

1 **Abstract**

2 Curated databases of species interactions are instrumental to exploring and un-
3 derstanding the spatial distribution of species and their biotic interactions. In the
4 process of conducting such projects, data development and curation efforts may
5 give rise to a data product with utility beyond the scope of the original work,
6 but which becomes inaccessible over time. Data describing insect host-pathogen
7 interactions are fairly rare, and should thus be preserved and curated with appro-
8 priate metadata. Here, we introduce the `insectDisease` *R* package, a mechanism
9 for curating, updating, and distributing data from the *Ecological Database of the*
10 *World's Insect Pathogens*, a database of insect host-pathogen associations, includ-
11 ing attempted inoculations and infection outcomes for insect hosts and pathogens
12 (bacteria, fungi, nematodes, protozoans, and viruses). This dataset has been uti-
13 lized for several projects since its inception, but without a well-defined, curated
14 and permanent repository, its existence and access have been limited to word-of-
15 mouth connections. The current effort presented here aims to provide a means to
16 preserve, augment, and disseminate the database in a documented and versioned
17 format. This project is an example of the type of effort that will be necessary to
18 maintain valuable databases after the original funding disappears.

19 **Running title:** Ecological Database of the World's Insect Pathogens (EDWIP)

20

21 **Introduction**

22 There are a number of data sources documenting host-pathogen associations, es-
23 pecially for pathogens of mammals (Gibb et al., 2021, Patrick et al., 2017), birds
24 (Bensch et al., 2009), and fish (Strona and Lafferty, 2012). Recent work from
25 the Verena Consortium has developed a dynamically updated host-virus associa-
26 tion database for all vertebrate hosts (VIRION) (Carlson et al., 2021), representing
27 the largest collection of host-virus association data to date. These resources have
28 been fundamental to our understanding of what determines pathogen host range,
29 pathogen species richness across a set of hosts, and overall host-pathogen network
30 structure (e.g, Carlson et al. (2020), Dallas et al. (2018)). But while some host
31 groups are well-studied, there are taxonomic gaps in our understanding of host-
32 pathogen associations. Insect host-pathogen relationships have considerably less
33 open-source data available, despite their inherent importance to scientific studies
34 and assessments of impacts to agricultural crops and spread of vector-borne dis-
35 ease, in addition to the sheer numerical dominance of insect species over other
36 taxa (Stork et al., 2015). This is a clear knowledge gap.

37 Many of the existing species interaction databases have dedicated researchers,
38 resources, and infrastructure to enable data deposition and curation in openly
39 accessible formats. However, some data have not been as lucky, at no fault of the
40 original data curators. These data run the risk of disappearing into a file drawer or
41 on an external hard drive, potentially shared with a small number of researchers

42 but not accessible to the scientific community at large. One data resource arguably
43 close to this point of disappearance is the *Ecological Database of the World's Insect*
44 *Pathogens* (EDWIP) (Onstad, 1997).

45 The EDWIP data consist of experimental infections and field observations of
46 the interactions between insect hosts and a number of bacterial, fungal, nema-
47 tode, protozoan, and viral pathogens (Braxton et al., 2003). One particularly
48 unique component of EDWIP is the existence of negative associations – attempts
49 to inoculate a host with a given pathogen that failed to infect – for some host
50 groups (Figure 1). Failed infections represent *true* absences or incompatibilities
51 between a given host and pathogen. These data are incredibly useful to pathogen
52 host range estimation and host-pathogen interaction modeling, but we rarely have
53 data on these known non-interactions.

54 Initially created in 1992, the data have been updated prior to 2000, but no clear
55 semantic versioning was used. As such, it is unclear how long or how frequently
56 this updating and curation continued, and thus, how many different versions of
57 the data may be in existence presently. The database we present here, as the
58 backbone of this R package, represents the most up-to-date version that we know
59 of, though this may differ slightly from previous descriptions of the data (Braxton
60 et al., 2003). Generally, we have attempted to preserve all of the original data in
61 the original format.

62 **Solution statement**

63 To preserve these data in a format that is well-documented, openly accessible, ver-
64 sioned, and flexible for continued development, we created the **`insectDisease`**
65 R package. In doing so, we implicitly adhere to the FAIR (Findable, Acces-
66 sible, Interoperable, Resuable) guidelines for managing data (Wilkinson et al.,
67 2016). By hosting the data openly on GitHub, and versioning releases of the data
68 with a permanent identifier (DOI), we ensure the longevity and versioned cura-
69 tion of this data resource. Finally, the incorporation of taxonomic data through
70 **`taxize`** (Chamberlain and Szöcs, 2013) ensures that host and pathogen taxonomic
71 names are updated periodically to accommodate for dynamic data or changing
72 taxonomies.

73 **Data specification**

74 **Package structure** Data products are broken down by pathogen group; ne-
75 matodes (**`data(nematode)`**), viruses (**`data(viruses)`**), and non-viral pathogens,
76 which include protozoan, fungi, and bacteria (**`data(nvpassoc)`**). Data on neg-
77 ative associations is stored collectively instead of being delineated by pathogen
78 group (**`data(negative)`**), but information on pathogen group is provided within
79 each of these files, allowing for sorting of negative interactions based on the initial
80 pathogen groupings (Table 1). This data structure is inherited from the original
81 structure of the EDWIP data files, and code to process and join these different

82 data files is provided in the *R* package vignette.

83 Each of the pathogen groups differs slightly in the available ancillary data on
84 experimental infections. For instance, nematode infections contain information on
85 soil type and associated bacteria, virus infection data have information on viral
86 dose, and non-viral pathogens (protozoans, fungi, and bacteria) have information
87 on intermediate host species. We recommend the user explore these data and
88 associated metadata from within *R*, as the metadata and data are neatly in the
89 same place.

90 Data are also available on the insect host species themselves (e.g., `data(hosts)`).
91 These data contain some information on the Canadian province where the host is
92 found (`ProvinceI` column), what it eats (`Food` column), and what type of habitat
93 it is found in (`Habitat` column). Additionally, a column on host insect pest status
94 is present, offering the opportunity to explore study effort and pathogen specificity
95 dependent on the pest status of the insect host.

96 **FAIR data** The FAIR principles represent guidelines for making data more per-
97 sistent, findable, and well-documented. Structuring the data as an R package
98 ensures that metadata and data are packaged together, where R manual files con-
99 tain column names and data descriptions for each data product (*Findable*). All
100 code to take data from the raw data (`data-raw` folder) to the end product `.RData`
101 and `.csv` files is contained in the versioned R data package, and integration with

102 Zenodo (<https://doi.org/10.5281/zenodo.5821896>) provides a DOI for each
103 release (*Accessible*). Apart from providing data in these multiple formats, user
104 access is aided by structuring the data as a package in a very popular comput-
105 ing language among biologists (and other folks too) and providing all code for
106 data processing and serving in an open and public-facing repository (*Interoper-*
107 *able*). Having all code and data in a streamlined, open, and versioned format,
108 serving the data through an interactive web portal, and publishing this software
109 note collectively serve to promote the use of this data resource (*Reusable*).

110 **Metadata and package documentation** Differences in features across the
111 data on different pathogen types (e.g., `?nematodes` relative to `?viruses`) make
112 combining these data non-straightforward, without a degree of loss of information.
113 We provide some example code in the package `vignette` on how to go about
114 combining or linking the data across types, with the caveats of information loss,
115 and have standardized some key column names across the different data products.
116 Further, we have documented each data resource using *R* package documentation,
117 allowing the metadata of each data product to be examined directly from R using
118 the `help()` function or the question mark notation (e.g., `?viruses`).

119 **Data cleaning and taxonomic resolution** We attempted to maintain as much
120 of the original data structure from the raw data files provided by David Onstad,
121 principal maintainer of the EDWIP data resource (Onstad, 1997). This includes

122 files such as `new_assoc`, as this was likely a test file containing pathogen species
123 such as “wormy thing”, and `newnema`, a dataset identical to `nematode`. We docu-
124 ment these idiosyncrasies in the metadata for each data product, providing a clear
125 overview of the state of each data subproduct.

126 The first, and perhaps most important, novel augmentation, is the resolution of
127 host and pathogen taxonomic information. We achieved this by using the R pack-
128 age `taxize`, specifically the NCBI taxonomic backbone (Chamberlain and Szöcs,
129 2013), making the data interoperable with existing data efforts by the Verena
130 Consortium (e.g., VIRION; Carlson et al. (2021)). Cached versions of host and
131 pathogen taxonomic information are provided (`data(hostTaxonomy)` and
132 `data(pathTaxonomy)`), and the *R* code to generate these taxonomic backbones
133 and clean the data are provided in the package vignette. This taxonomic backbone
134 serves to both standardize host and pathogen nomenclature, while also correcting
135 any taxonomic changes that have occurred in the past couple decades. This in-
136 cludes the consideration of microsporidian parasites as fungi, not protozoans, a
137 change affecting a large set of records in the EDWIP data. All of the data within
138 the `data` and `csv` folders have already gone through these data cleaning steps.
139 However, these data may be dynamic, such that some form of continuous integra-
140 tion or updating of the host and pathogen taxonomy may be necessary. As such,
141 we provide a vignette which transparently shows the steps to clean and augment
142 the data resource, as well as reproduce figures from this manuscript. Finally, we

143 opt to store processed data in the `csv` folder, which contains all data files in `.csv`
144 format. This allows non-*R* users to access the csv-formatted data easily, and en-
145 sures long-term stability of the data, as `csv` is a stable text file format. These data
146 are also provided as `.rda` files in the `data` folder.

147 Maintaining the data dynamically as described above allows users to access
148 the data programmatically or as versioned flatfiles (i.e., `.csv` files). However, for
149 users who do not wish to download the entire data resource, and simply want to
150 quickly query a static version of the database, there is also a standalone web user
151 interface (<https://edwip.ecology.uga.edu/>) that allows users to easily subset
152 and explore the data. The web interface serves arguably the most important
153 subset of the overall data (data files `nematode`, `viruses`, `nvpassoc`, `negative`, and
154 `hosts`). This interface allows users to quickly query based on host or parasite
155 taxonomy as a dropdown list. This is perhaps more useful as a teaching tool or
156 for initial exploration of the data, while the programmatic interface and dynamic
157 data may be more useful for more rigorous analysis. This version of the EDWIP
158 data will also only be deployed with a single static copy of the data, such that
159 users wanting to benefit from versioned and dynamic data will need to access the
160 data through the GitHub repository. Future efforts to integrate the web interface
161 and the existing dynamic data structure will be explored, but this is not currently
162 integrated.

163 **Case study: covariance among pathogen groups in parasite** 164 **species richness**

165 Hosts that are infected by more pathogens of one type may also be more infected by
166 pathogens of another type, mediated by host life history traits, metabolic demands,
167 geographic distribution, and intensity of scientific study (Dallas and Becker, 2021).
168 We explore this in the EDWIP data by measuring the number of known positive
169 associations of each of the pathogen groups for each insect host species, visualizing
170 the relationship between the number of pathogens per insect host as a correlation
171 matrix (Figure 2). We find very little evidence that pathogen groups have positive
172 covariance, which would be expected if host species traits or trait-based sampling
173 biases drove infection process across pathogen groups in the same manner. The
174 failure to detect strong positive relationships, and indeed some negative relation-
175 ships appearing, could be a signal of the targeted nature of data collection, as
176 many insect host species were selected to study due to their potential as a crop
177 pest, and many pathogens were selected to study based on their potential use as
178 biocontrol or perhaps for their ease of culture.

179 This potential sampling bias among insect host species would be evident if there
180 were a positive relationship between the number of positive interactions and the
181 number of negative interactions for a host species, as it would indicate that host
182 species with lots of known interactions also tended to appear in many studies and

183 have some negative interactions as well. We find evidence for a significantly neg-
184 ative relationship based on a Spearman’s rank correlation ($\rho = -0.1$, $p < 0.0001$),
185 indicating no discernible influence of this relationship. This does not imply that
186 there is no sampling bias in the insect host species researchers opt to study, but
187 that such bias was not so strong as to be clearly detected.

188 **Concluding comments**

189 While ecological data are growing in availability, size, accessibility, and stability,
190 there are still data resources that are aging in place, and should not be allowed
191 to fade out of existence. The EDWIP data provided to the authors were in a
192 proprietary format (‘Claris FileMaker Pro 5’) that was already over 10 major
193 versions behind. With limited inter-version operability (e.g., `.fmp5` files cannot
194 be opened in more recent versions of the software, or require multiple conversion
195 steps), these data seemed as if headed towards obsolescence. The `insectDisease`
196 package ensures that these data will be available to the broadest set of researchers,
197 be bound to relevant metadata, and be properly versioned. By hosting the data
198 openly, we welcome contributions from researchers interested in augmenting the
199 data or building off the existing resource.

200 **Data accessibility**

201 The `insectDisease` R package is currently available on GitHub
202 (github.com/viralemergence/insectDisease), with ‘.csv’ files in the csv directory for
203 long-term data stability. GitHub releases of the data ensure versioning is main-
204 tained and all versions are accessible. At the time of this writing, the current
205 version is 1.2.0 (available at
206 <https://github.com/viralemergence/insectDisease/releases/tag/1.2.0>). Re-
207 leases are given a DOI through integration with Zenodo (<https://doi.org/10.5281/zenodo.58218>).

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Tables

Table 1: Files associated with the EDWIP data resource. Metadata is stored in *R* package documentation, allowing the data and metadata to be intrinsically linked. For instance, users can use the help functionality from within *R* to see more information on data columns and unit (e.g., `?nematode`).

filename	rows	columns	description
<code>assocref</code>	11005	16	references for some host-pathogen associations
<code>citation</code>	1966	7	references but no host-pathogen association information
<code>hosts</code>	4392	21	insect host trait data
<code>hostTaxonomy</code>	4489	7	host taxonomic data updated with the <code>getNCBI()</code> function
<code>negative</code>	529	21	information on negative host-pathogen associations
<code>nemaref</code>	338	5	references from nematode pathogens
<code>nematode</code>	234	24	host-nematode interaction data
<code>new_asso</code>	19	25	likely a training document (perhaps do not use)
<code>noassref</code>	569	16	references for some host-pathogen associations
<code>nvpassoc</code>	7164	23	non-viral pathogen infection data
<code>pathogen</code>	2041	9	pathogen trait data
<code>pathTaxonomy</code>	2282	7	pathogen taxonomic data updated with the <code>getNCBI()</code> function
<code>viraref</code>	2124	16	references from viral infections
<code>viruses</code>	1659	25	host-viral interaction data

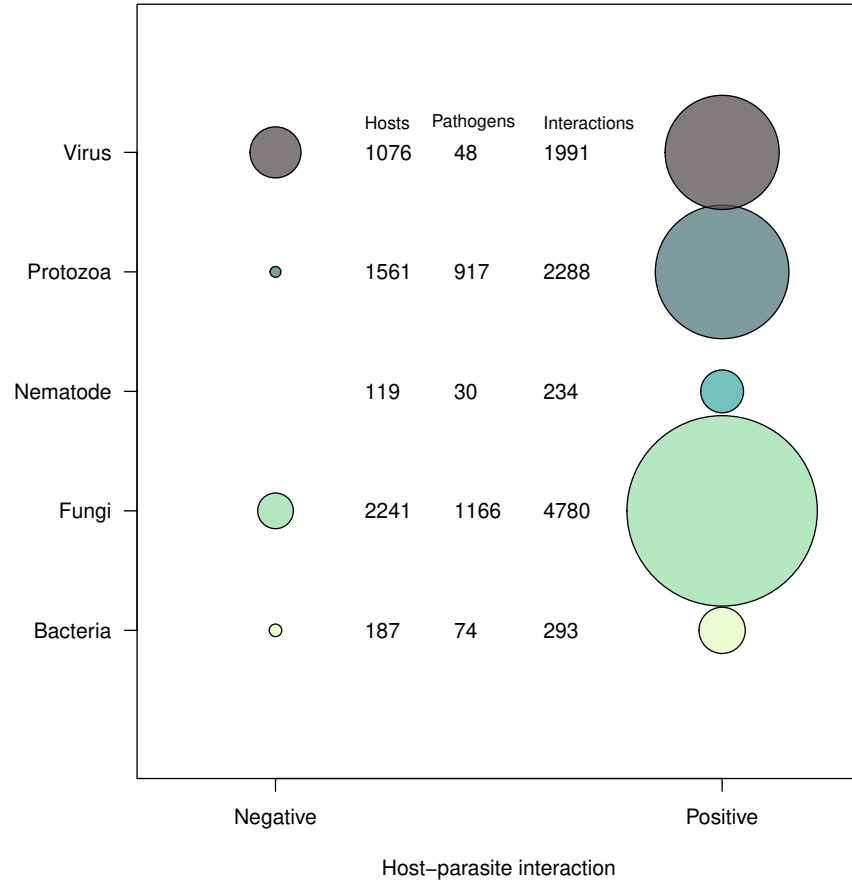


Figure 1: The number of known non-interactions (*negative* left panel) and known interactions (*positive* right panel) for the set of bacterial, fungal, nematode, protozoan, and viral pathogens (*y*-axis). Bubble size is proportional to the total number of interactions associated with that pathogen group and interaction type (i.e., *negative* or *positive*). Numeric columns correspond to the number of unique host species, pathogen species, and interactions for each pathogen group.

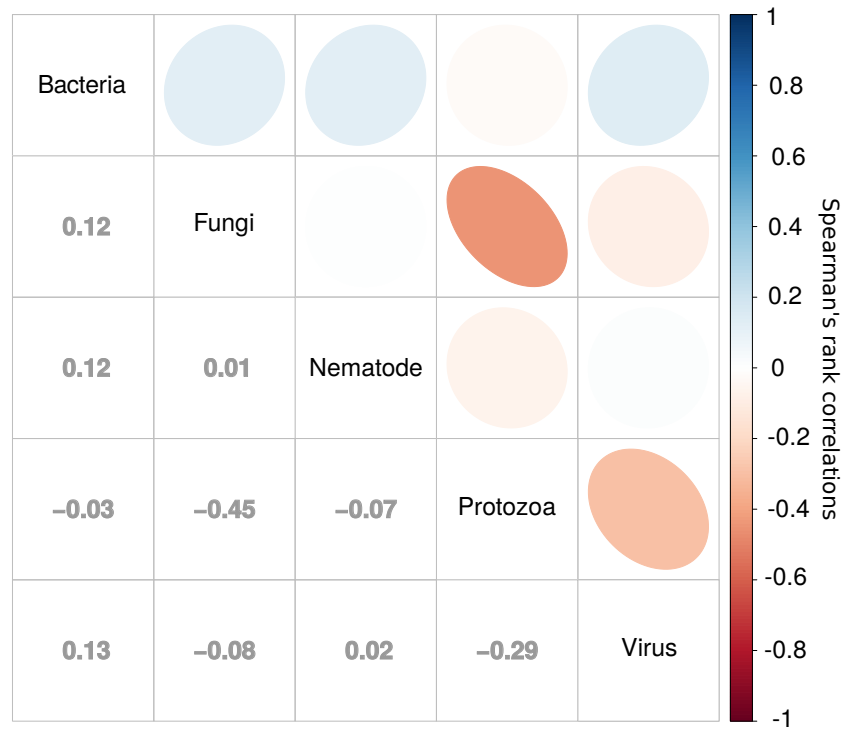


Figure 2: Correlations between each pathogen group in terms of pathogen richness of insect host species, where color corresponds to Spearman's rank correlation values (provided in the lower diagonal matrix). Fungal and protozoan pathogens were negatively related, as were viruses and protozoans. Understanding to what extent this is driven by sampling effects or insect host ecology is an outstanding research question that these could be used to begin addressing.