

Peters, R. L., Pappas, C., Hurley, A., Poyatos, R., Flo, V., Zweifel, R., Goossens, W., Steppe, K. (2021): Assimilate, process and analyse thermal dissipation sap flow data using the TREXrpackage. - Methods in Ecology and Evolution, 12, 2, 342-350.

<https://doi.org/10.1111/2041-210X.13524>

1 **Assimilate, process, and analyse thermal dissipation sap flow data using the TREX R**
2 **package**

3 Richard L. Peters^{a,b*}, Christoforos Pappas^{c,d,e}, Alexander G. Hurley^{f, g}, Rafael Poyatos^{h,i},
4 Victor Flo^g, Roman Zweifel^b, Willem Goossens^a, Kathy Steppe^a

5 ^aLaboratory of Plant Ecology, Department of Plants and Crops, Faculty of Bioscience
6 Engineering, Ghent University, Coupure links 653, B-9000 Ghent, Belgium

7 ^bForest Dynamics, Swiss Federal Research Institute for Forest, Snow and Landscape Research
8 (WSL), Zürcherstrasse 111, CH-8903 Birmensdorf, Switzerland

9 ^cDépartement de géographie, Université de Montréal, Montréal, H2V 0B3, Montreal, QC,
10 Canada

11 ^dCentre d'étude de la forêt, Université du Québec à Montréal, C.P. 8888, Succursale Centre-
12 ville, Montréal, QC, H3C 3P8, Canada

13 ^eDépartement Science et Technologie, Téliuq, Université du Québec, 5800 rue Saint-Denis,
14 Bureau 1105, Montréal, QC, H2S 3L5, Canada

15 ^gGFZ German Research Centre for Geosciences, Section 4.3 Climate Dynamics and
16 Landscape Evolution, Potsdam, Germany

17 ^fSchool of Geography, Earth and Environmental Sciences, University of Birmingham,
18 Birmingham B15 2TT, United Kingdom

19 ^hCREAF, E08193 Bellaterra (Cerdanyola del Vallès), Catalonia, Spain

20 ⁱUniversitat Autònoma de Barcelona, E08193 Bellaterra (Cerdanyola del Vallès), Catalonia,
21 Spain

22 *Corresponding Author:

23 Tel: +32 9 264 61 12, Fax: +32 9 224 44 10, e-mail: Richard.Peters@UGent.be.

24 **Abstract**

25 1. A key ecophysiological measurement is the flow of water (or sap) along the tree's water-
26 transport system, which is an essential process for maintaining the hydraulic connection within
27 the soil-plant-atmosphere continuum. The thermal dissipation method (TDM) is widespread in
28 the scientific community for measuring sap flow and has provided novel insights into water use
29 and its environmental sensitivity, from the tree- to the forest-stand level. Yet, methodological
30 approaches to determine sap flux density (SFD) from raw TDM measurements remains case-
31 specific, introducing uncertainties and hampering data syntheses and meta-analyses.

32 2. Here, we introduce the R package TREX (TRee sap flow EXtractor), incorporating a wide
33 range of sap flow data-processing procedures to quantify SFD from raw TDM measurements.
34 TREX provides functions for: i) importing and assimilating raw measurements, ii) data quality
35 control and filtering, and iii) calculating standardized SFD outputs and their associated
36 uncertainties according to different data-processing methods.

37 3. A case study using a Norway spruce tree illustrates TREX's functionalities, featuring
38 interactive data curation and generating outputs in a reproducible and transparent way. The
39 calculations of SFD in TREX can, for instance, use the original TDM calibration coefficients,
40 user-supplied calibration parameters or calibration data from a recently compiled database of
41 22 studies and 37 species. Moreover, the package includes an automatic procedure for
42 quantifying the sensitivity and uncertainty of the obtained results to user-defined assumptions
43 and parameter values, by means of a state-of-the-art global sensitivity analysis.

44 4. Time series of plant ecophysiological measurements are becoming increasingly available and
45 enhance our understanding of climate change impacts on tree functioning. TREX allows for
46 establishing a baseline for data processing of TDM measurements and supports comparability
47 between case studies, facilitating robust, transparent, and reproducible large-scale syntheses of
48 sap flow patterns. Moreover, TREX facilitates the simultaneous application of multiple
49 common data-processing approaches from raw data to physiological relevant quantities. This
50 allows for robust quantification of the impact (i.e., sensitivity and uncertainty) of user-specific
51 choices and methodological differences, which is necessary for process understanding and
52 policy making.

53 **Keywords:** thermal dissipation method, transpiration, sap flux density, calibration, uncertainty
54 analysis, global sensitivity analysis, whole-tree water use

55 1 | Introduction

56 A key ecophysiological measurement is tree water transport, as terrestrial plant
57 transpiration plays a key role in the local, regional and global water cycle (e.g., Reyes-Acosta
58 & Lubczynski 2013; Schlesinger & Jasechko 2014; Fatichi & Pappas 2017). More specifically,
59 sap flow (SF; total flow of water, often expressed in kg H₂O per h) attracts large interest across
60 several scientific disciplines, as it is essential for maintaining the soil-plant-atmosphere
61 continuum (e.g., Zweifel, Item & Häslner 2001; Steppe, Vandegehuchte, Tognetti & Mencuccini
62 2015). SF can be measured in different plant organs, although it is typically measured in stems.
63 For decades, heat-based SF measurements have been used to quantify whole-tree water use for
64 partitioning ecosystem-level water fluxes into transpiration and evaporation (e.g., Poyatos *et al.*
65 2016). In addition, species comparisons on stomatal regulation in mature trees have been
66 performed using SF, revealing species-specific environmental responses to drought and heat
67 waves (e.g., Brinkmann, Eugster, Zweifel, Buchmann & Kahmen 2016; Dietrich, Delzon, Hoch
68 & Kahmen 2019).

69 A large variety of heat-based SF methods exist (as reviewed by Smith & Allen 1996;
70 Peters *et al.* 2018), yet no method suits all needs for tree ecophysiological research. The thermal
71 dissipation method (TDM) is the most widely applied for measuring SF in trees, due to its
72 simplicity and low cost (Flo, Martinez-Vilalta, Steppe, Schuldt & Poyatos 2019). In short, the
73 TDM measures how moving sap within the tree's water conducting xylem (i.e. sapwood) affects
74 the dissipation of heat supplied by a continuous heater (Granier 1985). The TDM is applied by
75 installing two axially aligned probes into the xylem and measuring the temperature difference
76 between a continuously heated probe and the non-heated reference probe (ΔT in °C, or as
77 voltage difference, ΔV in mV), where ΔT decreases due to the convective cooling with moving
78 sap over the heating probe.

79 The range of user-specific choices (e.g., parameter values, calculation procedures)
80 applied during data processing between studies represents a key challenge for homogenising
81 and synthesising SF measurements obtained from TDM (Peters *et al.* 2018). A large variety of
82 TDM data-processing steps are applied to convert raw data into physiologically meaningful
83 variables, namely sap flux density (SFD in cm³ of sap per cm² of sapwood area per h), yet often
84 only a single data-processing path is utilized. Typically suggested data-processing steps
85 include: i) determining time periods of zero flow (Rabbal, Diekkrüger, Voigt & Neuwirth
86 2016), ii) correcting for the partial insertion of the probe into non-conducting heartwood
87 (Clearwater, Meinzer, Andrade, Goldstein & Holbrook 1999), iii) accounting for signal

88 dampening due to wound response (Peters *et al.* 2018), and iv) sensor calibration (Fuchs,
89 Leuschner, Link, Coners & Schuldt 2018). Yet, although data-processing steps can impact the
90 quantification of SFD (Köstner, Granier & Cermák 1998; Peters *et al.* 2018), no consensus
91 exists on a standardized method.

92 The lack of a standardized method of TDM data processing has stimulated the recent
93 development of software tools for specific TDM data-processing steps (i.e., Oishi, Hawthorne
94 & Oren 2016; Speckman, Ewers & Beverly 2018). Although such tools provide an array of data
95 cleaning procedures with a graphical interface and a fixed chain of data-processing options,
96 they do not facilitate the exploration of multiple approaches and parameter values (i.e., multi-
97 method ensemble for SFD data analyses). Moreover, although there are techniques to identify
98 parameters which critically impact the output of interest (e.g., De Pauw, Steppe & De Baets
99 2008; Pappas, Fatichi, Leuzinger, Wolf & Burlando 2013), such systematic sensitivity analyses
100 have yet to be considered for TDM data-processing. The consequence is that SFD data from
101 TDM studies may still be difficult to compare or to apply in broader ecological research
102 questions.

103 Here we present the R package TREX (TRee sap flow Extractor; R Core Team 2017)
104 freely accessible via the Comprehensive R Archive Network (CRAN: <http://cran.r-project.org>).
105 TREX imports and assimilates raw TDM measurements, provides a complete multi-modular
106 data-processing workflow and facilitates SFD-related outputs and analyses. TREX provides
107 functionalities for interactive data curation, quantification of the sensitivity and uncertainty of
108 the obtained results to user-defined assumptions and parameter values, and produces outputs in
109 a reproducible and transparent way. Notwithstanding, a multitude of the TREX functionalities
110 are also applicable to other heat-based SF measurements.

111 **2 | TREX functionalities**

112 We demonstrate the functionalities of TREX using an example dataset (Figure 1) from
113 a mature Norway spruce (*Picea abies* (L.) Karst.; see *tdm.data()*). TREX functionalities are
114 structured in three modules (Figure 2), where each module contains multiple functions which
115 can be used separately or in a processing chain (short descriptions of the functions are provided
116 in Table S1). The first module “Import & assimilation” guides users to import data, adjust time-
117 step-related issues, perform outlier detection and gap filling. The second module “Data
118 processing” provides a chain of functions which apply different TDM data-processing
119 procedures. The final module “Output & analyses” generates standardized data output and

120 quantifies the output sensitivity and uncertainty associated to variations in selected parameters.
121 Some functions from the “Import & assimilation” and “Output & analyses” modules can also
122 be used with other SF methods (Figure 2).

123 *TREX functions for data import and assimilation*

124 TREX utilizes TDM time series of ΔT (or ΔV) provided for each individual sensor in
125 text file format, with ΔT (or ΔV) data recorded for a specific timestamp, providing information
126 on time zone and daylight saving time (see *example.data()*). It is crucial to know the time zone
127 in which the data were logged and whether day-time-saving was applied. The input can be
128 presented in a *Timestamp* format: including a timestamp of the measurements column, and
129 value of ΔT (or ΔV) of a specific sensor (Table 1; but also see *DOY* format in
130 *?TREX::example.data*). Multiple reference probes used to correct for natural temperature
131 gradients in the sapwood (Lindén, Fonti & Esper 2016) can also be added as additional columns
132 (labelled as *ref1*, *ref2*, ..., *refn*). The user can verify whether the imported dataset fits the
133 requirements for TREX data format by calling the function *is.trex()*. After specifying the
134 timestamp format and time zone, the users can select whether the time series should be
135 standardized to solar time (i.e., mid-day corresponding to solar noon which enhances
136 comparability across sites) by providing the longitude (in decimal degrees) of the location of
137 the sensor. Although TREX functionalities are applied per individual sensor, a looping
138 procedure could allow for analysing multiple individual trees or sensors.

139 The *dt.steps()* function manipulates the temporal extent and resolution (indicated with
140 *time.int* in minutes) of an *is.trex* object (i.e., to facilitate comparison with other time series).
141 Outliers in ΔT (or ΔV) time series need to be identified and removed before proceeding with
142 data processing (Figure 3). This task is addressed in the TREX package with an accompanying
143 Shiny application that is launched with the *outlier()* function and i) visualizes raw and outlier-
144 free time series interactively (Sievert 2018), ii) highlights automatically detected outliers (see
145 *?TREX::outlier()* for method specifications), iii) allows the user to revise the automatically
146 detected outliers and manually include data points interactively, and iv) exports the original
147 data and the outlier-free time series in a *is.trex()* object that can be further processed. The
148 cleaned or raw time series can additionally be fed to the *gap.fill()* function to interpolate
149 sporadic gaps of a user-defined length (using *max.gaps* defined in minutes).

150 *TREX data-processing functions*

151 SFD is estimated by first determining the relative thermal difference (known as the heat
 152 index and denoted as K , unitless) between ΔT and zero-flow conditions (denoted as ΔT_{\max} or
 153 ΔV_{\max} in equation 1; Granier 1985). Multiple methods exist to determine ΔT_{\max} (equation 1;
 154 Rabbel, Diekkrüger, Voigt & Neuwirth 2016). Four classes of methods (as reviewed by Peters
 155 *et al.* 2018) are incorporated in the function `tdm_dt.max()`, including: i) the pre-dawn ΔT_{\max}
 156 (*pd*), where zero-flow is assumed to occur each night, ii) daily moving window (*mw*; e.g., 7
 157 days, defined with *max.days*), in which the centred maximum pre-dawn ΔT_{\max} over a moving
 158 window is assumed to represent zero-flow conditions, iii) the double regression method (*dr*;
 159 Lu, Urban & Zhao 2004), determining the mean pre-dawn ΔT_{\max} over the *max.days* period and
 160 removes all points below this moving-window mean and calculates a second running mean
 161 moving window, and iv) the environmental dependent method (*ed*) assuming that one can find
 162 zero-flow conditions based on environmental conditions (Oishi, Hawthorne & Oren 2016; see
 163 `?TREX::tdm_dt.max` for details). As the selected parameters are likely site-, and even tree-
 164 specific, users should carefully inspect the selected nights of zero-flow conditions resulting
 165 from the selected criteria (e.g., Figure S1). Yet, when determining ΔT_{\max} , users have to keep in
 166 mind that it is a subjective process where the four presented methods may not be suitable for
 167 all regions or tree species. For example, in some regions where zero-flow conditions do not
 168 occur over many weeks, an alternative approach needs to be sought. Moreover, zero-flow
 169 conditions might not be homogeneous across the radial profile of the sapwood, which may
 170 require a different ΔT_{\max} method for the inner- *versus* outer-sapwood.

171

Equation 1.

$$K = \frac{\Delta T_{\max} - \Delta T}{\Delta T}$$

172

173 The correction for the proportion of the probe that is inserted into the non-conducting
 174 heartwood is another important data-processing step (see Peters *et al.* 2018). This correction is
 175 incorporated in the `tdm_hw.cor()` function using the sapwood thickness (*sapwood.thickness* in
 176 mm). Together with the length of the probe (*probe.length* in mm), the proportion of the probe
 177 inserted into the sapwood (γ in mm mm⁻¹) vs. the proportion of the probe in the inactive
 178 heartwood is used to calculate ΔT_{sw} (Equation 2) which replaces ΔT (Equation 1; Clearwater,
 179 Meinzer, Andrade, Goldstein & Holbrook 1999).

180

Equation 2

$$\Delta T_{sw} = \frac{(\Delta T - (1 - \gamma) \cdot \Delta T_{max})}{\gamma}$$

181

182

183

184

185

186

187

188

189

190

191

192

193

194

195

Equation 3

$$K_{res} = \frac{(a + b \cdot t)}{(1 + c \cdot t + d \cdot t^2)}$$

196

197

198

199

200

201

202

203

204

205

206

The installation of probes into living tissues results in signal dampening in K that is particularly pronounced in multi-annual observation periods (see Wiedemann, Marañón-Jiménez, Rebmann, Herbst & Cuntz 2016). When probes are inserted into the xylem without reinstalling each year, one can correct for the dampening effect with the statistical method proposed by Peters *et al.* (2018) with the `tdm_damp()` function. Yet, it should be noted that no empirical evidence (i.e., gravimetric validation) exists to confirm the correction shape of this function and should thus be implemented with caution. Moreover, such data-driven statistical approach may hamper data analyses related to inter-annual variability. Moreover, more mechanistic approaches are needed to address this issue and improve the accuracy of the approaches to correct for signal dampening. For K values obtained from preceding functions, a dampening model can be fitted to calculate corrected K values (using `nls()`; equation 3 and see `?TRES::tdm_damp` for details). After careful visual inspection of the fitted model, the fitted parameters (a , b , c , and d) can be used to correct K and scaled to the maximum values within the first year since installation (Figure 4).

SFD is defined by the relationship between K and SFD using a power-type function (equation 4). Recent calibration studies and meta-analyses have urged the application of case- or species-specific calibration experiments (e.g., Steppe, De Pauw, Doody & Teskey 2010; Fuchs, Leuschner, Link, Coners & Schuldt 2018; Flo, Martinez-vilalta, Steppe, Schuldt & Poyatos 2019). Thus, within the `tdm_cal.sfd()` one can either supply raw calibration experiment data or provide a and b parameters from the obtained calibration experiments (equation 4). Alternatively, the package also contains a database of raw calibration measurements, from 22 studies covering 37 species covering the main wood anatomy types, (Flo, Martinez-vilalta, Steppe, Schuldt & Poyatos 2019; Figure 5; see `cal.data`).

Equation 4.
$$\text{SFD} = a \cdot K^b$$

207

208 3 | TREX uncertainty and sensitivity analyses and output generation

209 *Uncertainty and sensitivity analyses*

210 Each parameter choice in TDM data-processing affects the SFD output. The uncertainty,
 211 i.e., variability in relevant SFD output caused by input parameters used in the data-processing
 212 cascade, could impact the level of confidence in tree water-use estimates which is relevant for
 213 both, mechanistic understanding of ecophysiological processes and environmental decision
 214 making (Maier, Ascough, Wattenbach, Renschler & Labiosa 2008). Besides providing a
 215 framework to quantify uncertainty, assessing the contribution of specific input parameters to
 216 the overall output uncertainty could support the identification of key parameters, that could be
 217 better constrained with additional data collection, and thus reducing the final output uncertainty
 218 (e.g., Pappas, Fatichi, Leuzinger, Wolf & Burlando 2013).

219 The *tdm_uncertain()* function performs uncertainty and global sensitivity analysis of
 220 TDM time series (using the Sobol's total sensitivity indices as implemented in the "sensitivity"
 221 R package; Iooss, Janon & Pujol 2019; see Note S1 for details). For conducting sensitivity and
 222 uncertainty analyses of TDM data with TREX, the user needs to: i) select the output variable
 223 of interest, ii) identify the relevant input parameters (some of them are method-specific), and
 224 iii) determine their range and statistical distributions (Table 2). For a given time series three
 225 output variables are considered, calculated as the mean over the entire time period where data
 226 are available, namely: i) mean daily sum of water use ("Sum", expressed as SFD or K), ii) the
 227 variability in maximum SFD or K values ("CV", coefficient of variation in % as this alters
 228 climate response correlations), and iii) the duration of daily sap flow based on SFD or K below
 229 a user-defined threshold ("Duration", expressed in hours per day dependent on a threshold, see
 230 *min.sfd* and *min.k*). The relevant input parameters are presented for each of the ΔT_{\max} methods
 231 in Table 2, (providing default values and sampling distributions; excluding *tdm_damp()* as its
 232 application demands detailed visual inspecting). Yet, users should note that parameter ranges
 233 and distributions represent an important critical component of any sensitivity analysis and
 234 should be carefully assessed (Wallach & Genard 1998).

235 The *tdm_uncertain()* function provides both graphical (Figure 6) and tabular output
 236 (Table S2) depending on the ΔT_{\max} calculation method (Figure S2-S3). The sensitivity output
 237 for each ΔT_{\max} calculation method can be compared to identify key parameters which need to

238 be constrained. In the Norway spruce example dataset, the total sensitivity indices for the Sum,
239 CV and Duration based on K values illustrate that both the Sum and CV of K are highly sensitive
240 to the sapwood correction, suggesting the need for carefully establishing sapwood depth at the
241 location of probe installation (Figure 6). For all outputs based on SFD the b parameter of the
242 power-type calibration curve appears to strongly impact both absolute water flow as well as
243 their variability, illustrating the need for robust calibration curves to constrain the output
244 variables. High sensitivity of CV to some of the input parameters suggests potential alterations
245 of SF environmental responses caused by methodological variability (Peters *et al.* 2018).

246 *Relevant output generation*

247 The package provides the functionality to temporarily aggregate the required SFD data
248 into any user-defined interval (*agg.data()*; e.g., daily sap flow values in $\text{cm}^3 \text{cm}^{-2} \text{d}^{-1}$).
249 Moreover, the package provides the *out.data()* function to generate either SFD (expressed as
250 F_d in $\text{mmol m}^{-2} \text{s}^{-1}$) or crown conductance estimates (G_C in $\text{cm}^3 \text{cm}^{-2} \text{h}^{-1} \text{kPa}^{-1}$) outputs in an
251 exportable format. Here, G_C is defined as the ratio of SFD to VPD and is analogue to stomatal
252 conductance under conditions of negligible stem capacitance, but expressed per unit of sapwood
253 area (Meinzer *et al.* 2013). The function *out.data()* offers the opportunity to decide about the
254 conditions that fulfil these negligible capacitance assumptions (see Note S2 for details). A non-
255 linear model of the form $G_C = \alpha + \beta \text{VPD}^{-0.5}$ can then be automatically fitted with the *out.data()*
256 function to the selected peak-of-day mean values of G_C on VPD to quantitatively describe the
257 observed patterns (Figure S4; as described in Pappas *et al.* 2018). This can be used to get
258 preliminary insights into tree water use strategies, yet caution is needed for the interpretations
259 of these values (i.e., capacitance assumptions need to be verified and G_C response models to
260 VPD have statistical limitations due to estimation of G_C using VPD).

261 **4 | Conclusion**

262 The TREX package contains advanced functionalities to assimilate, process and analyse
263 raw sap flow measurements obtained with the TDM. The package provides means for
264 transparent and reproducible TDM data processing and for enhancing comparability of SFD
265 estimates between studies. Moreover, to our knowledge, this is the first study to provide a state-
266 of-the-art systematic quantification of sap flow uncertainty and sensitivity due to data-
267 processing parameter inputs and assumptions. We believe that TREX provides a structured and
268 transparent pathway for sap flow data processing, which will emphasize the utility of heat-based
269 sap flow measurements for future research. The package structure eases the implementation of

270 future processing methods when made available from the scientific community, facilitating thus
271 multi-method comparisons and robust sensitivity and uncertainty analyses.

272 **Acknowledgements**

273 We thank Patrick Fonti for providing constructive comments on an earlier version of the
274 manuscript and supplying the example dataset collected at the Löttschental transect. We thank
275 Dirk J. W. De Pauw and Ana Stritih for their discussions on uncertainty analyses. We also thank
276 all researchers and their funding agencies, whose work was essential for obtaining all the
277 calibration experiment data. RLP acknowledges the support of the Swiss National Science
278 Foundation (SNSF), Grant P2BSP3_184475. CP acknowledges the support of the SNSF
279 (Grants P2EZP2_162293 and P300P2_174477), the Canada Research Chair in Atmospheric
280 Biogeosciences in High Latitudes and the Global Water Futures project Northern Water
281 Futures. RP acknowledges the support of the Spanish grant RTI2018-095297-J-I00 and of a
282 Humboldt Research Fellowship for Experienced Researchers. Data collection was performed
283 in the framework of the SNSF projects INTEGRAL (121859) and LOTFOR (150205). All
284 authors declare that there is no conflict of interest,

285 **Data availability statement**

286 The TREX package is available as a package in R. The program R is freely available from the
287 Comprehensive R Archive Network (CRAN; <http://cran.r-project.org/>). The package and
288 presented example data is also available for download and installation via a GitHub
289 (<https://github.com/the-Hull/TREX>) and Zenodo (<http://doi.org/10.5281/zenodo.4121258>)
290 repository.

291 **Authors' Contributions**

292 RLP, CP and AGH initiated the concept of the R package TREX and developed the main
293 functions. RP and VF aided in further developing the functionalities and AH mainly compiled
294 the R package. RLP and KS raised the funding. RLP and CP wrote the manuscript and all
295 authors contributed to the manuscript drafts.

296 **References**

- 297 Brinkmann, N., Eugster, W., Zweifel, R., Buchmann, N., & Kahmen, A. (2016). Temperate
298 tree species show identical response in tree water deficit but different sensitivities in sap
299 flow to summer soil drying. *Tree Physiology*, *36*(12), 1508–1519.
300 doi:10.1093/treephys/tpw062
- 301 Clearwater, M. J., Meinzer, F. C., Andrade, J. L., Goldstein, G., & Holbrook, N. M. (1999).
302 Potential errors in measurement of nonuniform sap flow using heat dissipation probes.
303 *Tree Physiology*, *19*(10), 681–687. doi:10.1093/treephys/19.10.681
- 304 De Pauw, D. J. W., Steppe, K., & De Baets, B. (2008). Identifiability analysis and
305 improvement of a tree water flow and storage model. *Mathematical Biosciences*, *211*,
306 314–332. doi:10.1016/j.mbs.2007.08.007
- 307 Dietrich, L., Delzon, S., Hoch, G., & Kahmen, A. (2019). No role for xylem embolism or
308 carbohydrate shortage in temperate trees during the severe 2015 drought. *Journal of*
309 *Ecology*, *107*(1), 334–349. doi:10.1111/1365-2745.13051
- 310 Fatichi, S., & Pappas, C. (2017). Constrained variability of modeled T:ET ratio across
311 biomes. *Geophysical Research Letters*, *44*, 6795–6803. doi:10.1002/2017GL074041
- 312 Flo, V., Martinez-vilalta, J., Steppe, K., Schuldt, B., & Poyatos, R. (2019). Agricultural and
313 Forest Meteorology A synthesis of bias and uncertainty in sap flow methods.
314 *Agricultural and Forest Meteorology*, *271*, 362–374.
315 doi:10.1016/j.agrformet.2019.03.012
- 316 Fuchs, S., Leuschner, C., Link, R., Coners, H., & Schuldt, B. (2017). Calibration and
317 comparison of thermal dissipation, heat ratio and heat field deformation sap flow probes
318 for diffuse-porous trees. *Agricultural and Forest Meteorology*, *244–245*, 151–161.
319 doi:10.1016/j.agrformet.2017.04.003
- 320 Granier, A. (1985). Une nouvelle méthode pour la mesure du flux de sève brute dans le tronc
321 des arbres. *Annals of Forest Science*, *42*, 193–200.
- 322 Iooss, B., Janon, A. & Pujol, G. (2019) Global Sensitivity Analysis of Model Outputs. R
323 Package ‘sensitivity’ version 1.16.2. Link: [https://cran.r-](https://cran.r-project.org/web/packages/sensitivity/)
324 [project.org/web/packages/sensitivity/](https://cran.r-project.org/web/packages/sensitivity/)
- 325 Köstner, B., Granier, A., & Cermák, J. (1998). Sapflow measurements in forest stands:
326 Methods and uncertainties. *Annales Des Sciences Forestieres*, *55*(1–2), 13–27.
327 doi:10.1051/forest:19980102
- 328 Lindén, J., Fonti, P., & Esper, J. (2016). Urban Forestry & Urban Greening Temporal
329 variations in microclimate cooling induced by urban trees in Mainz, Germany. *Urban*
330 *Forestry & Urban Greening*, *20*, 198–209. doi:10.1016/j.ufug.2016.09.001
- 331 Lu, P., Urban, L., & Zhao, P. (2004). Granier’s Thermal Dissipation Probe (TDP) Method for
332 Measuring Sap Flow in Trees: Theory and Practice. *Acta Botanica Sinica*, *46*(6), 631–
333 646.

- 334 Maier, H.R., Ascough, J.C., Wattenbach, M., Renschler, C.S., & Labiosa, W.B. (2008)
335 Uncertainty in Environmental Decision Making: Issues, Challenges and Future
336 Directions. Publ. from USDA-ARS/UNL Fac.
- 337 Meinzer, F. C., Woodruff, D. R., Eissenstat, D. M., Lin, H. S., Adams, T. S., & McCulloh, K.
338 A. (2013). Above-and belowground controls on water use by trees of different wood
339 types in an eastern US deciduous forest. *Tree Physiology*, *33*(4), 345–356.
340 doi:10.1093/treephys/tpt012
- 341 Oishi, A. C., Hawthorne, D. A., & Oren, R. (2016). Baseline: An open-source, interactive
342 tool for processing sap flux data from thermal dissipation probes. *SoftwareX*, *5*, 139–143.
343 doi:10.1016/j.softx.2016.07.003
- 344 Pappas, C., Fatichi, S., Leuzinger, S., Wolf, A., & Burlando, P. (2013). Sensitivity analysis of
345 a process-based ecosystem model: Pinpointing parameterization and structural issues.
346 *Journal of Geophysical Research: Biogeosciences*, *118*(2), 505–528.
347 doi:10.1002/jgrg.20035
- 348 Pappas, C., Matheny, A. M., Baltzer, J. L., Barr, A. G., Black, T. A., Bohrer, G., ... Stephens,
349 J. (2018). Boreal tree hydrodynamics: Asynchronous, diverging, yet complementary.
350 *Tree Physiology*, *38*(7), 953–964. doi:10.1093/treephys/tpy043
- 351 Peters, R. L., Fonti, P., Frank, D. C., Poyatos, R., Pappas, C., Kahmen, A., ... Steppe, K.
352 (2018). Quantification of uncertainties in conifer sap flow measured with the thermal
353 dissipation method. *New Phytologist*, *219*, 1283–1299. doi:10.1111/nph.15241
- 354 Peters, R. L., Speich, M., Pappas, C., Kahmen, A., von Arx, G., Graf Pannatier, E., Steppe,
355 K., Treydte, K., Stritih, A., & Fonti, P. (2019) Contrasting stomatal sensitivity to
356 temperature and soil drought in mature alpine conifers. *Plant, Cell & Environment*, *42*,
357 1674–1689. doi: 10.1111/pce.13500
- 358 Poyatos, R., Granda, V., Molowny-Horas, R., Mencuccini, M., Steppe, K., & Martínez-
359 Vilalta, J. (2016). SAPFLUXNET: Towards a global database of sap flow measurements.
360 *Tree Physiology*, *36*(12), 1449–1455. doi:10.1093/treephys/tpw110
- 361 R Core Team (2017). R: A language and environment for statistical computing. R Foundation
362 for Statistical Computing, Vienna, Austria. Link: <https://www.R-project.org/>.
- 363 Rabbel, I., Dieckrüger, B., Voigt, H., & Neuwirth, B. (2016). Comparing ΔT_{max}
364 determination approaches for Granier-based sapflow estimations. *Sensors*, *16*(1), 1–16.
365 doi:10.3390/s16122042
- 366 Reyes-Acosta, J. L., & Lubczynski, M. W. (2013). Mapping dry-season tree transpiration of
367 an oak woodland at the catchment scale, using object-attributes derived from satellite
368 imagery and sap flow measurements. *Agricultural and Forest Meteorology*, *174–175*,
369 184–201. doi:10.1016/j.agrformet.2013.02.012
- 370 Schlesinger, W. H., & Jasechko, S. (2014). Agricultural and Forest Meteorology
371 Transpiration in the global water cycle. *Agricultural and Forest Meteorology*, *189–190*,
372 115–117. doi:10.1016/j.agrformet.2014.01.011

- 373 Sievert, C. (2018) Interactive web-based data visualization with R, plotly, and shiny. Link:
374 <https://plotly-r.com>.
- 375 Smith, D.M., & Allen, S.J. (1996) Measurement of sap flow in plant stems. *Journal of*
376 *Experimental Botany* 47:1833–1844. doi: 10.1093/jxb/47.12.1833
- 377 Speckman, H. Ewers, B.E., & Beverly, D.P. (2018). AquaFlux : Rapid , transparent and
378 replicable analyses of plant transpiration. *Methods in Ecology and Evolution*, (11), 44–
379 50. doi:10.1111/2041-210X.13309
- 380 Steppe, K., De Pauw, D. J. W., Doody, T. M., & Teskey, R. O. (2010). A comparison of sap
381 flux density using thermal dissipation, heat pulse velocity and heat field deformation
382 methods. *Agricultural and Forest Meteorology*, 150(7–8), 1046–1056.
383 doi:10.1016/j.agrformet.2010.04.004
- 384 Steppe, K., Vandegehuchte, M. W., Tognetti, R., & Mencuccini, M. (2015). Sap flow as a key
385 trait in the understanding of plant hydraulic functioning. *Tree Physiology*, 35(4), 341–
386 345. doi:10.1093/treephys/tpv033
- 387 Wiedemann, A., Marañón-jiménez, S., Rebmann, C., Herbst, M., & Cuntz, M. (2016). An
388 empirical study of the wound effect on sap flux density measured with thermal
389 dissipation probes. *Tree Physiology*, (36), 1471–1484. doi:10.1093/treephys/tpw071
- 390 Zweifel, R., Item, H., & Häslér, R. (2001). Link between diurnal stem radius changes and tree
391 water relations. *Tree Physiology*, (21), 869–877.

392 **Figure legends**

393 **Figure 1.** Example of a thermal dissipation time series from a mature Norway spruce (*Picea*
394 *abies* Karst.) growing in an alpine valley at 1300 m a.s.l. (Lötschental, Switzerland; Peters *et*
395 *al.* 2019).

396 **Figure 2.** Schematic overview of the modules (a-c), functions and workflow of the TREX
397 package. Functions which are specifically designed to work with thermal dissipation method
398 data are indicated with “tdm_...”, while other functions can be applied on any type of sap flow
399 data.

400 **Figure 3.** An illustration of the *outlier()* function, launching a Shiny application for interactive
401 time series visualization and inspection. Users can adjust the *alpha* parameter for the automatic
402 detection of outliers. Red points in the upper panel highlight the automatically detected outliers,
403 while red points in the lower panel are manually selected points.

404 **Figure 4.** Example of the dampening model function to correct *K* (using the *tdm_damp()*
405 function) for the Norway spruce example dataset (see Figure 1). As a smooth curve is fitted and
406 it shows good agreement with the maximum values over the years presented, we consider this
407 an appropriate correction.

408 **Figure 5.** Calibration curves obtained from literature, including all wood types (using the
409 *tdm_cal.sfd()* function; mean [μ] and confidence interval [CI] over studies). The grey
410 background shows the *K* range of the specific zero flow method, which was provided as input
411 for the function and illustrates whether the uncertainty of the calibration curve (presented with
412 CI) will impact the *K* time series.

413 **Figure 6.** Visual output from the *tdm_uncertain()* function when considering the double-
414 regression ΔT_{\max} method for the Norway spruce example dataset (see Figure 1). Total Sobol'
415 sensitivity indices of the investigated parameters with their mean (coloured dots) and 95%
416 confidence intervals (vertical lines) are provided for *K* and SFD, respectively (see Note S1 for
417 details). The sensitivity analysis was conducted on the example dataset considering the time
418 window May to November of 2013. Less parameters are presented for *K* as the calibration
419 parameters (*a* and *b* in equation 2) do not affect *K*. The smaller panels on the right present the
420 uncertainty over the selected output variables. ‘Sum’ indicates the daily sum of water use
421 (expressed in $\text{cm}^3 \text{cm}^{-2} \text{d}^{-1}$ for SFD or unitless for *K*), ‘CV’ the variability of maximum SFD or
422 *K* values (coefficient of variance in %), and ‘Duration’ the duration of daily transpiration based
423 on SFD or *K* (expressed in h per day dependent on a threshold).

424 **Tables**

425 **Table 1.** Example of TREX data format of a thermal dissipation time series
 426 *example.data(type= "timestamp")*. Data of an individual tree has to be provided, including
 427 timestamp and value (in ΔT or ΔV ; here ΔV).

timestamp	value
17-4-2012 15:00	0.444
17-4-2012 15:15	0.541
17-4-2012 15:30	0.560
17-4-2012 15:45	0.568
17-4-2012 16:00	0.572
17-4-2012 16:15	0.545
...	...

428

429 **Table 2.** Selection of data-processing parameters relevant for the uncertainty and sensitivity
 430 analyses. Each ΔT_{\max} method is considered with its unique set of parameters (indicated with
 431 square brackets). Default values incorporated into TREX, operating on the example dataset, are
 432 provided as a baseline and are based on expert judgement and existing literature. Incorporated
 433 data-processing functions include: *tdm_dt.max()* calculating zero-flow conditions,
 434 *tdm_hw.cor()* implementing the heartwood correction, and *tdm_cal.sfd()* quantifying the sap
 435 flux density. The ΔT_{\max} methods include: pre-dawn (pd), moving window (mw), double
 436 regression (dr) and environmental dependent (ed).

Data-processing [ΔT_{\max} method]	Parameter	Sampling distribution	Default values
tdm_dt.max[pd/mw/dr/ed]	zero.start	Integer sampling range	2 hours
	zero.end	Integer sampling range	2 hours
tdm_dt.max[mw/dr]	max.days	Integer fixed sampling range	5-15 days
tdm_dt.max[ed]	ed.window	Integer fixed sampling range	2-4 hours
	criteria[sr]	Dependent sampling range	30%
	criteria[vpd]	Fixed sampling range	0.05-0.5 kPa
	criteria[cv]	Fixed sampling range	0.5-1 %
tdm_hw.cor [pd/mw/dr/ed]	sapwood.thickness [sw.cor]	Normal distribution	$\sigma = 16$ mm
tdm_cal.sfd [pd/mw/dr/ed]	a	Normal distribution (Log)*	$\mu = 4.085$; $\sigma = 0.628$
	b	Normal distribution*	$\mu = 1.275$; $\sigma = 0.262$

*See *cal.sfd* with Use all; Coniferous in example

437

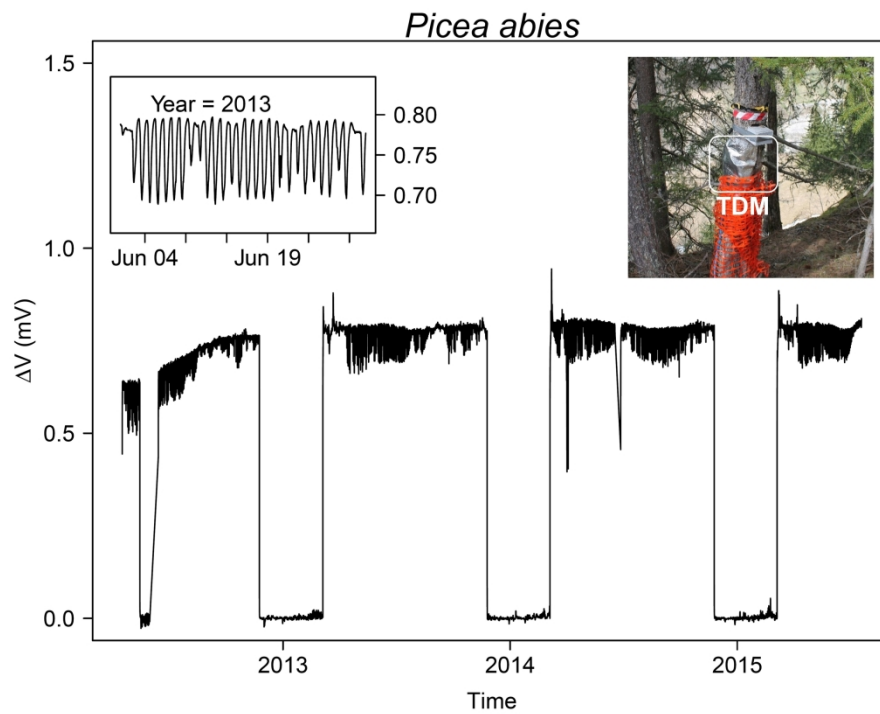


Figure 1. Example of a thermal dissipation time series from a mature Norway spruce (*Picea abies* Karst.) growing in an alpine valley at 1300 m a.s.l. (Lötschental, Switzerland; Peters *et al.* 2019).

254x190mm (300 x 300 DPI)

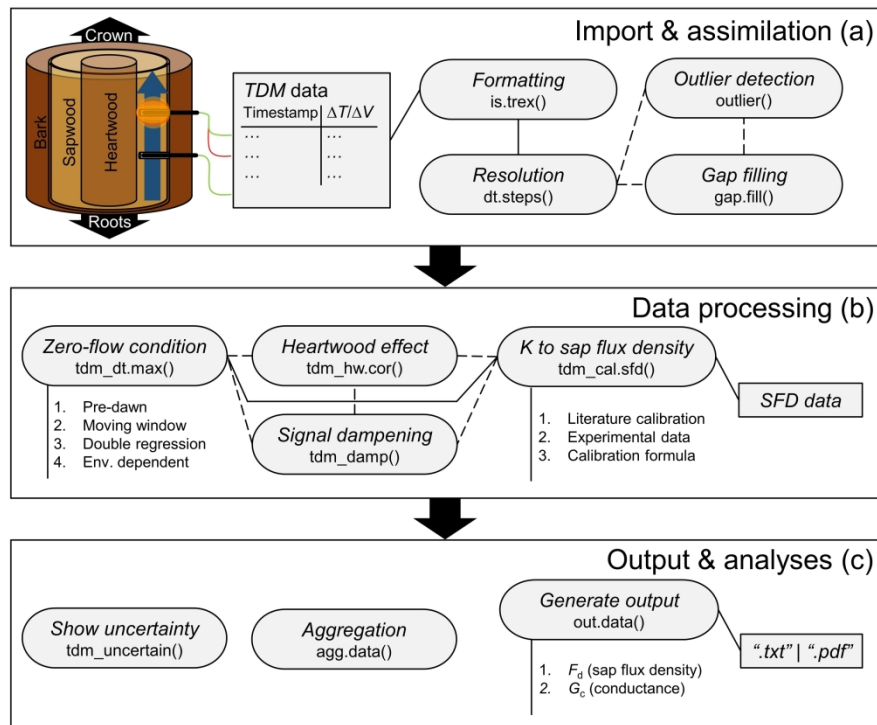


Figure 2. Schematic overview of the modules (a-c), functions and workflow of the TRES package. Functions which are specifically designed to work with thermal dissipation method data are indicated with "tdm...", while other functions can be applied on any type of sap flow data.

254x190mm (300 x 300 DPI)



Figure 3. An illustration of the *outlier()* function, launching a Shiny application for interactive time series visualization and inspection. Users can adjust the *alpha* parameter for the automatic detection of outliers. Red points in the upper panel highlight the automatically detected outliers, while red points in the lower panel are manually selected points.

254x190mm (300 x 300 DPI)

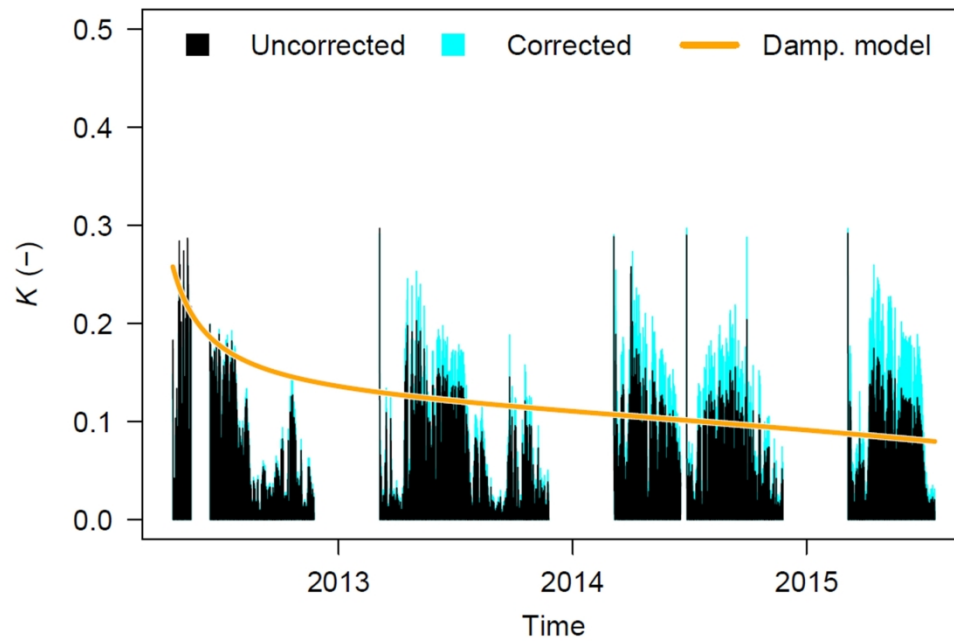


Figure 4. Example of the dampening model function to correct K (using the `tdm_damp()` function) for the Norway spruce example dataset (see Figure 1). As a smooth curve is fitted and it shows good agreement with the maximum values over the years presented, we consider this an appropriate correction.

254x190mm (300 x 300 DPI)

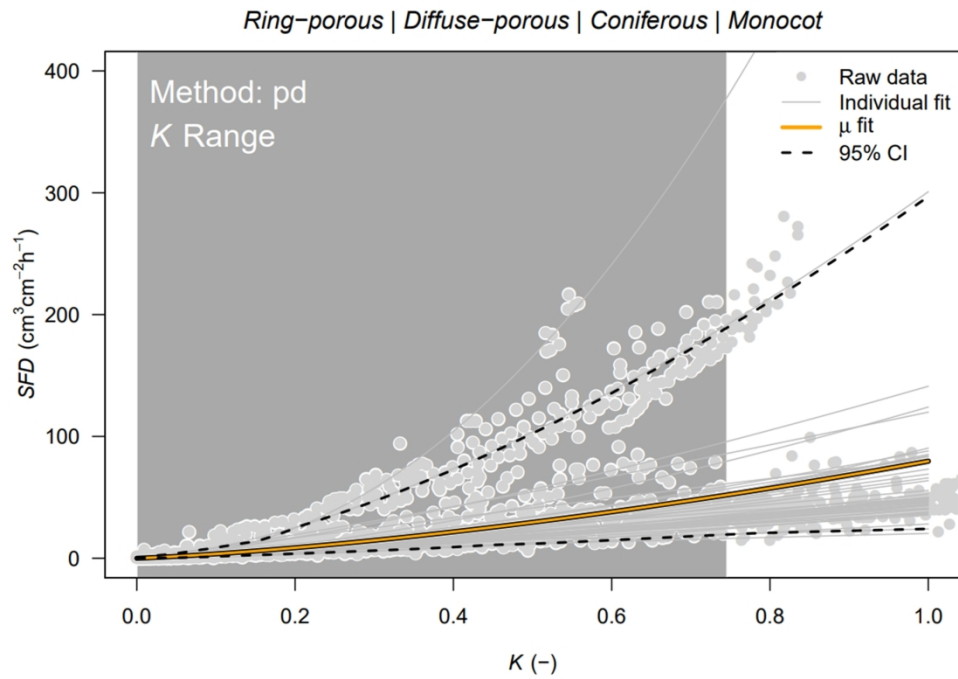


Figure 5. Calibration curves obtained from literature, including all wood types (using the *tdm_cal.sfd()* function; mean [μ] and confidence interval [CI] over studies). The grey background shows the K range of the specific zero flow method, which was provided as input for the function and illustrates whether the uncertainty of the calibration curve (presented with CI) will impact the K time series.

254x190mm (300 x 300 DPI)

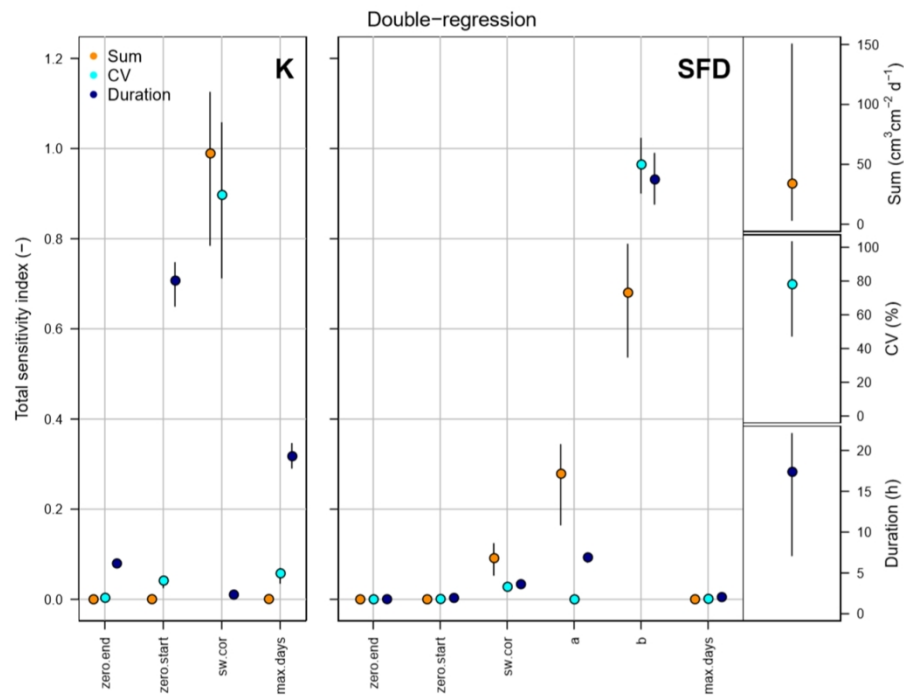


Figure 6. Visual output from the *tdm_uncertain()* function when considering the double-regression ΔT_{\max} method for the Norway spruce example dataset (see Figure 1). Total Sobol' sensitivity indices of the investigated parameters with their mean (coloured dots) and 95% confidence intervals (vertical lines) are provided for *K* and SFD, respectively (see Note S1 for details). The sensitivity analysis was conducted on the example dataset considering the time window May to November of 2013. Less parameters are presented for *K* as the calibration parameters (*a* and *b* in equation 2) do not affect *K*. The smaller panels on the right present the uncertainty over the selected output variables. 'Sum' indicates the daily sum of water use (expressed in $\text{cm}^3 \text{cm}^{-2} \text{d}^{-1}$ for SFD or unitless for *K*), 'CV' the variability of maximum SFD or *K* values (coefficient of variance in %), and 'Duration' the duration of daily transpiration based on SFD or *K* (expressed in h per day dependent on a threshold).

254x190mm (300 x 300 DPI)