

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 caused by insect defoliation. The ‘dfoliatR’ library is based on the FORTRAN V program OUTBREAK, and builds on existing resources in the R computing environment. ‘dfoliatR’ expands on OUTBREAK to provide greater control of suppression thresholds, additional output tables, and high-quality graphics. To use ‘dfoliatR’ requires standardized ring-width measurements from insect host trees and an indexed tree-ring chronology from local non-host trees. 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 available for summary statistics and graphics of tree- and site-level series.

Key words: Dendroecology, spruce budworm, Choristoneura, pandora moth, Coloradia pandora Blake, larch-bud-moth

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27 1. Introduction

28 Variation in the width and morphology of annual radial growth rings in
29 trees permits dating and quantification of past forest insect defoliator outbreaks.
30 Defoliation can be distinguished from climate- and other disturbance-related
31 influences by comparing ring-width or other annually-resolved features in the
32 wood of host species to non-host species or to climate records. The effect of
33 defoliation on radial growth of trees has been recognized since the mid-1800s,
34 but it was not until the 1980s that dendrochronology techniques were applied
35 to identify and quantify defoliation events (Lynch, 2012; Speer, 2010; Swetnam
36 et al., 1985). The first studies (Swetnam and Lynch, 1993, 1989; Swetnam
37 et al., 1985) focused on developing historical reconstructions of western spruce
38 budworm (WSBW; *Choristoneura freemani*). The methodology has since been
39 successfully applied to a wide range of defoliator species, most of which are
40 conifer herbivores, and has evolved in sophistication and application to a wide
41 range of ecosystem situations (Lynch, 2012).

42 The main dendrochronological tool for inferring, dating, and quantifying
43 defoliator outbreaks from tree-ring records has been the software routine OUT-
44 BREAK (Holmes and Swetnam, 1986; Swetnam and Lynch, 1989; Swetnam et
45 al., 1985). OUTBREAK computes indices (described later in detail) of sup-
46 pressed growth by subtracting a detrended and standardized climate series (a
47 “control” chronology) from host individual-tree detrended and standardized ra-
48 dial growth series after the host and non-host series have been brought to a
49 common variance. The non-host chronology often consisted of a site chronology
50 developed from non-host tree species, but a gridded climate data point series,
51 like the North American Drought Atlas (Cook and Krusic, 2004) also suffices.
52 If the host and non-host species respond similarly to climate (which can be
53 tested), the derived series retains variability that the host and non-host series
54 do not have in common, generally the insect signal, along with some unexplained
55 variability (noise). The user defines a rule base specifying the magnitude and
56 duration that a period of indexed growth suppression must meet or surpass for

57 a period of suppressed growth to be inferred as a defoliation event at the tree
58 level.

59 Though powerful, OUTBREAK is outdated and increasingly difficult to use
60 in modern computing environments. It was written in FORTRAN V with in-
61 herently severe restrictions, as RAM and disk space were limited at that time
62 (256 kb and 10 MB, respectively) and FORTRAN conventions imposed very
63 strict formatting, file naming, and output conventions. The program provides
64 no graphical interface or capabilities, forcing users to import generated text files
65 into spreadsheet or other software for assessing results and performing analy-
66 ses. Furthermore, OUTBREAK can only handle one test at a time, creating
67 barriers to batch operation and a large burden for researchers with a datasets
68 including multiple sites. We developed the software library **dfoliatR** as an R-
69 and **dplR**-based routine to overcome these issues.

70 **dfoliatR** adds to a growing suite of dendrochronology packages the R com-
71 puting environment (R Core Team, 2019). Stemming from the **dplR** library
72 (Bunn, 2008) that enables R users to read and write an array of tree-ring data
73 formats, standardize ring width series, build and evaluate chronologies, and
74 perform quality control (to name a few), one can now also measure ring widths
75 from scanned images of prepared samples (Lara et al., 2015; Shi and Xiang,
76 2019), perform and check crossdating (Bunn, 2010), and perform many analyti-
77 cal tests (Jevšenak and Levanič, 2018; Zang and Biondi, 2015). Tools for assess-
78 ing stand dynamics and disturbance analyses are under rapid development, with
79 new packages for assessing growth and release events (**TRADER**: Altman et al.,
80 2014), metrics of growth resilience (**pointRes**: Maaten-Theunissen et al., 2015),
81 and fire history (**burnr**: Malevich et al., 2018). The key objective of **dfoliatR**
82 is to provide tools to identify and analyze insect defoliation and outbreak events
83 by building on the methods employed by OUTBREAK. It capitalizes on the ro-
84 bust software already available in R by using **dplR** data formats for incoming
85 tree-ring series and providing output data formats embodied by the **tidyverse**
86 (Wickham et al., 2019) that include efficient data manipulation (Wickham et
87 al., 2020) and graphics (Wickham, 2016).

88 In this paper, we describe the statistical methods employed by `dfoliatR`,
 89 compare results to those produced by OUTBREAK, and present an example
 90 analysis. Users need not have much experience in R to replicate the analyses
 91 and graphics as presented. The R code below is executable in an R session
 92 once the required libraries are installed and loaded. Support documentation in
 93 addition to this paper is provided within the package via standard help menus
 94 (accessed by typing `?` before a function name) and on the package website
 95 (<https://chguiterman.github.io/dfoliatR/>), which includes up-to-date vignettes
 96 that describe various software routines. Code to generate a preprint of this
 97 manuscript including the R scripts, and tubular and graphical output is available
 98 from https://github.com/chguiterman/dfoliatR_paper.

99 2. Overview of the software

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

- 102 • Standardized ring-width series for individual trees of the host species
- 103 • Standardized tree-ring chronology from a local non-host species

104 Users can develop these data sets in software of their choosing, such as
 105 `dplR` (Bunn, 2008) or ARSTAN (Cook and Holmes, 1996). It is important
 106 that the host-tree data include only one tree-ring series per tree. Both `dplR`
 107 and ARSTAN have options for averaging multiple sample series into a tree-level
 108 series.

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

112 2.1. Identifying Defoliation of Trees

The `defoliate_trees()` function is the point of entry to the `dfoliatR` li-
 brary. 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. `dfoiatR` removes that 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

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

113 where H and NH are the host-tree series and the non-host chronology, in year
 114 i , respectively. Only the common period between the host-tree series and the
 115 non-host chronology are used in Equation 1. The host and non-host chronolo-
 116 gies are brought to common variance by scaling the non-host chronology by its
 117 mean ($\overline{\text{NH}} \approx 1.0$) and multiplying by the ratio of host and non-host standard
 118 deviations ($\frac{\sigma_H}{\sigma_{\text{NH}}}$), which approximates the variance of the host tree series.

119 Negative departures in the normalized GSI (NGSI) that surpass user-defined
 120 thresholds in duration and magnitude are defined as *defoliation events*. As
 121 in OUTBREAK, magnitude is assessed on a single year within the departure
 122 sequence. The default setting is -1.28 (in units of standard deviation), which
 123 was previously determined to be representative of WSBW effects (Swetnam and
 124 Lynch, 1989). Duration is assessed by examining sequences of negative NGSI
 125 before and after the year of maximum departure. Each defoliation event is
 126 allowed one positive excursion on each side of the maximum departure year.
 127 Duration is assessed across the entire sequence that includes up to two positive
 128 excursions. The default duration is eight years, as is commonly used in WSBW
 129 studies (Swetnam and Lynch, 1989). Different species of defoliation insects
 130 vary in the length of defoliation and the degree to which they can suppress
 131 tree growth. Researchers can, and should, adjust the duration and magnitude
 132 parameters accordingly and critically evaluate the results.

133 Diverging from OUTBREAK, `dfoiatR` allows users to extend defoliation
 134 events by bridging successive events and also by allowing potentially short-

duration events that occur at the end of the series. In cases where two defoliation events are separated by a single year, bridging will link them into a single event (**Show figure?**). We urge careful use of this option because there is no setting to limit the number or length of potentially bridged events. The series end option can be used in cases when the host trees were actively being defoliated at the time of sampling. This option eliminates the duration parameter for an event at the recent end of the series, but all other thresholds apply. The advantage of this parameter is that it can aid in identifying the start-year for the current defoliation event or outbreak, which is both useful in management and allows the current event to be included in return-interval estimates.

2.2. Inferring Outbreak Events

Defoliation of one or a few trees does not constitute an outbreak. To determine when defoliation becomes an *outbreak event*, `dfoliatR` composites the individual tree defoliation series into a site-level chronology with the `outbreak()` function. Users have options to define the number and/or the proportion of trees required for an event to be considered an outbreak. Three parameters control whether a defoliation event constitutes an outbreak: the minimum number of trees available, the minimum number of trees recording defoliation, and the percent of trees recording defoliation. The first allows the researcher to make a judgement call as to the confidence ascribed to reduced sample depth toward the ends of their chronologies, thus compensating for the “fading record problem” (Swetnam et al., 1999). The second two parameters adjust the scale of defoliation considered to be an outbreak. Absolute numbers of trees and percentages can be applied separately or in conjunction, following filtering conventions in tree ring fire history studies (Malevich et al., 2018). We urge users to carefully consider the choice of absolute numbers in situations where the number of trees represented in the series varies with time, or the choice of percentages when sample size is small.

163 3. Evaluation

164 3.1. Approach

165 We tested `dfoliatR` against OUTBREAK by comparing NGSI to corrected
166 indices for individual trees and years, defoliation status for individual trees
167 and years, and percentage of trees recording outbreaks at the site level, using
168 standardized ring-width data from eight sites in British Columbia, Colorado,
169 and New Mexico, and author-provided non-host site chronologies.

170 We detrended host data for both `dfoliatR` and OUTBREAK using AR-
171 STAN v6.1 (Cook and Holmes, 1996) with default double detrending (128 year
172 wavelength and a 50% smoothing spline). In both `dfoliatR` and OUTBREAK
173 we used event thresholds of -1.28 normalized index, 8 years duration, and al-
174 lowed for events at the end of series in seven of eight sites. These seven sites
175 were sampled during ongoing outbreak events (Swetnam and Lynch, 1993). We
176 found it necessary to be consistent in how we detrended and in what software
177 (e.g., ARSTAN vs `dplR`) we employed because subtle differences in standard-
178 ized ring-width indices created between the programs exacerbated differences
179 between `dfoliatR` and OUTBREAK. In the end, we chose to only use the stan-
180 dardization output files from ARSTAN, which are easily read into `R` (and then
181 `dfoliatR`) using the `dplR` package.

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

184 3.2. Findings

185 Since both programs employ Equation 1, `dfoliatR` and OUTBREAK com-
186 pute identical growth suppression indices at 0.000 precision (Figure 3). In all
187 there are 43,280 ring-width indices from 222 trees included in our test. Of
188 these, the programs identified 11,530 years of defoliation. The programs agreed
189 on 92% of the years, leaving 927 “difference” years in which only one program
190 identified defoliation on an individual tree. The differences included 102 events
191 on 85 trees. We carefully inspected each of these events in the full context of
192 each tree’s ring series, and categorized the differences as follows

- 193 • *Series-end events* (40% of the total) in which we allowed OUTBREAK to

194 include “truncated outbreaks” (for seven sites) at the end of each series.

195 In `dfoliatR`, this option is controlled by the “series_end_events” param-

196 eter to `defoliate_trees()`. In OUTBREAK, the option appears while

197 changing the duration parameter (option 3). When selected, OUTBREAK

198 will include any sequences of negative indices at the either the beginning

199 or the end of each tree series. No thresholds are applied in identifying

200 these as defoliation events. In `dfoliatR`, the negative sequences are eval-

201 uated as with any other possible defoliation, but the duration threshold is

202 omitted. Each of the 13 events included in these differences did not meet

203 the “max_reduction” parameter (-1.28 NGSI) in `dfoliatR` and were ex-

204 cluded. In two cases, OUTBREAK included events at the beginning of

205 the series where `dfoliatR` does not allow truncated events. In four cases,

206 OUTBREAK omitted the last year of the series because it was positive.

207 In two cases, `dfoliatR` omitted possible events because it had already

208 included a positive NGSI excursion after the “max_reduction” year, and

209 since it will only allow one excursion on either side of the max year, the

210 events were omitted due to short duration.
- 211 • *Sequential events* (36%) in which OUTBREAK omitted events that occur

212 one year prior or one year following an identified event. OUTBREAK

213 selects the event with the greatest negative index year. On two trees,

214 OUTBREAK omitted two of three sequential events. `dfoliatR` allows all

215 such events. While inspecting these differences, we added an option to

216 `defoliate_trees()` that would “bridge” between sequential events, mak-

217 ing them into single, long events. We felt that this was ecologically justi-

218 fied, especially for studies of WSBW, because the insect will occasionally

219 maintain minor defoliation f individuals for decades even while not in an

220 outbreak phase.
- 221 • *Undetermined differences* (22%) occurred in cases where OUTBREAK

222 omitted events without clear cause.

223 • *Rounding* (2%) differences in the indices either omitted or cut short events
224 on two trees. In both cases the indices were very close to zero.

225 Our inspections revealed what we believe are short-comings in how OUT-
226 BREAK identifies defoliation events on individual trees. In every one of the
227 102 cases we inspected, we felt that `dfoliatR` provided a more biologically and
228 statistically appropriate assessment of defoliation.

229 and use them to infer identical tree-level outbreak events, including onset and
230 termination dates and date and magnitude of the maximum growth suppression
231 when `dfoliatR` bridging matches the OUTBREAK protocols. Once the issues
232 with ARSTAN (described earlier) were resolved, `dfoliatR` and OUTBREAK
233 produced identical indices at 0.000 precision.

234 When inferring outbreak events at the tree level, our initial results did not
235 always match OUTBREAK, leading to development of the bridging option.
236 Identifying a minimum within a period while allowing for positive excursions is
237 a tricky computation. OUTBREAK deliberately disallows back-to-back events
238 – after a minimum value within a period of negative values, a second positive
239 index terminates an event. OUTBREAK does not allow a period of highly
240 negative values to be appended to an earlier outbreak if the intervening period is
241 1 year, even if those negative values fall below the minimum in the earlier period.
242 Due in large part to reconstructions using OUTBREAK (see papers cited by
243 Lynch, 2012), we now know considerably more about forest defoliator outbreak
244 regimes than we did in the 1980s when OUTBREAK was written. We think that
245 two prolonged events separated by a single year should in some situations be
246 considered a single outbreak at both the tree and site levels. This is particularly
247 relevant to western spruce budworm and spruce budworm (*C. fumiferana*), for
248 which regimes have been reconstructed for several geographic areas and for
249 which researchers and forest health experts have gained considerable knowledge
250 (**Sanders et al. 1985, Brookes et al. 1987 and many later publications**).
251 We now know that the greatest growth suppression often occurs late in the
252 outbreak due to cumulative effects (**rrr**). We developed the bridging option to

253 permit these periods to be linked to preceding events separated by a single year.

254 Bridging operates at the tree level in `dfoliatR`, and these effects generally
255 have minor effects on site-level reconstructions (**Fig.Bridging**). Occasionally
256 site-level events are inferred differently (**Table.Bridging**), and this can affect
257 return interval and duration statistics.

258 3.3. Best Practice Recommendations

259 We encourage authors to report software and library version numbers and
260 download dates where applicable. This is standard practice for statistical pack-
261 ages, but has often been ignored for R- and dplR-based analyses. R-based soft-
262 ware evolves over time, and results cannot always be replicated if the software
263 versions are unknown. Similarly, significant changes (especially fixes) imply
264 that re-analysis of data might differ from earlier published results and inter-
265 pretations. Wherever possible, refereed publications supporting the software
266 should be cited. Our experience with ARSTAN demonstrates the importance of
267 adopting this practice. In the United States, analytic methods not documented
268 thusly and supported by the literature might be questioned if the results are
269 used in a National Environmental Protection Act (NEPA) process.

270 We recommend that researchers compare `dfoliatR` results with and without
271 bridging, and see which option agrees best with known insect biology and ecol-
272 ogy. Bridging may be a logical choice for insects with prolonged non-outbreak
273 periods, such as western spruce budworm, but not for situations where impacted
274 stands barely recover from one outbreak before another begins, such as occurs
275 with pine processionary caterpillars (*Thaumetopoea pityocampa* (Lepidoptera:
276 Thaumetopoeidae)) (Carus 2004, 2009).

277 4. Availability and installation

278 The `dfoliatR` library (Guiterman et al., 2020) is provided free and open
279 source from the Comprehensive R Archive Network (CRAN; [https://cran.r-](https://cran.r-project.org/)
280 [project.org/](https://cran.r-project.org/)). To install `dfoliatR` from CRAN use

```
install.packages("dfoIiatR")
```

281 In each R session, `dfoIiatR` can be loaded via

```
library(dfoIiatR)
```

282 Development versions of `dfoIiatR` are available on GitHub and installed
283 using the `devtools` library,

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

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

289 5. Example Usage

290 In `dfoIiatR` we provide two sets of tree-ring data to aid users in exploring the
291 functions, graphics, and outputs. Each set consists of individual host-tree series,
292 standardized using 50-year splines with a 50% frequency response, and a local
293 non-host chronology. The non-host ring-width data were standardized using
294 150-year splines with a 50% frequency response. The non-host chronologies
295 were averaged following standard procedures (Speer, 2010). Host trees from
296 Demijohn Peak (DMJ; 2902 m asl) in the San Juan Mountains of southern
297 Colorado include Douglas-fir (*Pseudotsuga menziesii*) compared against a local
298 non-host ponderosa pine (*Pinus ponderosa*) chronology (Ryerson et al., 2003).
299 The East Fork site (EF; 2580 m asl) in the Jemez Mountains of northcentral
300 New Mexico include Douglas-fir and white fir (*Abies concolor*) host trees and
301 a ponderosa pine non-host chronology (Swetnam and Lynch, 1993).

302 5.1. Tree-Level Defoliation Events

303 To assess defoliation of trees, we begin with a set of standardized host-tree
304 series and a local non-host chronology or climate reconstruction. First, the
305 `dfoliatR` library needs to be loaded (after installation as above) and the in-
306 cluded datasets can be loaded. The function `defoliate_trees()` performs the
307 GSI indexing procedure on each host-tree series and then identifies defoliation
308 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)
```

309 The result of running `defoliate_trees()` is long-format (stacked) data
310 frame with five variables: “year”, “series”, “gsi”, “ngsi”, and “defol_status.”
311 The “gsi” variable is the “corrected” tree-level growth suppression index. It was
312 normalized or scaled into z-scores to produce “ngsi”, which is used to identify de-
313 foliation events. The “defol_status” column provides a set of character strings,

314 “nd” for non-defoliation year, “defol” for a defoliation year, “max_defol” for
315 the year of maximum suppression (that acts as the basis for individual events),
316 “bridge_defol” to identify years that link subsequent events (only one is present
317 at DMJ), and “series_end_defol” to identify defoliation at the present-end of
318 the series. Selecting `list_output = TRUE` provides a list-object of data frames,
319 each with the `rw1` object that combines the host tree and non-host chronology
320 and the columns created in `defoliate_trees()`. This option is not used by
321 subsequent functions in `dfoliatR`, but researchers can examine it to check the
322 results of the GSI calculation (Equation 1).

323 We can assess the results of running `defoliate_trees()` through graphical
324 and table outputs. The function `get_defol_events()` will provide a list of
325 every defoliation event for every tree, with the corresponding mean “ngsi” value.
326 A summary table of the results for each tree is produced by `defol_stats()`
327 (Table 1).

```
defol_stats(dmj_defol)
View(dmj_defol)
```

328 The `plot_defol()` function produces a “ggplot” graphics object with line
329 segments showing the measured sequence of each series and a filled segment for
330 each identified defoliation event (Figure 1). The defoliation segments are colored
331 by their relative severity based on their average NGSi value. The default cut-off
332 values between “Severe” and “Moderate” is the overall mean across all events
333 and between “Moderate” and “Minor” is the first quartile. Users can define
334 these breaks to suit their needs.

```
plot_defol(dmj_defol)
## This creates a ggplot object, so additions can
## be made to adjust plotting themes and
## aesthetics, like color.
## For more on the features of ggplot:
## https://ggplot2.tidyverse.org/
```

```
library(ggplot2)
plot_defol(dmj_defol) +
  scale_color_manual(values = c("red", "orange", "purple"))
```

335 These basic output functions aid in assessing the sensitivity of input param-
 336 eters to `defoliate_trees()` including the duration and magnitude thresholds
 337 for identifying defoliation events. Using `plot_defol()` also provides a direct
 338 assessment of the between-tree variability in deflation. For further analyses, we
 339 composite the host-tree defoliation series to the site level using `outbreak()`.

340 5.2. Site-Level Events

341 Once we have identified defoliation events on individual trees, we can infer
 342 outbreak events at the site level. The function `outbreak()` acts to composite
 343 tree-level defoliation series into a single chronology, with input parameters that
 344 control thresholds in the number and proportions of trees recording a defoliation
 345 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)
```

346 Input parameters to `outbreak()` include “`filter_min_series`” to control the
 347 chronology cut-off points with regard to sample depth, “`filter_min_defol`” and
 348 “`filter_perc`” to control the minimum number and percent of trees, respec-
 349 tively, recording a defoliation event in a given year. `Outbreak()` produces a
 350 new data frame with eight variables: “`year`”, “`num_defol`”, “`percent_defol`”,
 351 “`num_max_defol`”, “`mean_gsi`”, “`mean_ngsi`”, and “`outbreak_status`.” All of
 352 these variables are populated regardless of an inferred outbreak event, providing
 353 a continuous disturbance chronology. The “`num_max_defol`” variable counts
 354 the number of trees recording their maximum defoliation in a given year. The

355 “mean_gsi” and “mean_ngsi” variables provide averages of these indices across
356 all available trees. Finally, the “outbreak_status” column shows whether an
357 outbreak event is inferred (“outbreak”) or not (“not_obr”).

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

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

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

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

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

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

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

375 6. Conclusions

- 376 • Encourage use of `dfoliatR` over OUTBREAK – more accurate, greater
377 user control, more easily conduct sensitivity tests to evaluate parameters,
378 data, and ecological patterns
- 379 • Facilitates high-level statistical analyses by easily transferring to other
380 libraries (MASS, etc.)
- 381 • A community resource that can be easily maintained and updated into
382 the future.

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390 Forest Service Rocky Mountain Research Station. Any use of trade, firm, or
391 product names is for descriptive purposes only and does not imply endorsement
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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

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

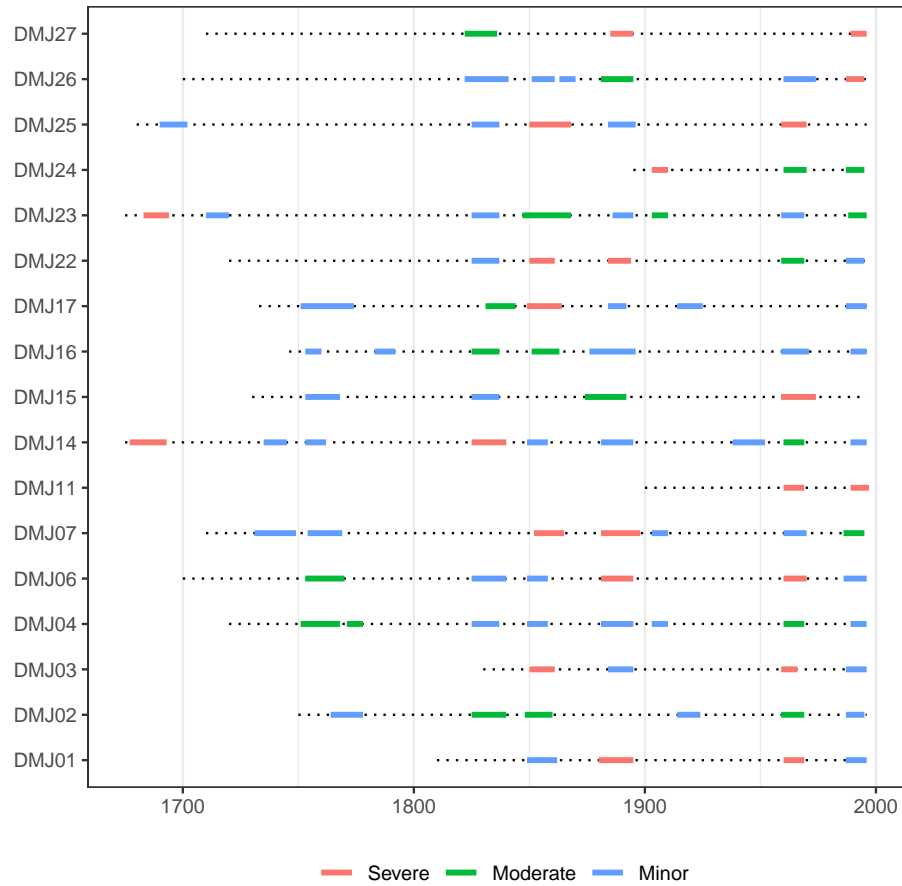


Figure 1: Default dfoliatR graphics for individual trees, produced by the 'plot_defol()' function. Dotted lines represent the series length for each tree and colored segments show periods of defoliation. Users can define the Cut-off values to determine severe-moderate-minor defoliation intensities, and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

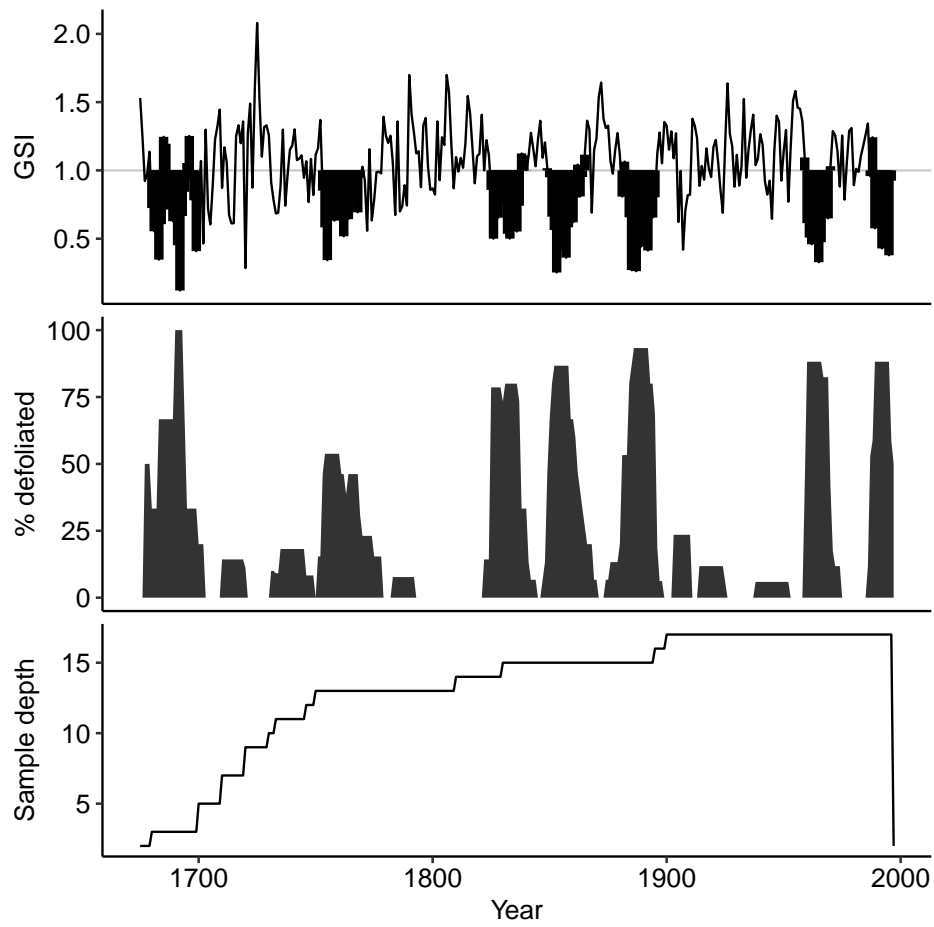


Figure 2: Default graphic to show outbreak events at the DMJ site. Top panel shows the mean GSI through time, with inferred outbreak events filled. The middle panel shows the percent of trees defoliated, and the bottom panel shows the number of trees through time.

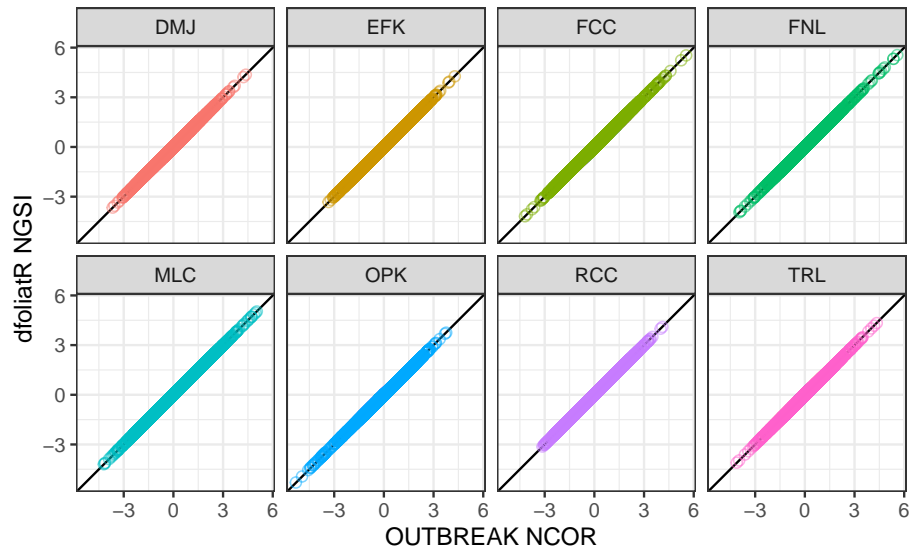


Figure 3: Comparison of disturbance indices between `dfoliatR` and `OUTBREAK`. Diagonal 1:1 lines show equal values.