

insectDisease: programmatic access to the *Ecological Database of
the World's Insect Pathogens*

Tad A Dallas^{a,b,*} and Colin Carlson^c

^a*Department of Biological Sciences, Louisiana State University, Baton Rouge, LA, USA*

^b*Department of Biological Sciences, University of South Carolina, Columbia, SC, USA*

^c

^d

*Corresponding author: tad.a.dallas@gmail.com

Author contributions: TAD developed the R package, with taxonomic functions developed by CC. All authors contributed to manuscript writing.

Acknowledgements: We sincerely thank all individuals associated with the collection and curation of the *Ecological Database of the World's Insect Pathogens* data, specifically David Onstad at the US Department of Agriculture.

Conflict of interest: The authors have no conflicts of interest to declare.

Keywords: insect pathogens, crop pest, experimental infection,

¹ **Abstract**

² **Running title:** Ecological Database of the World's Insect Pathogens (EDWIP)

³

4 Introduction

5 There are a number of data sources documenting host-pathogen associations, es-
6 pecially for pathogens of mammals (Gibb et al., 2021, Patrick et al., 2017), birds
7 (Bensch et al., 2009), and fish (Strona and Lafferty, 2012). These have been in-
8 credibly important to our understanding of what determines pathogen host range,
9 pathogen species richness across a set of hosts, and overall host-pathogen net-
10 work structure (e.g., (Carlson et al., 2020, Dallas et al., 2018)). But while some
11 host groups are well-studied, there are taxonomic gaps in our understanding of
12 host-pathogen associations. For instance, insect host-pathogen relationships have
13 considerably less open-source data available, despite their inherent importance to
14 agricultural crops and vector-borne disease, in addition to the numerical dominance
15 of insect species over other taxa (Stork et al., 2015). This is a clear knowledge
16 gap.

17 Additionally, all of the previous data sources listed before have dedicated re-
18 searchers and government agencies responsible for data deposition and curation
19 in openly accessible formats. However, some data have not been as lucky, at no
20 fault of the original data curators. These data run the risk of disappearing into a
21 file drawer or on an external hard drive, potentially shared with a small number
22 of researchers but not accessible to the scientific community at large. One data
23 resource arguably close to this point of disappearance is the *Ecological Database*
24 *of the World's Insect Pathogens* (EDWIP) (Onstad, 1997).

25 The EDWIP data consist of experimental infections and field observations of the
26 interactions between insect hosts and a number of nematode, viral, protozoan, fun-
27 gal, and bacterial pathogens (Braxton et al., 2003). In the EDWIP data, a number
28 of pathogen infection types were excluded, including vector-borne pathogens which
29 do not infect the insect host. An interesting component of EDWIP is the existence
30 of negative associations – attempts to inoculate a host with a given pathogen that
31 failed to infect – for some host groups (Figure 1). Failed infections represent *true*
32 absences or incompatibilities between a given host and pathogen. These data are
33 incredibly useful to pathogen host range and host-pathogen interaction modeling,
34 but we rarely have data on these known non-interactions. Initially created in 1996,
35 the data have been dynamically updated over time. However, it is unclear how
36 long this curation continued, and how many different versions of the data may be
37 in existence without proper versioning. To this point, the numbers of insect-host
38 pathogen interactions differs slightly from previously published versions of the data
39 (Braxton et al., 2003), but these changes are relatively minor.

40 **Solution statement**

41 To preserve these data in a format that is well-documented, openly accessible,
42 versioned, and flexible for continued development, we created the `insectDisease`
43 R package. In doing so, we implicitly adhere to the FAIR (Findable, Accessible,
44 Interoperable, Resuable) guidelines for managing data. By hosting the data openly

45 on GitHub, and versioning releases of the data with a permanent identifier (DOI),
46 we ensure the longevity and versioned curation of this data resource. Finally, the
47 incorporation of taxonomic data through `taxize` (Chamberlain and Szöcs, 2013)
48 makes sure that host and pathogen taxonomic names are updated.

49 **Data specification**

50 **Package structure** Data products are broken down by pathogen group; ne-
51 matodes (`data(nematode)`), viruses (`data(viruses)`), and non-viral pathogens,
52 which include protozoan, fungi, and bacteria (`data(nvpassoc)`). Data on neg-
53 ative associations is stored collectively instead of being delineated by pathogen
54 group (`data(negative)`), but information on pathogen group is provided within
55 each of these files, allowing for sorting of negative interactions based on the initial
56 pathogen groupings.

57 Each of the pathogen groups slightly differs in the available data on experimental
58 infections. For instance, nematode infections contain information on soil type and
59 associated bacteria, virus infection data has information on viral dose, and non-
60 viral pathogens (protozoans, fungi, and bacteria) have information on intermediate
61 host species.

62 Data are also available on the insect host species themselves (e.g., `?hosts` or
63 `data(hosts)`). These data contain some information on Canadian province where

64 the host is found (`ProvinceI` column), what it eats (`Food` column), and what
65 type of habitat it is found in (`Habitat` column). Additionally, a column on host
66 insect pest status is present, offering the opportunity to explore study effort and
67 pathogen specificity dependent on the pest status of the insect host.

68 **Metadata and package documentation** Differences in features across the
69 data on different pathogen types (e.g., `?nematodes` relative to `?viruses`) make
70 combining these data difficult without a loss of information. However, there is
71 clear utility in having all of the data in a single standardized form. As such, we
72 have documented each data resource using R package documentation, allowing the
73 metadata of each data product to be examined directly from R using the `help()`
74 function or the question mark notation (e.g., `?viruses`).

75 **Data cleaning and taxonomic resolution** The initial data structure was
76 maintained from the original raw data files provided by David Onstad, principal
77 maintainer of the EDWIP data resource (Onstad, 1997). This includes files
78 such as `new_assoc`, as this was likely a test file containing pathogen species such as
79 "wormy thing". We clean and augment this existing data source programmatically,
80 with much of this code in the `insectDisease` package vignette.

81 First, we resolve host and pathogen names using the R package `taxize`, using
82 NCBI taxonomic backbone (Chamberlain and Szöcs, 2013). While this seems like
83 a small change, it both standardizes host and pathogen nomenclature, and catches

any taxonomic changes that have occurred in the past couple decades. This includes the consideration of microsporidian parasites as fungi, not protozoans, a change affecting a large set of records in the EDWIP data. Second, we provide a vignette which combines the different host-pathogen data together to form an informative set of host-pathogen associations, including both both known associations and failed experimental infections, as delineated by the `interaction` column in the resulting `edwip.csv` file. Finally, processed data by the vignette is output in `.csv` format in the `vignettes` folder, but all data stored as R data objects (`.rda`) are also converted to csv and placed in the `csv` folder, allowing non-R users to still access the data in a stable format.

Case study: covariance among pathogen groups in parasite species richness

Hosts that are infected by more pathogens of one type may also be more infected by pathogens of another type, mediated by host life history traits, metabolic demands, geographic distribution, and intensity of scientific study (Dallas and Becker, 2021). We explore this in the EDWIP data by measuring the number of known positive associations of each of the pathogen groups for each insect host species, visualizing the relationship between the number of pathogens per insect host as a correlation matrix (Figure 2). We find very little evidence that pathogen groups have positive covariance, which would be expected if host species traits drove infection process

104 across pathogen groups in the same manner. The failure to detect positive rela-
105 tionships, and indeed some negative relationships appearing, could be a signal of
106 the targeted nature of data collection, as many insect host species were selected
107 to study due to their potential as a crop pest, and many pathogens were selected
108 to study based on their potential use as biocontrol or perhaps for their ease of
109 culture.

110 This potential sampling bias among insect host species would be clear if there
111 were a positive relationship between the number of positive interactions to the
112 number of negative interactions for a host species, as it would indicate that in-
113 sect host species with lots of known interactions also tended to appear in many
114 studies and have some negative interactions as well. We find evidence for a signif-
115 icantly negative relationship based on a Spearman’s rank correlation ($\rho = -0.1$, p
116 < 0.0001).

117 **Concluding comments**

118 Ecological data – while growing in availability, size, and stability – are still a re-
119 source that should not be allowed to fade out of existence. The EDWIP data
120 provided to the authors were in a proprietary format (‘Claris FileMaker Pro 5’)
121 that was already over 10 major versions behind. With limited inter-version oper-
122 ability (e.g., `.fmp5` files cannot be opened in more recent versions of the software,
123 or require multiple conversion steps), these data seemed as if headed towards ob-

124 solescence. The `insectDisease` package ensures that these data will be available
125 to the broadest set of researchers, be bound to relevant metadata, and be properly
126 versioned. By hosting the data openly, we welcome contributions from researchers
127 interested in augmenting the data or building off the existing resource.

128 **Data accessibility**

129 The `insectDisease` R package is currently available on GitHub <https://github.com/viralemergence/>
130 with '.csv' files in the csv directory for long-term data stability.

131 Finally, the data are periodically archived at major version changes via Zenodo

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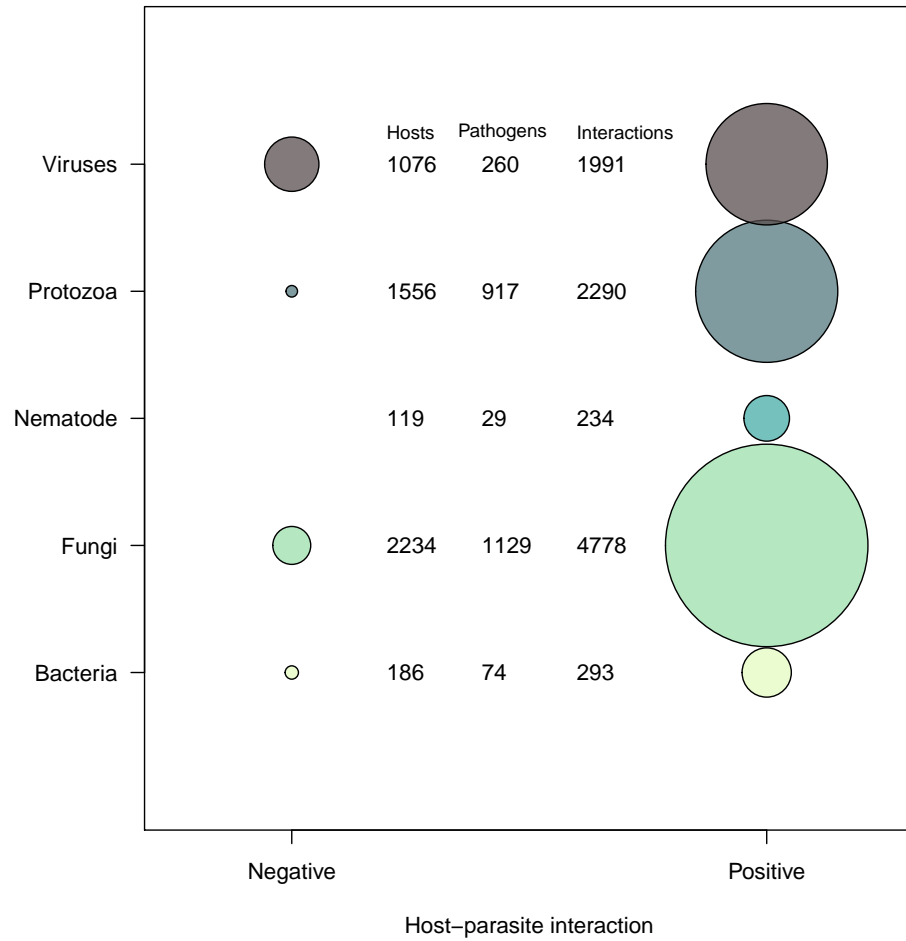


Figure 1: Bubble plot, where points are proportional to the total number of negative (on the left) and positive (on the right) host-pathogen interactions for each pathogen group (y -axis). Numeric columns correspond to the number of unique host species, pathogen species, and interactions for each pathogen group.

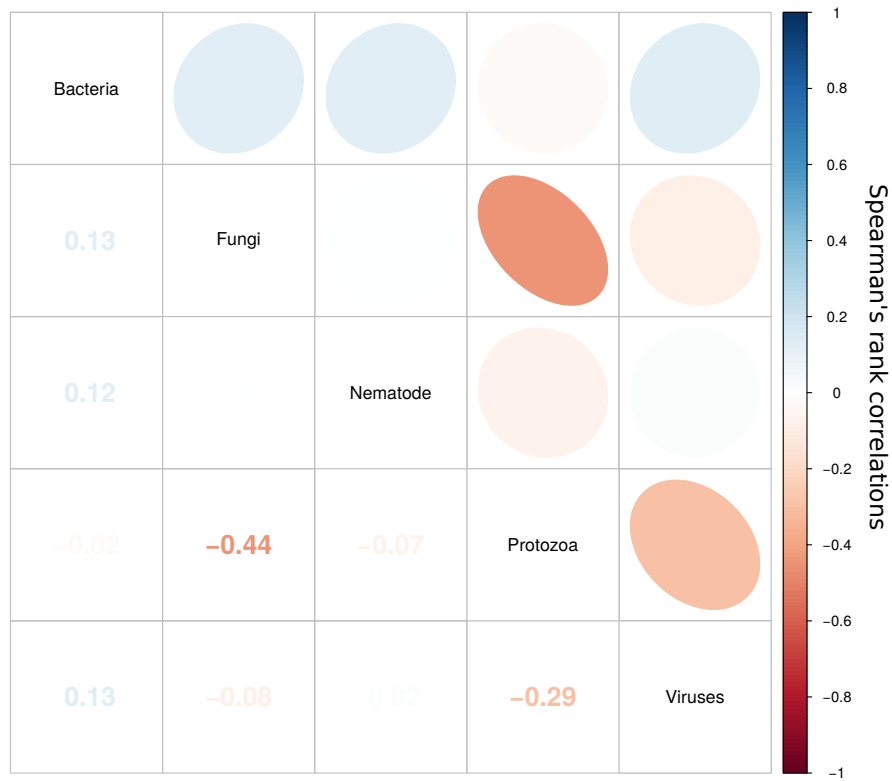


Figure 2: Correlations between each pathogen group in terms of pathogen richness of insect host species, where color corresponds to Spearman's rank correlation values (provided in the lower diagonal matrix). Fungal and protozoan pathogens seemed to negatively covary, as did viruses and protozoans. Understanding 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.