

dfoliatR: An R package for detection and analysis of insect defoliation signals in tree rings

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Abstract

We present a new R package to provide dendroecologists with tools to infer, quantify, analyze, and visualize growth suppression events in tree rings. *dfoliatR* is based on the OUTBREAK program and builds on existing resources in the R and *dplR* computing environments. It is designed to aid research in the ecology of insect defoliation events and to reconstruct defoliator outbreak chronologies, but can be applied to other studies where host–non-host comparisons are useful. *dfoliatR* performs an indexing procedure to remove climatic signals in the host-tree series that are represented in the non-host chronologies, or other annually-resolved climate series. It then infers defoliation events in individual trees based on user-specified thresholds. Site-level analyses identify outbreak events that synchronously affect user-defined numbers or proportions of involved host trees. Functions are provided for summary statistics and graphics of tree- and site-level series. We evaluated *dfoliatR* against OUTBREAK, using eight datasets including 222 host-trees, and found that *dfoliatR* improves on OUTBREAK with greater user control, identification of defoliation events, computing capacity, and both the statistical summary and graphical outputs. We provide two example data sets and script to enable users to gain familiarity with the package and its capabilities. The source code is available in the Comprehensive R Archive Network (CRAN) and on GitHub.

Key words: Dendroecology, spruce budworm, *Choristoneura*, pandora moth,

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14 **1. Introduction**

15 Variation in the width and morphology of annual radial growth rings in trees
 16 permits dating and quantification of past forest insect defoliator outbreaks. De-
 17 foliation can be distinguished from climate- and other disturbance-related influ-
 18 ences by comparing ring-width or other annually-resolved features in the wood
 19 of host species to that of non-host species or annually-resolved climate records.
 20 The effect of defoliation on radial growth of trees has been recognized since the
 21 1860s, and used to reconstruct outbreak regimes since the 1950s (Blais, 1954;
 22 Lynch, 2012). It was not until the 1980s, however, that precise dendrochronolog-
 23 ical techniques were applied for inferring defoliation events and reconstructing
 24 defoliator outbreak regimes (Swetnam et al., 1985; Speer, 2010; Lynch, 2012).
 25 The first studies (Swetnam et al., 1985; Swetnam and Lynch, 1989, 1993) fo-
 26 cused on developing historical outbreak reconstructions of western spruce bud-
 27 worm (WSBW; *Choristoneura freemani* (Lepidoptera: Tortricidae); previously
 28 known as *C. occidentalis*). The methodology has since been successfully applied
 29 to a wide range of defoliator species, most of which are conifer herbivores, and
 30 has evolved in sophistication for a wide range of ecosystem situations (Lynch,
 31 2012).

32 The main dendrochronological tool for inferring, dating, and characteriz-
 33 ing defoliator outbreaks from tree-ring records has been the software routine
 34 OUTBREAK (Swetnam et al., 1985; Holmes and Swetnam, 1986; Swetnam
 35 and Lynch, 1989). OUTBREAK computes indices (described later in detail) of
 36 suppressed growth by subtracting a detrended and standardized climate series
 37 (a “control” chronology) from individual host-tree detrended and standardized
 38 radial growth series after the host and non-host series have been brought to a
 39 common variance. The non-host chronology usually consists of a site chronol-
 40 ogy developed from non-host tree species growing on a climate-sensitive site,
 41 but a gridded climate data point series, like the North American Drought Atlas

42 (Cook and Krusic, 2004) also suffices. If the host and non-host species respond
43 similarly to climate (which can and should be tested), the derived series retains
44 variability that the host and non-host series do not have in common, generally
45 the insect signal and some unexplained variability (noise). The user defines
46 a rule base specifying the magnitude and duration that a period of indexed
47 growth suppression must meet or surpass for a period of suppressed growth to
48 be inferred as a defoliation event at the tree level.

49 Though powerful, OUTBREAK is outdated and increasingly difficult to use
50 in modern computing environments. It was written in FORTRAN V with in-
51 herently severe restrictions, as RAM and disk space were limited at that time
52 (256 kb and 10 MB, respectively) and FORTRAN conventions imposed very
53 strict formatting, file naming, and output conventions. The program lacks a
54 graphical interface or capabilities, forcing users to import generated text files
55 into spreadsheets or other software to assess results and perform analyses. Fur-
56 thermore, OUTBREAK can only handle one test at a time, creating barriers
57 to batch operation and a large burden for researchers with datasets including
58 multiple sites. We developed **dfoliatR** (Guiterman et al., 2020) as an R- and
59 **dp1R**-based library to overcome these issues.

60 **dfoliatR** adds to a growing suite of dendrochronology packages in the R
61 computing environment (R Core Team, 2019). Stemming from the **dp1R** library
62 (Bunn, 2008) that enables R users to read and write an array of tree-ring data
63 formats, standardize ring width series, build and evaluate chronologies, and per-
64 form quality control (to name a few), one can now also measure ring widths from
65 scanned images of prepared samples (Lara et al., 2015; Shi et al., 2019), conduct
66 and check crossdating (Bunn, 2010), analyze sub-annual anatomical features
67 (Rathgeber et al., 2011; Campelo et al., 2016), and perform many analytical
68 tests (Zang and Biondi, 2015; Jevšenak and Levanič, 2018). Tools for assessing
69 stand dynamics and disturbance analyses are under rapid development, with
70 new packages for assessing growth and release events (**TRADER**: Altman et al.,
71 2014), metrics of growth resilience (**pointRes**: van der Maaten-Theunissen et
72 al., 2015), and fire history (**burnr**: Malevich et al., 2018). The key objective

of `dfoliatR` is to provide tools to identify and analyze insect defoliation and outbreak events by building on the methods employed by `OUTBREAK`. It capitalizes on the robust software already available in `R` by using `dplR` data formats for incoming tree-ring series and providing output data formats embodied by the `tidyverse` (Wickham et al., 2019) that include efficient data manipulation (`dplyr`: Wickham et al., 2020) and graphics (`ggplot2`: Wickham, 2016).

In this paper, we describe the statistical methods employed by `dfoliatR`, compare results to those produced by `OUTBREAK`, and present an example analysis including test data sets and script. Users need not have much experience in `R` to replicate the analyses and graphics as presented. The `R` code below is executable in an `R` session once the required libraries are installed and loaded. Support documentation in addition to this paper is provided within the package via standard help menus and on the package website (<https://chguiterman.github.io/dfoliatR/>), which includes up-to-date vignettes that describe various routines. Code to generate a preprint of this manuscript, including the `R` scripts and tabular and graphical output is available from https://github.com/chguiterman/dfoliatR_paper.

2. Overview of the software

The `dfoliatR` library requires two sets of tree-ring data to infer defoliation and outbreak events:

- Standardized ring-width series for individual trees of the host species
- A standardized tree-ring chronology from a local non-host species, or a climate reconstruction

Users can develop these data sets in the software of their choosing, such as `dplR` or `ARSTAN` (Cook and Holmes, 1996). It is important that the host-tree data include only one tree-ring series per tree. Both `dplR` and `ARSTAN` have options for averaging multiple sample series into a tree-level series.

At the heart of `dfoliatR` lies two functions: `defoliate_trees()` and `outbreak()`. These identify defoliation events on individual trees (Figure 1) and then composite across multiple trees to identify outbreak events (Figure 2).

2.1. Identifying Defoliation of Trees

The `defoliate_trees()` function is the point of entry to the `dfoliatR` library. It performs two processes, removing climate-related growth signals from the host-tree series and identifying tree-level defoliation events. The climatic or non-defoliation signals in each host-tree series are characterized by a non-host chronology or climate reconstruction. `dfoliatR` removes the non-defoliation signal by subtracting the non-host series from each host-tree series, which generates a residual index. In `OUTBREAK`, this residual index was termed the “corrected index.” We call it the “growth suppression index” (GSI). The GSI is calculated the same as in `OUTBREAK` for each host tree as

$$\text{GSI}_i = H_i - (\text{NH}_i - \overline{\text{NH}}) \frac{\sigma_H}{\sigma_{\text{NH}}} \quad (1)$$

where H and NH are the host-tree series and the non-host chronology, in year i , respectively. Only the common period between the host-tree series and the non-host chronology are used in Equation 1. The host and non-host chronologies are brought to common variance by scaling the non-host chronology by its mean ($\overline{\text{NH}}$) and multiplying by the ratio of host and non-host standard deviations ($\frac{\sigma_H}{\sigma_{\text{NH}}}$), which approximates the variance of the host tree series.

Negative departures in the normalized GSI (NGSI, or GSI converted to z-scores) that surpass user-specified thresholds in duration and magnitude are defined as *defoliation events*. As in `OUTBREAK`, the lowest NGSI value in the particular sequence being assessed must reach the magnitude threshold. The default setting is -1.28 (NGSI is in units of standard deviation), which was previously determined to be representative of WSBW effects (Swetnam and Lynch, 1989) and is commonly used for other species. The year with the lowest value is termed the “year of maximum departure” and becomes a central point in time for assessing other thresholds before being included as a defoliation event.

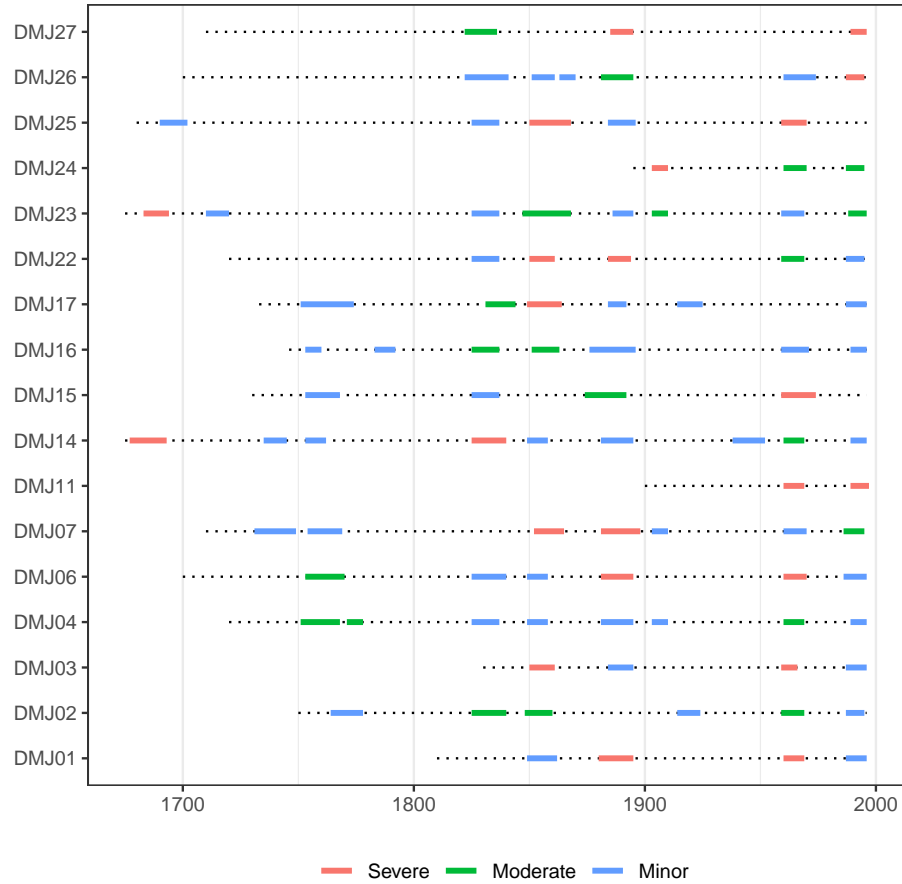


Figure 1: Default `dfoliatR` graphics for individual trees, produced by the `plot_defol()` function. Dotted lines represent the series length for each tree and colored segments show periods of defoliation. Users can define the cut-off values to determine severe–moderate–minor defoliation intensities, and use standard `ggplot2` graphics parameters to adjust styles and themes.

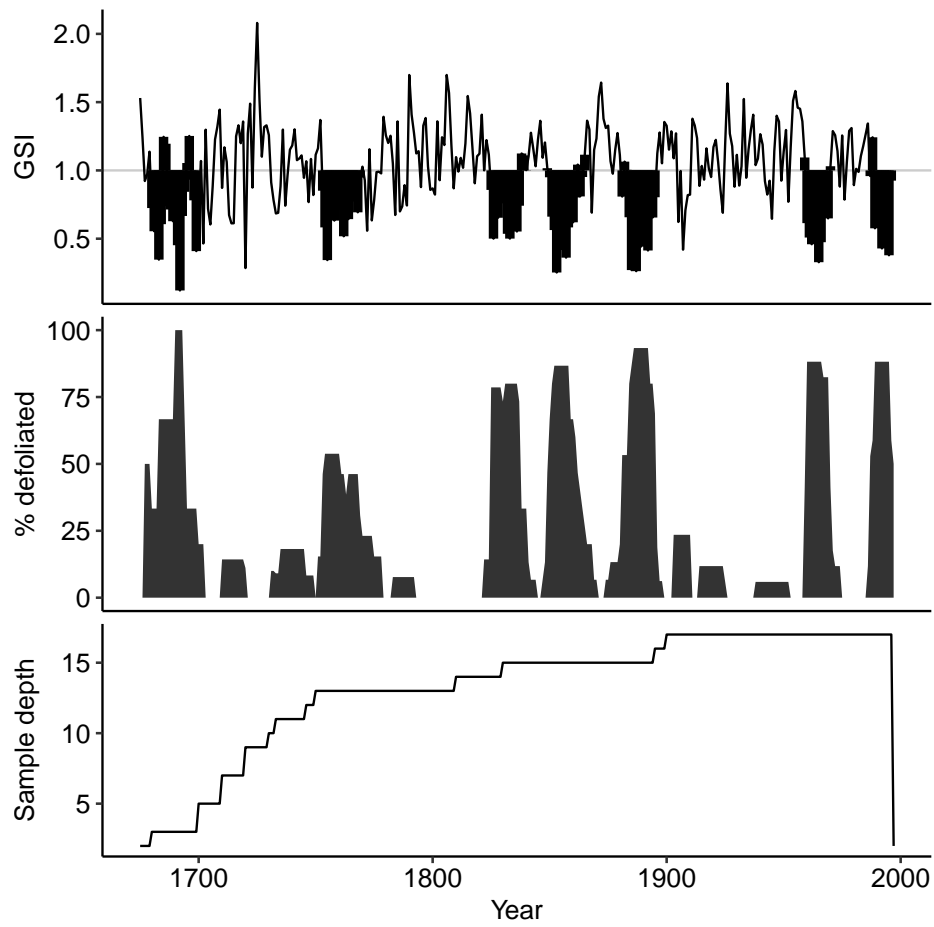


Figure 2: Default graphics showing outbreak events at the DMJ site. Top panel shows the mean GSI through time, with inferred outbreak events filled. The middle panel shows the percent of trees defoliated, and the bottom panel shows the number of trees through time.

119 If the year of maximum departure is higher than the threshold (i.e. $\text{NGSI}_{\text{lowest}} >$
120 -1.28), the sequence being assessed is omitted from the event results.

121 Event duration is assessed by examining sequences of negative NGSI (that
122 exceed the magnitude threshold) before and after the year of maximum depart-
123 ure year. Each defoliation event is allowed one single-year positive excursion
124 on each side of the year of maximum departure. Duration is computed across
125 the entire sequence that may include these two positive excursions. As in OUT-
126 BREAK, the user specifies a duration threshold (minimum number of years) for
127 a departure sequence to be inferred as a defoliation event. The default thresh-
128 old is eight years, as is commonly used in WSBW studies (Swetnam and Lynch,
129 1989). If the sequence is shorter than the duration threshold, the sequence is
130 omitted from the event results (i.e., both thresholds must be met). Researchers
131 can, and should, adjust the duration and magnitude parameters accordingly
132 and critically evaluate the results, as insect species vary in the length of their
133 outbreaks and the degree to which they can suppress tree growth.

134 Like OUTBREAK, users are provided an option to suspend the duration
135 threshold at the recent end of the series in cases where an outbreak event is
136 known to be ongoing. This should be used if the user has direct knowledge of
137 defoliation at the site during the sampling campaign. The advantage of allowing
138 potentially short, series-end events is that it allows the current event to be
139 included in return-interval estimates, and can aid in identifying the start-year
140 for the current defoliation event or outbreak.

141 Diverging from OUTBREAK, `dfoliatR` allows users to extend defoliation
142 events on individual trees by bridging between sequential events (Figure 3). In
143 cases where two defoliation events are separated by a single year, bridging will
144 link them into a single event. This option was added to `dfoliatR` during the
145 testing phase of development, when we realized that OUTBREAK deliberately
146 omits sequential, or back-to-back events, even when these events surpass both
147 the magnitude and duration thresholds. Instead, OUTBREAK will select the
148 one sequential event with the lowest negative departure year. In every case
149 we assessed (described below) we felt that the OUTBREAK-omitted defolia-

150 tion events should have been maintained and recorded. Due in large part to
 151 reconstructions using OUTBREAK (see papers cited by Lynch, 2012), we now
 152 know considerably more about forest defoliator outbreak regimes than we did
 153 in the 1980s when OUTBREAK was under development. We think that two or
 154 more prolonged events separated by a single year should in some situations be
 155 considered a single event. This is particularly relevant to WSBW and spruce
 156 budworm (*C. fumiferana*), for which outbreak regimes have been reconstructed
 157 for several geographic areas and for which researchers and forest health ex-
 158 perts have gained considerable knowledge (Brookes et al., 1982, and many later
 159 publications; Schmitt et al., 1984; Sanders et al., 1985). In many cases, the
 160 greatest growth suppression often occurs late in the outbreak due to cumula-
 161 tive effects accrued on a tree’s resources (Swetnam and Lynch, 1989, 1993).
 162 We urge caution in using the bridging option, however, because it may not be
 163 appropriate for all studied insects, such as in situations where impacted stands
 164 barely recover from one outbreak before another begins, as with pine procession-
 165 ary caterpillars (*Thaumetopoea pityocampa* (Lepidoptera: Thaumetopoeidae))
 166 (Carus, 2004, 2009).

167 2.2. Inferring Outbreak Events

168 Defoliation of one or a few trees does not constitute an outbreak. To de-
 169 termine when defoliation becomes an *outbreak event*, `dfoliatR` composites the
 170 individual tree defoliation series into a site-level chronology with the `outbreak()`
 171 function. Users have options to define the number and/or the proportion of trees
 172 required for an event to be considered an outbreak. Three parameters control
 173 whether a defoliation event constitutes an outbreak: the minimum number of
 174 trees available, the minimum number of trees recording defoliation, and the
 175 percent of trees recording defoliation. The first allows the researcher to make a
 176 judgement call as to the confidence ascribed to reduced sample depth toward the
 177 ends of their chronologies, thus compensating for the “fading record problem”
 178 (Swetnam et al., 1999). The second two parameters adjust the scale of defolia-
 179 tion considered to be an outbreak. Absolute numbers of trees and percentages

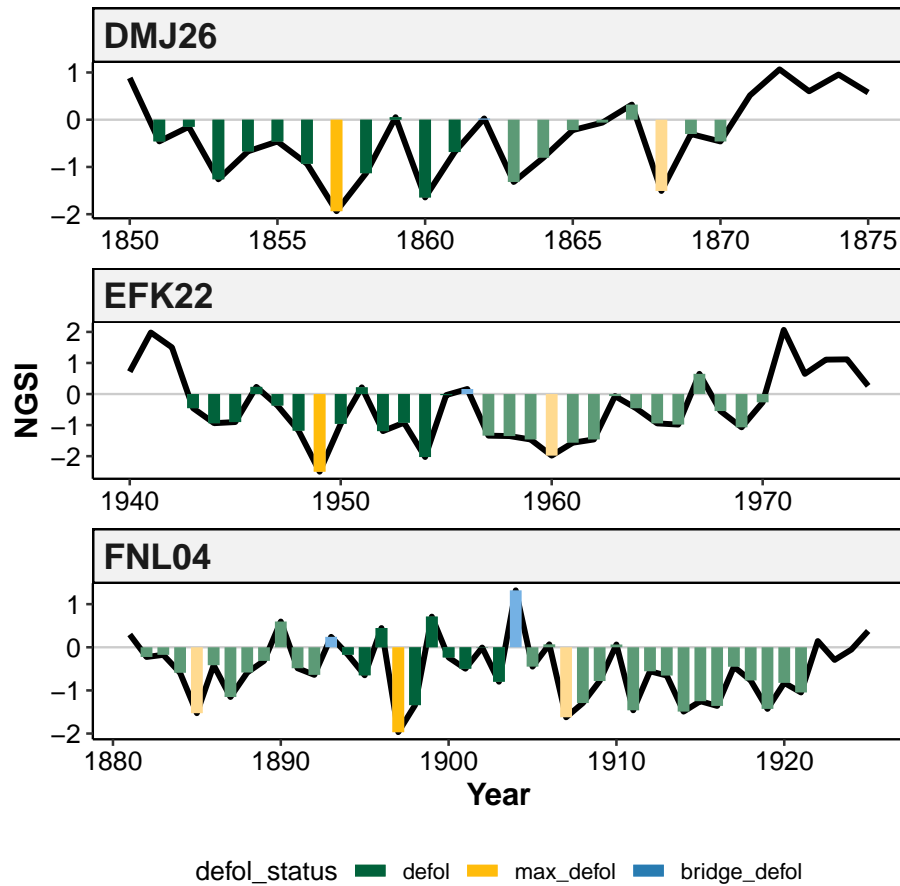


Figure 3: Examples of sequential defoliation events (bars), wherein two or three events occur separated by single years. In each case OUTBREAK recorded only one of the events, omitting those with a higher “max_defol” year, or the year of the lowest NGSI value. `dfoliatR`, by contrast, recorded all of the events, including two sequential events on DMJ26 and EFK22, and three on FNL04. Events recorded by both programs have a darker color, while those recorded only by `dfoliatR` are lighter in color. Adding the bridging option in `dfoliatR` links these events (“bridge_defol”) into single longer events that may more accurately capture the effects of long-term insect herbivory on individual trees.

180 can be applied separately or in conjunction, following filtering conventions in
181 tree-ring fire history studies (Malevich et al., 2018). We urge users to carefully
182 consider the choice of absolute numbers in situations where the number of trees
183 represented in the series varies with time, or the choice of percentages when
184 sample size is small.

185 **3. Evaluation**

186 *3.1. Approach*

187 We tested `dfoliatR` against OUTBREAK by comparing NGSI to OUT-
188 BREAK’s normalized corrected indices for individual trees and years, defoli-
189 ation status for individual trees and years, and percentage of trees recording
190 outbreaks at the site level. Our tests used standardized ring-width data from
191 eight sites in British Columbia, Colorado, and New Mexico, and author-provided
192 non-host site chronologies.

193 We detrended host data for both `dfoliatR` and OUTBREAK using AR-
194 STAN [v6.1] with cubic smoothing splines (50% frequency response on 100-150
195 year wavelengths depending on the site). In both `dfoliatR` and OUTBREAK
196 we used event thresholds of -1.28 normalized indices, 8 years duration, and al-
197 lowed for events at the end of series in seven of eight sites. These seven sites
198 were sampled during ongoing outbreak events (Swetnam and Lynch, 1993). We
199 found it necessary to be consistent in how we detrended and what software we
200 employed (e.g., ARSTAN vs `dplR`) because subtle differences in standardized
201 ring-width indices generated between the programs transferred into differences
202 between `dfoliatR` and OUTBREAK. In the end, we chose to only use the stan-
203 dardization output files from ARSTAN, which are easily read into `R` (and then
204 `dfoliatR`) using the `dplR` package.

205 The `R` code to replicate our comparisons is available from [https://github.](https://github.com/chguiterman/dfoliatR_paper)
206 [com/chguiterman/dfoliatR_paper](https://github.com/chguiterman/dfoliatR_paper).

207 3.2. Findings

208 Across the 43,280 ring-width indices from 222 trees included in our eval-
 209 uation, we found that `dfoliatR` and OUTBREAK compute identical growth
 210 suppression indices at 0.000 precision. We expected this outcome because both
 211 programs employ Equation 1. At the tree-level, the programs identified 11,530
 212 years of defoliation. The programs agreed on 97.9% of the years, leaving 927
 213 “difference” years in which only one program identified defoliation on an in-
 214 dividual tree. The differences included 102 events on 85 trees. We carefully
 215 inspected each of these events in the full context of each tree’s ring-series, and
 216 categorized the differences as follows

- 217 • *Series-end events* (40% of the total) in which OUTBREAK included “trun-
 218 cated outbreaks” (for seven sites) at the end of each series. In `dfoliatR`,
 219 this option is controlled by the “series_end_events” parameter to `defoliate_trees()`.
 220 In OUTBREAK, the option appears while changing the duration param-
 221 eter (option 3). When selected, OUTBREAK will include any sequences of
 222 negative indices at the either the beginning or the end of each tree series
 223 as a defoliation event, without consideration of either duration or magni-
 224 tude thresholds. In `dfoliatR`, the duration threshold is omitted and the
 225 magnitude threshold is retained in series-end-events. Each of the 13 events
 226 included in these differences did not meet the “max_reduction” parameter
 227 (-1.28 NGSI) in `dfoliatR` and were excluded. In two cases, OUTBREAK
 228 included events at the beginning of the series where `dfoliatR` does not
 229 allow truncated events. In four cases, OUTBREAK omitted only the last
 230 year of the series because the index was positive, but `dfoliatR` allowed this
 231 single positive excursion. Finally, there were two cases in which `dfoliatR`
 232 omitted possible events because it had already included a positive NGSI
 233 excursion after the “max_reduction” year, and since it will only allow one
 234 excursion on either side of the max year, the events were omitted due to
 235 short duration.
- 236 • *Sequential events* (36%) in which OUTBREAK omitted back-to-back events

that occur one year prior or one year following an identified event. When this occurs, OUTBREAK selects the one event sequence with the lowest negative index year (e.g., Figure 3). On two trees, OUTBREAK omitted two of three sequential events. While inspecting these differences, we added an option to `defoliate_trees()` that would “bridge” between sequential events (that surpass magnitude and duration thresholds) into single, long events. We felt that this was ecologically justified, especially for studies of WSBW, because the insect will occasionally defoliate individual trees for decades and reconstructed chronologies have shown outbreaks as long as **xxx (ANN CHECK)**.

- *Undetermined differences* (22%) occurred in cases where OUTBREAK omitted events without clear cause that `dfoliatR` correctly identified as defoliations.
- *Rounding differences* (2%) in the indices either omitted or cut short events on two trees. In both cases the indices were very close to zero, and the difference was less than the precision of the raw data measurement.

At the site level, OUTBREAK and `dfoliatR` produce similar time series of percent trees defoliated (Figure 4), which forms the basis for inferring outbreak occurrence, intensity, and duration. In nearly site comparison, `dfoliatR` included either more events or longer durations of inferred outbreak. These differences arise from the inclusion of tree-level events by `dfoliatR` that were omitted by OUTBREAK. The differences translate to greater numbers of trees in defoliation during certain events, or events recorded by single trees during periods of low sample depth.

This evaluation revealed what we believe are short-comings in how OUTBREAK identifies defoliation events on individual trees. In every one of the 102 cases we inspected, we felt that `dfoliatR` provided a more biologically and statistically appropriate assessment of defoliation. Therefore, `dfoliatR` also did a better job inferring outbreak events and associated statistics at the site level.

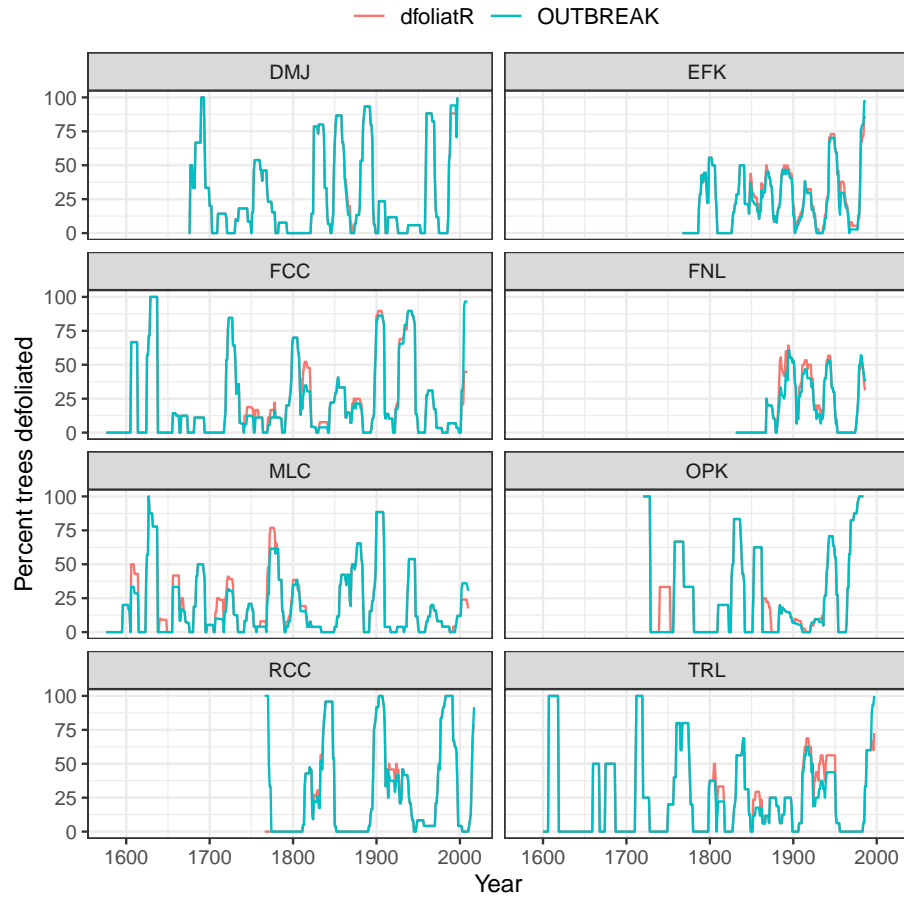


Figure 4: Comparison of reconstructed western spruce budworm outbreaks computed by **dfoliatR** and **OUTBREAK**. Input parameters were identical between programs. Difference arise because **dfoliatR** will identify and record more defoliation events on individual trees, and may more accurately capture herbivory dynamics.

266 4. Availability and installation

267 The `dfoliatR` library is provided free and open source from the Compre-
268 hensive R Archive Network (CRAN; <https://cran.r-project.org/>). To install
269 `dfoliatR` from CRAN use

```
install.packages("dfoliatR")
```

270 In each R session, `dfoliatR` can be loaded via

```
library(dfoliatR)
```

271 Development versions of `dfoliatR` are available on GitHub and installed
272 using the `devtools` library,

```
devtools::install_github("chguiterman/dfoliatR")
```

273 Issues, bug reports, and ideas for improving `dfoliatR` can be posted to
274 <https://github.com/chguiterman/dfoliatR/issues>. As an Open Source library,
275 we welcome and encourage community involvement in future development. The
276 best ways to contribute to `dfoliatR` are through standard GitHub procedures
277 or by contacting the first author.

278 5. Example Usage

279 Once `dfoliatR` is loaded into an R session (via `library(dfoliatR)`) users
280 can access two sets of tree-ring data to aid in exploring the functions, graph-
281 ics, and outputs. Each data set consists of individual host-tree series and a
282 local non-host chronology. The host-tree series were standardized using 128-
283 year splines with a 50% frequency response, while the non-host ring-width data
284 were standardized using 150-year splines with a 50% frequency response and
285 then averaged via Tukey's biweight robust mean procedure. Host trees from
286 Demijohn Peak (DMJ; 2902 m asl) in the San Juan Mountains of southern
287 Colorado include Douglas-fir (*Pseudotsuga menziesii*) compared against a local

288 non-host ponderosa pine (*Pinus ponderosa*) chronology (Ryerson et al., 2003).
289 The East Fork site (EFK; 2580 m asl) in the Jemez Mountains of northcentral
290 New Mexico includes Douglas-fir and white fir (*Abies concolor*) host trees and
291 a ponderosa pine non-host chronology (Swetnam and Lynch, 1993).

292 With `dfoliatR` loaded, the datasets are accessible using the `data()` func-
293 tion. The data object names are prefixed by their site codes. For instance, the
294 `dmj_*` objects come from the DMJ site and include the host-tree series (`dmj_h`),
295 the non-host chronology (`dmj_nh`), the defoliation series (`dmj_defol`), and the
296 outbreak series (`dmj_obr`). The same suite of data are available for EFK using
297 the `efk_*` prefix.

298 In our example scripts below, `##` and `#` denote user comments, per standard
299 R coding, which are colored in brown. Text in blue denote functions; black are
300 loaded objects, and green are quoted variables and links.

301 5.1. Tree-Level Defoliation Events

302 The function `defoliate_trees()` performs the GSI indexing procedure on
303 each host-tree series and then identifies defoliation events.

```
## Load the package
library(dfoliatR)

## Load data
data("dmj_h") # host tree series
data("dmj_nh") # non-host chronology
# View data structure for any data object via
# "View(dmj_h)" or "head(dmj_h)"

## Calculate indices, identify defoliation events
dmj_defol <- defoliate_trees(host_tree = dmj_h,
                             nonhost_chron = dmj_nh,
                             duration_years = 8,
```



```

max_reduction = -1.28,
bridge_events = TRUE,
series_end_event = TRUE,
list_output = FALSE)

```

304 The result is long-format (stacked) data frame with five variables: “year”,
 305 “series”, “gsi”, “ngsi”, and “defol_status.” The “defol_status” column indicates
 306 whether that year has defoliation or not, with a set of factors that include
 307 “nd” for non-defoliation year, “defol” for a defoliation year, “max_defol” for
 308 the year of maximum suppression (that acts as the basis for individual events),
 309 “bridge_defol” to identify years that link subsequent events (only one is present
 310 at DMJ), and “series_end_defol” to identify defoliation at the present-end of
 311 the series.

312 Selecting `list_output = TRUE` in `defoliate_trees()` provides a list-object
 313 of data frames, each with an `rwl` object that combines the host tree and non-
 314 host series and the other columns created by `defoliate_trees()`. This option
 315 is not used by subsequent functions in `dfoliatR`, but researchers can examine
 316 it to check the results of the GSI calculation (Equation 1), such as the non-host
 317 series after scaling to a common variance with a particular host-tree series.

318 The results of running `defoliate_trees()` can be assessed through graphi-
 319 cal and table outputs. The function `get_defol_events()` will provide a list of
 320 every defoliation event for every tree, with the corresponding mean “ngsi” value.
 321 A summary table of the results for each tree is produced by `defol_stats()` (Ta-
 322 ble 1).

```

defol_stats(dmj_defol)
View(dmj_defol)

```

323 The `plot_defol()` function produces a “ggplot” graphics object with line
 324 segments showing the measured sequence of each series and a filled segment for
 325 each identified defoliation event (Figure 1). The defoliation segments are colored
 326 by their relative severity based on their average NGSI value. The default cut-off

Table 1: Summary statistics for individual tree defoliation events.

series	first	last	years	num_events	tot_years	mean_duration
DMJ01	1810	1996	187	4	50	12
DMJ02	1750	1996	247	6	75	12
DMJ03	1830	1996	167	4	42	10
DMJ04	1720	1996	277	8	90	11
DMJ06	1700	1996	297	6	81	14
DMJ07	1710	1996	287	7	96	14
DMJ11	1900	1997	98	2	19	10
DMJ14	1675	1996	322	9	112	12
DMJ15	1730	1996	267	4	64	16
DMJ16	1746	1996	251	7	86	12
DMJ17	1733	1996	264	6	85	14
DMJ22	1720	1996	277	5	56	11
DMJ23	1675	1997	323	8	96	12
DMJ24	1895	1996	102	3	28	9
DMJ25	1680	1996	317	5	70	14
DMJ26	1700	1996	297	6	79	13
DMJ27	1710	1996	287	3	34	11

values between “Severe” and “Moderate” is the overall mean across all events.
The cut-off between “Moderate” and “Minor” is the first quartile by default.
Users can re-define the breaks to suit their needs via the “breaks” parameter in
`plot_defol()`.

```
plot_defol(dmj_defol)
## This creates a ggplot object. Additions can
## be made to adjust plotting themes and
```

```

## aesthetics, like color.
## For more on the features of ggplot see
## https://ggplot2.tidyverse.org/
## Adding 'scale_color_manual()' will
## change the colors that denote
## the average NGSI value for each defoliation
## event.
library(ggplot2)
plot_defol(dmj_defol) +
  scale_color_manual(values = c("red", "orange", "purple"))

```

331 These output functions aid in assessing the sensitivity of input parameters
 332 to `defoliate_trees()`, including the duration and magnitude thresholds for
 333 identifying defoliation events. Using `plot_defol()` also provides a direct as-
 334 sessment of the between-tree variability in defoliation.

335 5.2. Site-Level Events

336 To infer outbreak events at the site level, the function `outbreak()` compos-
 337 ites tree-level defoliation series into a single chronology, with input parameters
 338 that control thresholds in the number and proportions of trees recording a de-
 339 foliation event.

```

## Use the defol object "dmj_defol" created above
dmj_obr <- outbreak(dmj_defol,
  filter_min_series = 3,
  filter_min_defol = 1,
  filter_perc = 25)

```

340 Input parameters to `outbreak()` include “`filter_min_series`” to control the
 341 chronology cut-off points with regard to sample depth, “`filter_min_defol`” and
 342 “`filter_perc`” to control the minimum number and percent of trees recording a
 343 defoliation event in a given year. `Outbreak()` produces a new data frame with

344 eight variables: “year”, “num_defol”, “percent_defol”, “num_max_defol”, “mean_gsi”,
345 “mean_ngsi”, and “outbreak_status.” All of these variables are populated
346 regardless of an inferred outbreak event, providing a continuous disturbance
347 chronology. The “num_max_defol” variable counts the number of trees record-
348 ing their maximum defoliation in a given year. The “mean_gsi” and “mean_ngsi”
349 variables provide averages of these indices across all available trees. Finally, the
350 “outbreak_status” column shows whether an outbreak event is inferred (“out-
351 break”) or not (“not_obr”).

352 The default plotting function to visualize results from `outbreak()` creates a
353 three-panel graph showing the mean site-level chronology, the percent of trees
354 recording a defoliation, and the sample depth over time (Figure 2).

```
plot_outbreak(dmj_obr, disp_index = "mean_gsi")
```

355 Inferred outbreak events are shown in the top panel of Figure 2 as the filled-
356 in spaces. Users can change the time series in this panel with the “disp_index”
357 parameter, choosing between the mean NGSI (the default) or GSI.

358 A summary table of the inferred outbreak events is generated by the `outbreak_stats()`
359 function (Table 2).

```
dmj_obr_stats <- outbreak_stats(dmj_obr)  
View(dmj_obr_stats)
```

Table 2: Summary statistics for inferred outbreak events.

start	end	duration	num_trees_start	perc_trees_start	peak_outbreak_year	num_trees_outbreak	peak_defol_year	min_gsi	mean_ngsi
1680	1699	20	1	33.3	1690	3	1692	0.121	-2.407
1753	1769	17	6	46.2	1754	7	1755	0.343	-1.611
1825	1840	16	11	78.6	1831	12	1826	0.500	-1.304
1849	1865	17	7	46.7	1852	13	1853	0.252	-1.994
1881	1895	15	8	53.3	1886	14	1885	0.262	-1.945
1959	1970	12	7	41.2	1960	15	1965	0.328	-1.830
1987	1996	10	9	52.9	1989	15	1995	0.378	-1.640

360 The output table from `outbreak_stats()` shows the start and end years of
361 each event, along with their corresponding duration, the number and percent of
362 trees in defoliation at the start of the event, the the number of trees recording
363 the outbreak event (in part or full), the minimums of the mean GSI and NGSI
364 indices, and two points of “peak defoliation.” The first, “`peak_outbreak_year`”
365 is the year in which the greatest number of trees are recording the outbreak,
366 and the second, “`peak_defol_year`” is the year with the lowest average NGSI.
367 These output variables provide a array of options for assessing metrics of the
368 insect defoliation regime, including interval analyses:

```
## calculate mean return interval for outbreak start years  
intervals <- diff(dmj_obr_stats$start)  
mean(intervals) ## 51.2 year mean return interval
```

369 6. Conclusions

370 The `dfoliatR` package provides dendroecologists with tools to infer, quan-
371 tify, analyze, and visualize tree-ring growth suppression events and to recon-
372 struct forest insect defoliator outbreak regimes. It is built on the long-accepted
373 host to nonhost comparison methodology used in the 1980s FORTRAN program
374 OUTBREAK (Swetnam et al., 1985; Swetnam and Lynch, 1989, 1993). Our
375 evaluation of the two programs revealed that `dfoliatR` excelled in identifying
376 defoliation events on single trees, providing researchers with more consistent and
377 biologically-justifiable results. Other key benefits are that `dfoliatR` provides
378 easier control of suppression thresholds, additional output tables, high-quality
379 and customizable graphics, and operates in the open-source R environment that
380 is stable across computing platforms and under active development and main-
381 tenance by a large and growing community.

382 Using `dfoliatR` requires standardized ring-width measurements from insect
383 host trees and either an indexed tree-ring chronology from local non-host trees
384 or suitable climate chronology. It performs an indexing procedure to remove

the climatic signal represented in the non-host chronology from the host-tree series. It then infers defoliation events in individual trees. Site-level analyses identify outbreak events that synchronously affect a user-defined number or proportion of the host trees. Functions are provided for summary statistics and graphics of tree- and site-level series. The package produces publication-quality plots, and tabulates growth suppression indices and tree- and site-level outbreak event statistics for user-defined post-processing needs, including those suitable for charting and tabulating landscape- and regional-level results.

dfoliatR adds a new option for dendroentimology to combine, or “bridge,” sequential tree-level defoliation events into single events. In practice, we suggest that researchers carefully evaluate if bridging is ecologically applicable to study situations and insect ecology, and to carefully explore the data before deciding whether or not to use this option. It is probably not appropriate for insects with high-frequency, high-severity outbreaks, such as processionary caterpillars. Outbreak reconstructions of insects for which the interval is notably longer than typical outbreak duration, such as Douglas-fir tussock moth (*Orgyia pseudotsugata* (Lepidoptera: Erebidae)), are unlikely to be significantly affected. Species for which outbreak duration or individual tree resilience to defoliation varies considerably, or which may chronically infest trees or sites, or alternate between chronic, outbreak, and minimal activity states such as conifer-feeding *Choristoneura*, present more complicated challenges, and the researcher should use discretion with the bridge option.

dfoliatR adds to the on-going open-source software development for dendrochronological methods (e.g., Bunn, 2008; Brewer, 2014; Brewer and Guiterman, 2016). The R environment enables automation of analyses, allowing input/output processes to become routine, and empowering batch processing of large multi-site projects and sensitivity analyses. It also facilitates additional statistical analyses, such as spectral analyses and superposed epoch analyses (e.g., Malevich et al., 2018), with easy transfer from **dfoliatR** and **dplR** to other libraries in R. Source code for **dfoliatR** is available in the Comprehensive R Archive Network (CRAN) and GitHub <https://github.com/chguiterman/>

dfoliatR with updated descriptions and helpful vignettes on the package website <https://chguiterman.github.io/dfoliatR/>. Researchers wishing to contribute to the further development of **dfoliatR** are encouraged to do so via the GitHub repository.

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