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

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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.

# Introduction

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

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

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

dfoliatR adds to a growing suite of dendrochronology packages the R computing environment (R Core Team, 2019). Stemming from the dplR library (Bunn, 2008) that enables R users to read and write an array of tree-ring data formats, standardize ring width series, build and evaluate chronologies, and perform quality control (to name a few), one can now also measure ring widths from scanned images of prepared samples (Lara et al., 2015; Shi and Xiang, 2019), perform and check crossdating (Bunn, 2010), and perform many analytical tests (Jevšenak and Levanič, 2018; Zang and Biondi, 2015). Tools for assessing stand dynamics and disturbance analyses are under rapid development, with new packages for assessing growth and release events (TRADER: Altman et al., 2014), metrics of growth resilience (pointRes: Maaten-Theunissen et 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 (Wickham et al., 2020) and graphics (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. 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 (accessed by typing ? before a function name) and on the package website (<https://chguiterman.github.io/dfoliatR/>), which includes up-to-date vignettes that describe various software routines. Code to generate a preprint of this manuscript including the R scripts, and tubular and graphical output is available from <https://github.com/chguiterman/dfoliatR_paper>.

# 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
* Standardized tree-ring chronology from a local non-host species

Users can develop these data sets in software of their choosing, such as dplR (Bunn, 2008) 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 and then composite across multiple trees to identify outbreak events.

## 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. 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

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 () and multiplying by the ratio of host and non-host standard deviations (), which approximates the variance of the host tree series.

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

Diverging from OUTBREAK, dfoliatR allows users to extend defoliation 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.

## 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.

# Evaluation

## Approach

We tested dfoliatR against OUTBREAK by comparing NGSI to corrected indices for individual trees and years, outbreak status for individual trees and years, and percentage of trees recording outbreaks at the site level, using raw ring-width data from 8 sites in British Columbia, Colorado, and New Mexico, and author-provided non-host site chronologies.

We detrended host data for both dfoliatR and OUTBREAK using ARSTAN 6.1 (**downloaded 21 April 2002 from dpl; xxx Ann check the date, as 21 April 2002 is 6.05P**) with default double detrending (128 year wavelength and a 50% smoothing spline) and event thresholds of -1.28 normalized index and 8 years. Attempts to detrend with ARSTAN 44 in dplR were problematic for two reasons. Some versions of ARSTAN, including ARSTAN 44, do not include an option for producing tree-level averages from multiple cores from the same tree (.tre files). **More importantly, ring-width indices from ARSTAN 44 truncated the process at either end when sample size fell below 5 xxx (Jodi, what was the minimum?)**, and produced different values for years near the truncation point than did dpl-ARSTAN 6.05P. The majority of the detrended series values were identical, but not the tails, as the two routines do not use the same data to establish the tails. These differences produce different GSI (dfoliatR) and Corrected Indices (OUTBREAK) values at the tails, sometimes considerably, affecting whether or not outbreak events are inferred early or late in the tree series, timing of the first or last event onset, and subsequent computation of return intervals, duration, and periodicity. Note that there is nothing particularly important regarding the choice of ARSTAN 6.1 – it is simply the version that the second author used routinely and had available.

## Findings

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

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

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

## Best Practice Recommendations

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

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

# Availability and installation

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

install.packages("dfoliatR")

In each R session, dfoliatR can be loaded via

library(dfoliatR)

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

devtools::install\_github("chguiterman/dfoliatR")

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

# Example Usage

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

## Tree-Level Defoliation Events

To assess defoliation of trees, we begin with a set of standardized host-tree series and a local non-host chronology or climate reconstruction. First, the dfoliatR library needs to be loaded (after installation as above) and the included datasets can be loaded. The function defoliate\_trees() performs the GSI indexing procedure on 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)

The result of running defoliate\_trees() is long-format (stacked) data frame with five variables: “year”, “series”, “gsi”, “ngsi”, and “defol\_status.” The “gsi” variable is the “corrected” tree-level growth suppression index. It was normalized or scaled into z-scores to produce “ngsi”, which is used to identify defoliation events. The “defol\_status” column provides a set of character strings, “nd” for non-defoliation year, “defol” for a defoliation year, “max\_defol” for the year of maximum suppression (that acts as the basis for individual events), “bridge\_defol” to identify years that link subsequent events (only one is present at DMJ), and “series\_end\_defol” to identify defoliation at the present-end of the series. Selecting list\_output = TRUE provides a list-object of data frames, each with the rwl object that combines the host tree and non-host chronology and the columns created in defoliate\_trees(). This option is not used by subsequent functions in dfoliatR, but researchers can examine it to check the results of the GSI calculation (Equation 1).

We can assess the results of running defoliate\_trees() through graphical and table outputs. The function get\_defol\_events() will provide a list of every defoliation event for every tree, with the corresponding mean “ngsi” value. A summary table of the results for each tree is produced by defol\_stats() (Table 1).

defol\_stats(dmj\_defol)  
View(dmj\_defol)

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

The plot\_defol() function produces a “ggplot” graphics object with line segments showing the measured sequence of each series and a filled segment for each identified defoliation event (Figure @(fig-defol)). The defoliation segments are colored by their relative severity based on their average NGSI value. The default cut-off values between “Severe” and “Moderate” is the overall mean across all events and between “Moderate” and “Minor” is the first quartile. Users can define 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"))

![Figure 1: Default dfoliatR graphics for individual trees. (A) shows basic color scheme, (B) includes user-defined colors. Dotted lines represent the series length for each tree and colored segments show periods of defoliation.](data:application/pdf;base64,)

Figure 1: Default dfoliatR graphics for individual trees. (A) shows basic color scheme, (B) includes user-defined colors. Dotted lines represent the series length for each tree and colored segments show periods of defoliation.

These basic output functions aid in assessing the sensitivity of input parameters to defoliate\_trees() including the duration and magnitude thresholds for identifying defoliation events. Using plot\_defol() also provides a direct assessment of the between-tree variability in deflation. For further analyses, we composite the host-tree defoliation series to the site level using outbreak().

## Site-Level Events

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

Input parameters to outbreak() include “filter\_min\_series” to control the chronology cut-off points with regard to sample depth, “filter\_min\_defol” and “filter\_perc” to control the minimum number and percent of trees, respectively, recording a defoliation event in a given year. Outbreak() produces a new data frame with eight variables: “year”, “num\_defol”, “percent\_defol”, “num\_max\_defol”, “mean\_gsi”, “mean\_ngsi”, and “outbreak\_status.” All of these variables are populated regardless of an inferred outbreak event, providing a continuous disturbance chronology. The “num\_max\_defol” variable counts the number of trees recording their maximum defoliation in a given year. The “mean\_gsi” and “mean\_ngsi” variables provide averages of these indices across all available trees. Finally, the “outbreak\_status” column shows whether an outbreak event is inferred (“outbreak”) or not (“not\_obr”).

The default plotting function to visualize results form outbreak() creates a three-panel graph showing the mean site-level chronology, the percent of trees recording a defoliation, and the sample depth over time (Figure @(fig-obr)).

plot\_outbreak(dmj\_obr, disp\_index = "mean\_gsi")

![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.](data:application/pdf;base64,)

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.

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

A summary table of the inferred outbreak events is generated by the outbreak\_stats() function (Table @(tbl-obr)).

dmj\_obr\_stats <- outbreak\_stats(dmj\_obr)

The output table from outbreak\_stats() shows the start and end years of each event, along with their corresponding duration, the number and percent of trees in defoliation at the start of the event, the the number of trees recording the outbreak event (in part or full), the minimums of the mean GSI and NGSI indices, and two points of “peak defoliation.” The first, “peak\_outbreak\_year,” is the year in which the greatest number of trees is recording the outbreak, and the second, “peak\_defol\_year,” is the year in with the lowest average NGSI. These output variables provide a array of options for assessing basic metrics of 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

# Extensions

* Describe how dfoliatR can be combined with other R libraries
  + Mapping, what else?

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