i>practically</i> possible. Any unambiguous format is acceptable. However, do not use two-digit year abbreviations. If relevant, provide timezone in field or in header, a 24 hour clock is preferred, but should be made unambiguous as to which time format is being used.	"2012-04-27" "July 26, 2017" "2017-Jul-26" "2017-July-26 Morning " "2017-Jul-26 20:00 GMT "
End Time (for collection)	End time of the data sample collection. <i>e.g.</i> The trap was collected	See above. If instantaneous data collection (<i>e.g.</i> a tick drag), End Time may be the same as Start Time.	See above.
Location	The geographical location of sample collection.	As detailed as possible. Latitude and longitude if possible with specified accuracy Providing <i>both</i> a GPS point (decimalized GPS points are prefered) field and a geographical name field is prefered. Note only providing location <i>names</i> is highly discouraged as they change over time and can be ambiguous. Place / Trap names and GPS fields can be provided. If obfuscation was used, it should be indicated in the Metadata (Table 1). Splitting latitude and longitude further into two columns further reduces ambiguity.	"Kukar Maikiya, Jigawa State, Nigeria" "40.697" and " -74.015"

Collection method	Sampling apparatus (<i>e.g.</i> trap type, observation method)		"CDC light trap" "Tick drag" "Quadrat count" "BG Sentinel Trap" "Pitfall trap" "Larval dip" "Johnson suction trap" "Lindgren Funnel Trap"
Collection attractants	The attractant/ lures used to attract insects to a trap or collection		"None" "Carbon dioxide" "UV light" "BG-Sweetscent Mosquito Lure" "Human/animal bait"
Collection area	The spatial extent (area or volume) of the sample.	If relevant (<i>e.g.</i> , when collection method is transect or quadrat), in units of area or volume, the spatial coverage of the sampling unit Note this field would not typically be used for mosquito collections.	"100 m^2" "1 liter" "1 ha" "10m^3"
Taxonomy	Classification of sample collected.	Scientific genus and species preferred. Avoid abbreviation.	<i>"Ixodes scapularis</i> " "Aedes aegypti" 'Anopheles gambiae sensu stricto"

Unit(s) of measurement and observation	Description of exactly what was observed, the unit for "Value" below. For counts, should indicate life stage, sex, etc. Unit measures can be encoded into value field header. Consider multiple unit fields (e.g. separate fields for sex and stage.) See Figure 2.	Do not abbreviate. Coded data key should be provided in field name (<i>e.g.</i> "1 = species present 0= species absent")	"Number of individuals per m^2" "Female" and "Adult" "Male and Female" and "Nymphs"
Value	The numerical amount or result from the sample collection. Often this will be a quantity of observed individuals. Unit measures can be encoded into value field header. See Figure 2.	Units should be provided in a separate field.	"0" "23" "Yes" "Not present"
Additional sample information	This could be more than one field and should be used when more information is required to understand the experiment, for example experimental variables, sub- locations, etc. Some users may report wind speeds, temperatures, elevations etc.	Do not abbreviate.	"Forest" vs "Field" "Winter" vs "Summer" "Inside" vs "Outside" "200 meters above sea level"

Sample Name A human reason sample name May exist so the benefit o depositor in organizing the data, use the internal name conventions. May also be to tie related observations together.	e. restricted, but any encoded metadata should be revealed in the other datafields. For example, you may name a sample named 'Aphid1_StickyTrap_Jan4,' but you will still have "Sticky Trap" listed in a Collection Method field, and "Jan 4, 2017" in the date field. used	"Trap1_Night1" "KissingBug_2" "00004" "Jan08_animal_4,"
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Field names in bold should be considered also required. Remaining fields are optional or

282 depend on the complexity of the experimental design

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284 Author contributions

The project was conceptualized by Lauren Cator and Samraat Pawar. The original draft was
 prepared by Michael A. Johansson, Samuel S.C. Rund, Naveed Heydari, Kurt Vandegrift,

286 prepared by Michael A. Jonansson, Samuel S.C. Rund, Naveed Heydari, Kurt Vandegrift,

Matthew Watts, and Samraat Pawar. Visualization was prepared by Kurt Vandegrift, Samuel
 S.C. Rund, Samraat Pawar, and Michael A. Johansson. Review & Editing was performed by all

S.C. Rund, Samraat Pawar, and Michael A. Johansson. Review & Editing was performed by all
 the authors.

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291 Competing interest statement

292 The authors declare no competing interests.

293 Acknowledgements

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295 The seeds of this effort were planted in 2016 at a meeting of VectorBiTE, which is a cross-

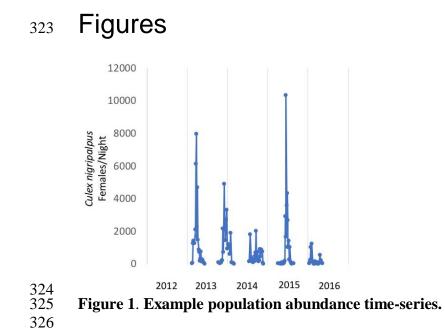
296 disciplinary research coordination network (RCN) for disease vectors. Samuel S.C. Rund,

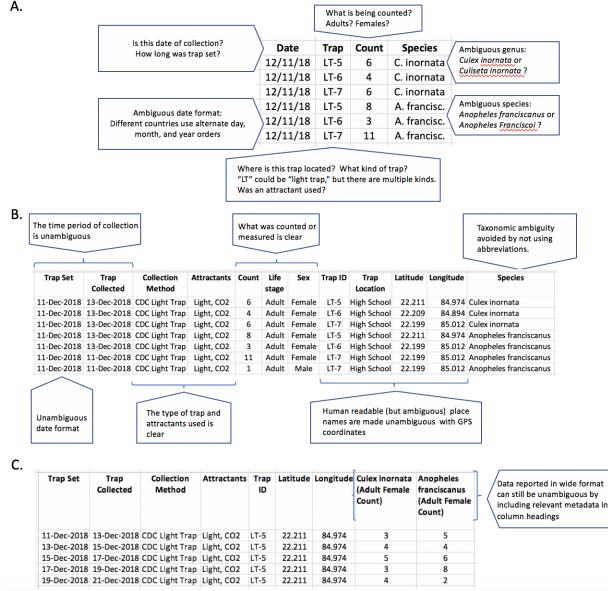
297 Matthew Watts, Kurt Vandegrift, Naveed Heydari, Cynthia Lord, Michael Johansson, Samraat

298 Pawar, and Sadie J. Ryan, received travel funding from NIH grant 1R01AI122284-01 and

- 299 BBSRC grant BB/N013573/1 as part of the joint [NIH-NSF-USDA-BBSRC] Ecology and
- 300 Evolution of Infectious Diseases program.

302	Samuel S.C. Rund was funded by the Royal Society (NF140517). Rund, Daniel Lawson, Robert
303	M. MacCallum, Sarah A. Kelly, Gloria I. Giraldo-Calderon and Scott J. Emrich were supported
304	by the National Institute of Allergy and Infectious Diseases, National Institutes of Health,
305	Department of Health and Human Services, under Contract No. HHSN272201400029C
306	(VectorBase Bioinformatics Resource Center).
307	
308	Kurt Vandegrift was funded by the National Science Foundation Ecology and Evolution of
309	Infectious Diseases program (1619072).
310	
311	Naveed Heydari and Sadie J. Ryan were funded by National Science Foundation (NSF DEB
312	EEID 1518681).
313	
314	Sadie J. Ryan was additionally funded by NIH 1R01AI136035-01, and CDC grant
315	1U01CK000510-01: Southeastern Regional Center of Excellence in Vector-Borne
316	Diseases: the Gateway Program. This publication was supported by the Cooperative Agreement
317	Number above from the Centers for Disease Control and Prevention. Its contents are solely the
318	responsibility of the authors and do not necessarily represent the official views of the Centers for
319	Disease Control and Prevention.
320	
321	Jennifer M. Zaspel was funded by the National Science Foundation Division of Biological
322	Infrastructure (NSF 1561448, NSF 1601957).





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328 Figure 2. MIReAAD reduces data ambiguity. A. Seemingly clean data can still lack key 329 information or have ambiguous metadata, hindering data reuse. B. MIReAAD compliant data 330 includes the metadata necessary for data reuse and removes ambiguity. C. Note data can be 331 formatted differently, but still be MIReAAD complaint, such as by presenting data in a wide 332 format

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