SurEau-Ecos v2.0: a trait-based plant hydraulics model for simulations of plant water status and drought-induced mortality at the ecosystem level

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Abstract. A widespread increase in tree mortality has been observed around the globe, and this trend is likely to continue because of ongoing climate-induced increases in drought frequency and intensity. This raises the need to identify regions and ecosystems that are likely to experience the most frequent and significant damage. We present SurEau-Ecos, a trait-based, plant hydraulic model designed to predict tree desiccation and mortality at scales from stand to region. SurEau-Ecos draws on the general principles of the SurEau model but introduces a simplified representation of plant architecture and alternative numerical schemes. Both additions were made to facilitate model parameterization and large-scale applications. In SurEau-Ecos, the water fluxes from the soil to the atmosphere are represented through two plant organs (a leaf and a stem, which includes the volume of the trunk, roots and branches) as the product of an interface conductance and the difference between water potentials. Each organ is described by its symplasmic and apoplasmic compartments. The dynamics of a plant’s water status beyond the point of stomatal closure are explicitly represented via residual transpiration flow, plant cavitation and solicitation of plants’ water reservoirs. Each organ is described by its symplasmic and apoplasmic compartments. The dynamics of a plant’s water status beyond the point of stomatal closure are explicitly represented via residual transpiration flow, plant cavitation and solicitation of plants’ water reservoirs. In addition to the “explicit” numerical scheme of SurEau, we implemented a “semi-implicit” and “implicit” scheme. Both schemes led to a substantial gain in computing time compared to the explicit scheme (> 10000 times), and the implicit scheme was the most accurate. We also observed similar plant water dynamics between SurEau-Ecos and SurEau but slight disparities in infra-daily variations of plant water potentials, which we attributed to the differences in the representation of plant architecture between models. A global model’s sensitivity analysis revealed that factors controlling plant desiccation rates differ depending on whether leaf water potential is below or above the point of stomatal closure. Total available water for the plant, leaf area index and the leaf water potential at 50% stomatal closure mostly drove the time needed to reach stomatal closure. Once stomata are closed, resistance to cavitation, residual cuticular transpiration and plant water stocks mostly determined the time to hydraulic failure. Finally, we illustrated the potential of SurEau-Ecos to simulate regional drought-induced mortality over France. SurEau-Ecos is a promising tool to perform regional-scale predictions of drought-induced hydraulic failure, determine the most vulnerable areas and ecosystems to drying conditions, and assess the dynamics of forest flammability.

1 Introduction

Forests across many regions worldwide are experiencing record-breaking droughts followed by widespread increase in climate-driven disturbance events, including tree mortality (Allen et al., 2015; Fettig et al., 2019; Schult et al., 2020), wildfires (Ruffault et al., 2020; Abram et al., 2021) and insect outbreaks (Jactel et al., 2012). Droughts are likely to become more frequent and more intense over the next decades because of the global increase in temperatures and heat waves, which are coupled in some regions to changes in the hydrological cycle (Trenberth et al., 2014). Given the importance of forests for biochemical cycles and ecosystem ser-
vices (Seidl et al., 2014), there is a growing need for the development of models that can simulate the response of forests to extreme drought. Process-based vegetation models can help to address these issues because they represent the mechanisms governing plant physiological responses to drought and account for the interspecific and intraspecific variations of tree traits and their acclimation to a rapidly changing climate.

The science of plant hydraulics seeks to understand the physical and physiological mechanisms driving water transport in plants. This research field has proven to be a relevant theoretical framework to study the effect of global changes on plant and the terrestrial water cycle (Choe et al., 2018; Brodribb et al., 2020). Advances in plant hydraulic modeling have accelerated over the last 2 decades (Mencuccini et al., 2019; Fatichi et al., 2016) and are used as mean to tackle diverse prediction challenges, such as tree mortality (Venturas et al., 2020; De Kauwe et al., 2020), water use efficiency (Domec et al., 2017; De Caceres et al., 2021) or species distribution (Sterck et al., 2011). Many of these models were also designed (or reformed) to be integrated into land surface models and improve the representation of the feedbacks between land and climate systems (Xu et al., 2016; Li et al., 2021; Kennedy et al., 2019; Christoffersen et al., 2016). Recently, modeling water transport in plants also proved to be a promising way to assess the seasonal dynamics of live fuel moisture (foliage and twigs water content, dead to live fuel ratio), a key variable for fire behavior that could play a major role in raising forests’ flammability under climate warming (Ruffault et al., 2018a; Nolan et al., 2020).

Most plant hydraulic models represent water fluxes in plants through the mathematical approach of the soil–plant–atmosphere (SPA) continuum, wherein diffusion laws control the water flow through the soil, roots and leaves (Mencuccini et al., 2019). Water flow through plants is considered to be analogous to the electrical current through a circuit with a series of resistance and/or capacitance factors (Sperry et al., 1998). SPA models, however, vary widely in their complexity, some of them representing trees as a single resistance (Mackay et al., 2003; Williams et al., 1996), while others include multiple resistances and capacitances (Sperry et al., 1998; Tuzet et al., 2017; Couvreur et al., 2018). How physiological processes regulate plant transpiration also differs between SPA models (Mencuccini et al., 2019). Some models describe stomatal conductance through semi-empirical models (Christoffersen et al., 2016; Williams et al., 1996; Li et al., 2021; Feng et al., 2018), while others are based on optimality approaches (Wang et al., 2020; Sperry et al., 2017).

The SurEau SPA model was developed specifically to simulate plant desiccation under extreme drought and heat waves (Martin-StPaul et al., 2017; Cochard et al., 2021). As in other SPA models, SurEau describes the soil–plant–atmosphere system as a network of resistances and capacitances and computes water exchanges until stomatal closure. Additionally, SurEau simulates plant tissue desiccation beyond the point of stomatal closure by accounting for residual plant transpiration and the discharge of internal plant water stores (Fig. 1a). Unlike most current approaches (Xu et al., 2016; Tuzet et al., 2017), SurEau explicitly accounts for the differences in capacitance of the symplasmic and apoplasmic compartments, which can be calibrated from pressure-volume curves for the symplasm and vulnerability curves for the apoplasm. Symplasmic capacitances mostly buffer water fluxes during well-watered conditions, whereas apoplasm capacitances come into play when cavitation occurs (Fig. 1a). Thus, SurEau accounts for the leading role of cavitation in the dynamics of plant desiccation (Mantova et al., 2021) and the probability of plant mortality (Adams et al., 2017).

SurEau has been successfully evaluated against field cavitation observations (Cochard et al., 2021; hereafter CPRM21), has been applied in different contexts (Lemaire et al., 2021; López et al., 2021) and has performed well in predicting plant water fluxes when compared to other plant hydraulic models (McDowell et al., 2022).

As noted in CPRM21, two characteristics of SurEau impede its use for large-scale ecological applications or its integration into terrestrial biosphere models. First, SurEau requires a high number of parameters because of its detailed representation of plant architecture and the mechanisms involved in plant water exchanges. The second limitation of SurEau is its high computation time, which is partly due to the use of a first-order “explicit” numerical scheme to compute water flows. This scheme requires that variations in water quantities be computed at very small time steps to avoid numerical instabilities due to the Courant–Friedrichs–Lewy condition (CFL; Dutký, 2016). A numerical method has been proposed to overcome these instabilities and increases the time step (Xu et al., 2016; Tuzet et al., 2017), but this is not directly compatible with SurEau’s specificities regarding capacitances and cavitation. Moreover, knowledge regarding numerical physics and methods for simulation have seldom been applied to plant hydraulics.

We present SurEau-Ecos, a new SPA model meant to improve the predictions of ecosystems’ transpiration, desiccation and drought-induced mortality at scales from stand to region. SurEau-Ecos draws on the physiological and physical framework of SurEau while limiting the number of parameters and reducing computational cost. In the following sections, we first describe the principles, functioning, main equations and numerical schemes of SurEau-Ecos. Second, we compare simulations produced with three numerical schemes (explicit, semi-implicit and implicit) in terms of prediction stability and computing time. Third, we further describe the differences in plant hydraulic architecture between SurEau-Ecos and SurEau (CPRM21) and their impacts on simulation results. Fourth, we perform a global sensitivity analysis of tree desiccation dynamics to the main SurEau-Ecos input, i.e., plant hydraulic traits and stand and soil parameters. Fifth, we illustrate the potentialities which SurEau-Ecos will provide by running prospective simulations of hy-
draulic failure probability at the regional scale under changing climate.

2 Description of SurEau-Ecos

2.1 Model overview

SurEau-Ecos is a plant hydraulic model that simulates water fluxes between the soil, plant and atmosphere for a monospecific layer of vegetation. In SurEau-Ecos the soil–plant system is discretized into three soil layers and two plant compartments: a leaf and a “stem” (Fig. 1c). Each of the two plant organs contains an apoplast and a symplasm. The stem apoplasm and symplasm include water volumes of all non-leaf compartments, i.e., trunk, root and branches.

Water dynamics of the SPA system (represented by nodes in Fig. 1c) are locally governed by a generic partial differential equation for mass conservation:

\[
\frac{dq}{dt} = \nabla \cdot (k \nabla \psi) + s,
\]

where \( q \) is the water quantity (kg m\(^{-3} \)), \( k \) is the conductivity, \( \psi \) is the water potential, \( k \nabla \psi \) is the water fluxes, and \( s \) is the local sink term (i.e., a negative sign for soil evaporation or transpiration) or source term (i.e., a positive sign for precipitation and water released by cavitation).

A spatially integrated form of Eq. (1) can be specified for each compartment of the plant (Fig. 1c) to derive the rate of change of its absolute water quantity (volumetric integration). For convenience, we use the water quantity per unit of leaf area \( Q \) (kg m\(^{-2} \)) as a state variable. To account for the water fluxes between compartments and the contribution of internal water stocks (i.e., capacitances), the computations of water fluxes between two adjacent compartments \( (F_{i\rightarrow j}) \) are simulated according to Darcy’s law as the product of compartment’s interface conductance \( (K_{ij}) \) and the gradient of water potential \( (\psi) \):

\[
F_{i\rightarrow j} = k_{ij} \nabla \psi \approx K_{ij} (\psi_j - \psi_i).
\]

These fluxes are described in Sect. 2.3.

In addition, solving Eq. (1) needs to describe the link between \( Q \) and \( \psi \). This is handled using the notion of capacitance for the plant compartments and water retention curves for the soil compartments. Plant capacitances \( (C) \) are defined as follows:

\[
C = \frac{dQ}{d\psi}.
\]

For any plant compartments a generic equation of the water balance can now be written:

\[
C \frac{d\psi_i}{dt} + \sum_j K_{ij} (\psi_i - \psi_j) - S = 0.
\]

According to the type of compartment, \( S \) includes cuticular or stomatal transpiration losses or water release from cavitation, which is also accounted as a source term in the apoplasm (Cruziat et al., 2002). Cuticular or stomatal transpiration fluxes are computed differently for each compartment (leaf symplasm includes stomatal transpiration, whereas stem symplasm only include cuticular transpiration). The contribution of capacitance \( (C) \) to the plant compartment water balance is related to the saturated (or initial) water quantity \( (Q_{\text{sat}}) \) in that compartment and takes different formulation for symplasm and apoplasm. A pressure–volume curve is used for the symplasmic capacitance (Tyree and Hammel, 1972), whereas a constant capacitance is used for the apoplasm (Sect. 2.5). To the best of our knowledge, this is the first formulation of symplasmic \( C \) and cavitation flux as Darcy’s law (see details in Sect. 2.3.3. and 2.5.1). These generic forms are needed for the numerical resolution of water balance at each plant node (described in Sect. 2.2.1).

For soil compartments, the water balance of a soil layer \( j \) is computed using a generic equation following Eqs. (1) and (2), such as:

\[
\frac{dQ_\text{soil}}{dt} + K_\text{soil}_{\text{Sapo}} (\psi_{\text{soil}_j} - \psi_{\text{Sapo}}) - S = 0,
\]

where \( K_{\text{soil}_{\text{Sapo}}} \) is the conductance from the soil layer \( j \) to the stem apoplasm (Sect. 2.3.1). \( S \) represents a source (when \( S > 0 \)) or sink (when \( S < 0 \)) term that can include soil water inputs from soil infiltration; drainage from other layers; or outputs such as deep drainage, soil evaporation, or capillarity depending on the soil layer (Sect. 2.2.2). A water retention curve for the soil (van Genuchten, 1980) is used to link \( Q_{\text{soil}_j} \) and \( \psi_{\text{soil}_j} \) and solve Eq. (5) (Sect. 2.5.2).

In addition to the core soil–plant hydraulic processes driving transpiration and plant water status \( (Q \) and \( \psi) \), SurEau-Ecos also includes an empirical module for leaf phenology that controls leaf area growth and decreases during senescence (described in Appendix A) and different modules to represent the stand water balance (interception, water transfers between soil layers and drainage; described in Ruffault et al. (2013). The list of input variables and their respective units is given in Table 1.

Temporal resolution varies according to each type of process (Fig. 1b). Phenology and stand water balances are computed at a daily time step. Soil–plant hydraulic processes (i.e., soil water uptake, transpiration and hydraulic redistribution) are computed at the finer time step (from 0.01 to 1800 s depending on the resolution scheme) and driven by hourly interpolated climate, which is derived from daily climate following (De Cáceres et al., 2021) (see Table B1 for the list of daily input weather variables). The three different numerical resolution schemes currently implemented in SurEau-Ecos are described in Sect. 2.6.

All variables and processes related to stand water balance processes (precipitation, interception, drainage) are expressed per unit of ground surface area, while plant hydraulic
processes are expressed per unit of leaf surface area, in accordance with usual practices in each research field. This implies that initial water volumes of the soil and the plant (leaf and stem) are expressed per unit of soil area. Following this, leaf area index (LAI) permits the conversion of quantities from a soil area basis to a leaf area basis. If the parametrization is performed from individual tree dimensions or from forest inventories and allometries, an additional parameter is needed, the average plant foot print (aPFP, in m²), in order to scale individual plant dimensions on leaf or a soil area basis.

SurEau-Ecos was implemented in the R programming language (R Core Team, 2020). The following sections describe the equations and resolution of the model in more details.

2.2 Water balance in each compartment

2.2.1 Plant

The water balance of each of the four-plant compartments (leaf and stem symplasm and apoplasm, Fig. 1c) is determined according to the generic Eq. (4) and solved at each time step.
Table 1. Input parameters in SurEau-Ecos.

| Parameter           | Description                                                                 | Unit                |
|---------------------|-----------------------------------------------------------------------------|---------------------|
| Stand               | Maximum leaf area index of the stand                                        | m^2_leaf m^-2       |
| t0                  | Initial date of the forcing period for leaf phenology                      | DOY                 |
| T_D                 | Minimum temperature to start cumulating temperature for budburst           | °C                  |
| F*                  | Amount of forcing temperature to reach budburst                             | °C                  |
| RLAI                | LAI growth rate per day                                                     | LAI d^-1            |
| cws                 | Canopy water storage capacity                                               | mm LAI^-1           |
| k                   | Light extinction parameter                                                 | –                   |
| Plant               | Modulus of elasticity of the leaf symplasm                                  | MPa                 |
| ε_L                 | Osmotic potential at full turgor of the leaf symplasm                       | MPa                 |
| ε_S                 | Osmotic potential at full turgor of the stem symplasm                       | MPa                 |
| slopeL              | Slope of rate of leaf embolism spread at ψ_50,L                           | % MPa^-1            |
| ψ_50,L              | Water potential causing 50 % loss of leaf hydraulic conductance            | MPa                 |
| slopeS              | Slope of rate of stem embolism spread at ψ_50,S                           | % MPa^-1            |
| ψ_50,S              | Water potential causing 50 % loss of stem hydraulic conductance            | MPa                 |
| K_R:SApo,max        | Maximum conductance from the root surface to the stem apoplasm             | mmol m^-2 leaf s^-1 MPa^-1 |
| K_SApo—LApo,max     | Maximum conductance from trunk apoplasm to the leaf apoplasm               | mmol m^-2 leaf s^-1 MPa^-1 |
| K_SSsym             | Conductance from the stem apoplasm to stem symplasm                        | mmol m^-2 leaf s^-1 MPa^-1 |
| K_LSsym             | Conductance from the leaf apoplasm to leaf symplasm                        | mmol m^-2 leaf s^-1 MPa^-1 |
| α_LApo              | Leaf apoplasmic fraction                                                   | –                   |
| α_SApo              | Stem apoplasmic fraction of the wood water volume                           | –                   |
| α_SSym              | Stem symplasmic fraction of the wood water volume                           | –                   |
| C_LApo              | Capacitance of the leaf apoplasm                                           | mmol m^-2 leaf s^-1 MPa^-1 |
| C_SApo              | Capacitance of the stem apoplasm                                           | mmol m^-2 leaf s^-1 MPa^-1 |
| V_S                 | Volume of tissue of the stem (includes the root, trunk and branches)       | L m^-1 soil         |
| Succulence          | Leaf succulence (water content per unit of leaf area)                      | g m^-2 leaf         |
| LDMC                | Leaf dry matter content (dry mass over saturated mass)                     | g m^-1              |
| LMA                 | Leaf mass per area                                                         | g m^-2 leaf         |
| β                   | Shape parameter for root distribution                                       | –                   |
| RaLa                | Root-to-leaf area ratio                                                    | –                   |
| d_R                 | Root diameter                                                              | m                   |
| ψ_gS50              | Water potential causing 50 % stomatal closure                               | MPa                 |
| slope_gS            | Rate of decrease in stomatal conductance at ψ_gS,50                       | % MPa^-1            |
| g_stom_min          | Minimum stomatal conductance                                               | mmol m^-2 leaf s^-1 |
| g_stom_max          | Maximum stomatal conductance                                               | mmol m^-2 leaf s^-1 |
| δ                   | Response of g_stom to light                                               | –                   |
| T_optim             | Temperature at maximal stomatal conductance                                 | °C                  |
| T_sens              | Stomatal sensitivity to temperature                                         | °C                  |
| g_crown0            | Reference crown conductance                                                | mmol m^-2 leaf s^-1 |
| g_cut20             | Cuticular conductance at 20 °C                                             | mmol m^-2 leaf s^-1 |
| Q_10a               | Temperature dependance of g_cut when T ≤ T_phase                           | –                   |
| Q_10b               | Temperature dependance of g_cut when T > T_phase                           | –                   |
| T_phase             | Temperature for transition phase of g_cut                                  | °C                  |
| Soil                | Rock fragment content of soil layer j                                      | %                   |
| rfc_j               | Thickness of soil layer j                                                  | m                   |
| th_j                | Soil water content at saturation                                            | –                   |
| θ_s                 | Residual soil water content                                                | –                   |
| α                   | Inverse of the air entry potential                                         | MPa^-1              |
| n                   | Pore size distribution index                                                | –                   |
| l                   | Shape parameter for the Van Genuchten equation                             | –                   |
| k_sat               | Soil hydraulic conductivity at saturation                                   | mmol m^-2 soil s^-1 |
| g_soil0             | Reference soil conductance to water vapor                                  | mmol m^-2 soil s^-1 |
For the leaf apoplasm, the water balance equation is as follows.

\[ \frac{d\psi_{\text{Lapo}}}{dt} + \frac{K_{\text{SApo-\text{Lapo}}} (\psi_{\text{Lapo}} - \psi_{\text{SApo}})}{\text{Flux to stem apoplasm}} + \frac{K_{\text{LSym}} (\psi_{\text{Lapo}} - \psi_{\text{LSym}}) - F_{\text{L cav}}}{\text{Flux to leaf symplasm}} = 0 \]

(6)

The first term represents the change in water quantity related to the leaf apoplastic capacitance \(C_{\text{Lapo}}\) (mmol m\(^{-2}\) MPa\(^{-1}\)), which releases or absorbs water according to volume changes due to water potential changes \( (\psi_{\text{Lapo}} \text{, MPa}) \). Contrary to symplasmic compartments, this term is very limited in the apoplasm because the xylem wall is inelastic. Note also that cavitation is not included in this capacitance. The second and third terms are the water exchanges between the leaf apoplasm and stem apoplasm and between the leaf apoplasm and leaf symplasm, respectively. \(\psi_{\text{SApo}}\) is the water potential of the stem apoplasm, \(\psi_{\text{LSym}}\) is the water potential of the leaf symplasm, \(K_{\text{SApo-\text{Lapo}}}\) (mmol m\(^{-2}\) S\(^{-1}\) MPa\(^{-1}\)) is the conductance from the stem apoplasm to leaf apoplasm and \(K_{\text{LSym}}\) is the conductance of the leaf symplasm. This equation applies to the non-cavitated part of the xylem, which receives water from the cavitated part. This source is represented by the fourth term \(F_{\text{L cav}}\) (mmol), which corresponds to the water release by the cavitated vessels towards the non-cavitated leaf apoplasm (Hölttä et al., 2009). This term is further described in Sect. 2.3.2, where we explain how it can be expressed as a function of \(\psi_{\text{Lapo}}\).

Water balance for the stem apoplasm is calculated as follows.

\[ \frac{d\psi_{\text{SApo}}}{dt} + \sum_{j} K_{\text{soil}_j-\text{SApo}} (\psi_{\text{SApo}} - \psi_{\text{soil}_j}) = 0 \]

(7)

The first term represents the water flux related to the stem apoplastic capacitance \(C_{\text{SApo}}\) and water potential \(\psi_{\text{SApo}}\) changes during the time step. As with the leaf apoplasm, this term is in general very limited for the stem apoplasm. The second term represents the water exchange between the stem apoplasm and the three soil layers. For each soil layer \(j\), \(K_{\text{soil}_j-\text{SApo}}\) is the conductance from the soil to the stem apoplasm and \(\psi_{\text{soil}_j}\) the soil water potential. The third and fourth terms represent flux to the leaf apoplasm and stem symplasm, respectively. \(\psi_{\text{SSym}}\) is the water potential of the stem symplasm and \(K_{\text{SSym}}\) is the stem–symplasm conductance. The fifth term \(F_{\text{S cav}}\) corresponds to the water released from cavitation to the non-cavitated stem apoplasm water reservoir.

Water balance for the leaf symplasm is as follows.

\[ \frac{d\psi_{\text{LSym}}}{dt} + \frac{K_{\text{LSym}} (\psi_{\text{LSym}} - \psi_{\text{Lapo}})}{\text{Flux to leaf apoplasm}} + \frac{E_{\text{cuti}}}{} = 0 \]

(8)

The first term represents the water flux related to \(C_{\text{LSym}}\) and water potential changes of the leaf symplasm \(\psi_{\text{LSym}}\) during the time step. The second term is the exchange between leaf apoplasm leaf symplasm. The third and fourth terms represent the losses of water from the plant to the atmosphere through leaf stomatal transpiration \(E_{\text{stom}}\) and cuticular leaf transpiration \(E_{\text{cuti}}\). Note that with this formulation, leaf water losses from leaf transpiration remains lower bounded by \(E_{\text{cuti}}\), even when stomata are fully closed (\(E_{\text{stom}} = 0\)).

Water balance for the stem symplasm is as follows.

\[ \frac{d\psi_{\text{SSym}}}{dt} + \frac{K_{\text{SSym}} (\psi_{\text{SSym}} - \psi_{\text{SApo}})}{\text{Flux to stem apoplasm}} + \frac{E_{\text{cuti}}}{\text{Leaf cuticular transpiration}} = 0 \]

(9)

The first term represents the water flux related to \(C_{\text{SSym}}\) and water potential changes of the stem symplasm \(\psi_{\text{SSym}}\) during the time step. The second term is the flux to the stem apoplasm. The third term represents the losses of water from the plant to the atmosphere through minimum cortical transpiration \(E_{\text{cuti}}\).

### 2.2.2 Soil

The water balance of each of the three soil layers (Fig. 1c) is determined according to the generic Eq. (5) and solved at each time step.

For the first soil layer, the following equation is required.

\[ \frac{dQ_{\text{soil}}}{dt} + k_{\text{soil}_1-\text{SApo}} (\psi_{\text{soil}_1} - \psi_{\text{SApo}}) \cdot \text{LAI} = 0 \]

(10)

The first term \(\frac{dQ_{\text{soil}}}{dt}\) (mmol m\(^{-2}\) soil\(^{-1}\)) represents the change in soil water quantity between two consecutive time steps. The second term is the flux to the stem apoplasm. This flux is multiplied by LAI to convert water quantities from a leaf area basis to a soil area basis. \(pp_{\text{soil}}\) (mmol m\(^{-2}\) soil\(^{-1}\)) is the precipitation reaching the soil, \(D_{1\rightarrow 2}\) is the drainage (mmol m\(^{-2}\) soil\(^{-1}\)) of the first to the second layer, and \(E_{\text{soil}}\) (mmol m\(^{-2}\) soil\(^{-1}\)) is soil evaporation that occurs only from this layer.
Similarly, for the second layer the following equation is required.

\[
\frac{dQ_{\text{soil}2}}{dt} = k_{\text{soil}2-\text{SApo}}(\psi_{\text{soil}2} - \psi_{\text{SApo}}) \cdot \text{LAI} + D_{1\rightarrow2} - D_{2\rightarrow3} = 0
\]  

(11)

For the third soil layer, the following equation is required.

\[
\frac{dQ_{\text{soil}3}}{dt} = k_{\text{soil}3-\text{SApo}}(\psi_{\text{soil}3} - \psi_{\text{SApo}}) \cdot \text{LAI} + D_{2\rightarrow3} - Dd = 0
\]  

(12)

Dd is the deep drainage (mmol m\(^{-2}\)). For any layer, drainage occurs when the field capacity of the soil layer (\(\theta_{fc}\)) is overpassed. Lateral water transfer processes and upward capillary transfers between layers are neglected. At the time step of the hydraulic model (\(dt\)) the water balance of each soil layer is treated according to the losses from transpiration and from evaporation (only for the first layer). Incoming fluxes from precipitation, drainage and transfers between soil layers are treated at a daily time step (Fig. 1b). Rainfall interception and drainage are treated as in SIERRA (Mouillot et al., 2001; Ruffault et al., 2013) and follow the design principles of several other water balance models (Rambal, 1993; De Cáceres et al., 2015; Granier et al., 1999).

2.3 Conductances and fluxes

2.3.1 Plant and soil conductances

The model includes four apoplasmic conductances (three root-to-stem and one stem-to-leaf conductance), two symplasmic conductances (one for the stem and one for the leaves) and three soil-to-root conductances (\(K_{\text{soil}j-Rj}\), one per soil layer \(j\)) (Fig. 1c). Symplasmic conductances of the leaves (\(K_{\text{LSym}}\)) and stem (\(K_{\text{SSym}}\)) drive the fluxes between the symplasmic and apoplasmic compartments. These conductances are set to a constant value throughout the simulation. Xylem (i.e., apoplasmic) conductances are composed of three root-to-stem conductances in parallel (\(K_{Rj-\text{SApo}}\), one per soil layer \(j\)) and one stem-to-leaf conductance (\(K_{\text{SApo}-\text{Lapo}}\)). These conductances vary throughout the simulation from their initial value down to 0 according to the level of cavitation (expressed by the percent loss in conductance).

In practice, it is also useful to define the total plant conductance \(K_{\text{Plant}}\) as follows:

\[
K_{\text{Plant}} = \frac{1}{\sum_{j=1}^{3} K_{Rj-\text{SApo}}} + \frac{1}{K_{\text{SApo}-\text{Lapo}}} + \frac{1}{K_{\text{LSym}}}. \tag{13}
\]

The stem-to-leaf apoplasmic conductance (\(K_{\text{SApo}-\text{Lapo}}\)) is expressed as a function of the percent loss of conductance due to xylem embolism in the leaf:

\[
K_{\text{SApo}-\text{Lapo}} = K_{\text{SApo}-\text{Lapo},\max} \frac{100 - \text{PLC}_L}{100}, \tag{14}
\]

where \(K_{\text{SApo}-\text{Lapo},\max}\) is the initial (maximum) root-to-leaf conductance and \(\text{PLC}_L\) (%) is the percent loss of conductance. \(\text{PLC}_L\) is proportional to the level of xylem embolism. It occurs when the water potential drops below the capacity of the leaf xylem to support negative water potential and is computed by using the sigmoidal function (Pammenter and Vander Willigen, 1998):

\[
\text{PLC}_L = \frac{100}{1 + e^{\left(\frac{\text{slopeL}}{\text{50\%}} (\psi_{\text{SApo}} - P_{50\%})\right)}}. \tag{15}
\]

where \(P_{50\%}\) (MPa) is the water potential causing 50% loss of plant hydraulic conductance and \(\text{slopeL}\) (% MPa\(^{-1}\)) is the slope of linear rate of embolism spread per unit of water potential drop at the inflection point \(P_{50\%}\).

The apoplasmic conductance from each root \(j\) to the stem apoplasm \(K_{Rj-\text{SApo}}\) is expressed as a function of the level of embolism computed at the node of the stem apoplasm:

\[
K_{Rj-\text{SApo}} = K_{Rj-\text{SApo},\max} \frac{100 - \text{PLC}_{\text{SApo}}}{100}, \tag{16}
\]

where \(\text{PLC}_{\text{SApo}}\) is computed as \(\text{PLC}_L\) with the stem apoplasm potential (\(\psi_{\text{SApo}}\)) and vulnerability curves parameters specific to the stem (\(\text{slope}_S\) and \(P_{50\%}\)). \(K_{Rj-\text{SApo},\max}\) is the maximal root-to-stem apoplasmic conductance of layer \(j\). It is derived from fine-root area of the layer \(j\) such as

\[
K_{Rj-\text{SApo},\max} = \text{RAI}_j \times K_{R-\text{SApo}}. \tag{17}
\]

where \(K_{R-\text{SApo}}\) is the total conductance of the root system. \(\text{RAI}_j\) is the fine-root area of the layer \(j\):

\[
\text{RAI}_j = \text{RAI} \times r_j, \tag{18}
\]

where \(\text{RAI}\) is the total fine-root area that is computed from the stand leaf area index and the root-to-leaf area ratio (\(\text{RAI}_a\)) and \(r_j\) the root fraction in each soil layer, which is determined according to the equation from Jackson et al. (1996):

\[
r_j = \left(\beta^{z_{h,j}} - 1\right) + \beta^{z_{h,j} \times 100 - \beta^{z_{h,j} \times 100}}. \tag{19}
\]

where \(z_{h,j}\) is the depth (m) from the soil surface to the interface between layers \(j\) and \(j+1\), the factor of 100 converts from meters to centimeters and \(\beta\) is a species-dependent root distribution parameter (Jackson et al., 1996). Following this, the conductance between each soil layer \(j\) and the stem apoplasm \((K_{\text{soil}j-Rj})\) is determined as the result of two conductances in series, \(K_{Rj-\text{SApo}}\) and the conductance from soil to root \((K_{\text{soil}j-Rj})\):

\[
K_{\text{soil}j-\text{SApo}} = \frac{1}{K_{Rj-\text{SApo}} + K_{\text{soil}j-Rj}}. \tag{20}
\]
The conductance of the soil to fine roots $K_{\text{soil},j-R_j}$ for each soil layer $j$ is computed as follows:

$$K_{\text{soil},j-R_j} = \frac{2\pi L_{a,j}}{\ln \left( \frac{1}{r_j \sqrt{\pi L_{a,j}}} \right)} k_{\text{sat}} \text{REW}_j \left[ 1 - (1 - \text{REW}_j)^{\frac{1}{m}} \right]^2,$$  \hspace{1cm} (21)

with $L_a$ and $L_v$ the root length per soil area and soil volume for each soil layer, respectively, with both computed from soil depth and RAI$_j$, whereas $r$ is the radius of fine absorbing roots. $k_{\text{sat}}$ is the soil hydraulic conductivity at saturation, $m$ is a parameter of shape from the van Genuchten equation and $\text{REW}$ is the relative extractable water content computed as follows:

$$\text{REW} = \frac{\theta - \theta_t}{\theta_s - \theta_t},$$  \hspace{1cm} (22)

where $\theta$ is the relative water content (soil water content per unit of soil volume) changing dynamically with changes in absolute soil water reserve in the rooting zone, $\theta_s$ is the relative soil water content at saturation and $\theta_t$ is the relative soil water content at wilting point. $\theta_s$ and $\theta_t$ are parameters measured in the laboratory or derived from soil surveys with pedotransfer functions.

The total available water (TAW) for the plant can also be computed as the difference between the water quantity at field capacity ($\theta_{fc}$) and the water quantity at $\theta_t$ summed over the three soil layers as follows:

$$\text{TAW} = \sum_{j=1}^{3} \theta_{fc,j} \cdot \frac{100 - \text{rfc}_j}{100} \left( \theta_{fc} - \theta_t \right),$$  \hspace{1cm} (23)

where $\text{rfc}_j$ and $\theta_{fc,j}$ are the rock fragment content (%) and thickness (m) of the soil layer $j$, respectively. TAW is not a parameter in SurEau-Ecos but is an integrative value resulting from the interaction between soil characteristics and rooting depth.

### 2.3.2 Cavitation

SurEau-Ecos also considers the capacitive effect of cavitation (Hölttä et al., 2009), i.e., the water released to the streamflow when cavitation occurs. The non-cavitated part of the xylem receives a water flux from the cavitated part, corresponding to $F_{\text{cav}}^\text{sat}$ in Eq. (6) ($F_{\text{cav}}^\text{sat} > 0$), and is then transferred to adjacent compartments. The amount of water corresponding to a new cavitation event is derived from the quantity of water in the apoplasma at saturation ($Q_{\text{Sat}}^{\text{Lapo}}$) and the temporal variations in PLC$_L$ as follows:

$$F_{\text{cav}}^\text{sat} = Q_{\text{Sat}}^{\text{Lapo}} \max \left( \frac{\text{dPLC}_L}{\text{dt}}, 0 \right).$$  \hspace{1cm} (24)

This flux is linearized in temporal variations in $\psi_{\text{Lapo}}$ in order to express this flux in the form of a Darcy’s law to match the generic form of Eq. (2). For that purpose, we introduce an equivalent conductance ($K_{\text{Lapo}}^\text{cav}$) as follows:

$$F_{\text{L}}^\text{cav} = Q_{\text{Sat}}^{\text{Lapo}} \frac{\text{dPLC}_L}{\text{dt}} \frac{\psi}{\text{dt}} \approx K_{\text{Lapo}}^\text{cav} \max \left( 0, \psi_{\text{Lapo}} - \psi_{\text{Lapo}}^\text{Sat} \right),$$  \hspace{1cm} (25)

where $K_{\text{Lapo}}^\text{cav} = -\frac{Q_{\text{Sat}}^{\text{Lapo}} \text{PLC} \left( \psi_{\text{Lapo}}^\text{Sat} \right)}{\text{dt}}$ is the derivative of the PLC with respect to $\psi$, which is computed from the cavitation curve, and $\psi_{\text{Lapo}}^\text{Sat}$ is the minimal value of potential ever reached over time, which controls the current cavitation level ($\text{PLC}_L = \text{PLC}_L(\psi_{\text{Lapo}}^\text{sat})$). PLC$^c$ is computed as follows:

$$\text{PLC}^c = -\frac{\text{slope PLC}}{25} \max \left( 0, \psi_{\text{Lapo}} - \psi_{\text{Lapo}}^\text{Sat} \right).$$  \hspace{1cm} (26)

Following the same approach, the flux derived from the stem when cavitation occurs is defined as follows:

$$F_{\text{S}}^\text{cav} = Q_{\text{Sat}}^{\text{Lapo}} \frac{\text{dPLC}_S}{\text{dt}} \approx K_{\text{Lapo}}^\text{cav} \max \left( 0, \psi_{\text{Lapo}} - \psi_{\text{Lapo}}^\text{Sat} \right).$$  \hspace{1cm} (27)

### 2.4 Sources and sinks

#### 2.4.1 Stomatal and cuticular plant transpiration

Plants lose water through stomatal transpiration ($E_{\text{stom}}$), cuticular transpiration of the leaf ($E_{\text{cutis}}$), and cuticular transpiration of the stem ($E_{\text{cutis}}$). Cuticular transpiration of the roots is considered to be negligible and is not taken into account. The total plant transpiration $E_{\text{Plant}}$ is decomposed as the sum of the leaf ($E_L$) and wood transpiration ($E_{\text{cutis}}$):

$$E_{\text{Plant}} = E_L + E_{\text{cutis}},$$  \hspace{1cm} (28)

where $E_{\text{leaf}}$ is computed as follows:

$$E_L = E_{\text{stom}} + E_{\text{cutis}},$$  \hspace{1cm} (29)

and $E_{\text{cutis}}$ is computed as follows:

$$E_{\text{cutis}} = \frac{1}{\frac{1}{g_{\text{stom}}} + \frac{1}{g_{\text{cutis}}} + \frac{1}{g_{\text{crown}}}} \frac{\text{VPD}_{\text{leaf}}}{P_{\text{atm}}},$$  \hspace{1cm} (30)

where $\text{VPD}_{\text{leaf}}$ (MPa) is the vapor pressure deficit of the leaf, $P_{\text{atm}}$ is the atmospheric pressure (MPa), $g_{\text{stom}}$ is the stomatal conductance, $g_{\text{cutis}}$ is the cuticular conductance of the leaf, $g_{\text{bound}}$ is the conductance of the leaf boundary layer and $g_{\text{crown}}$ is the conductance of the tree crown.

$\text{VPD}_{\text{leaf}}$ is a function of leaf temperature ($T_L$). $T_L$ is computed at the leaf surface by solving the energy budget as in CPRM21. $g_{\text{bound}}$ and $g_{\text{crown}}$ are computed following Jones...
(31). $g_{\text{bound}}$ varies with leaf shape, size ($d_{\text{leaf}}$) and wind speed; $g_{\text{crown}}$ is a function of wind speed.

$g_{\text{cuti}}$ is a function of $T_L$, which is based on a single or double $Q_{10}$ equation depending on whether leaf temperature ($T_L$) is above or below the transition phase temperature ($T_{\text{Phase}}$) (Cochard, 2021):

\[
T_L \leq T_{\text{Phase}} \quad g_{\text{cuti10}} = g_{\text{cuti20}} \frac{T_{L} - 20}{10},
\]

\[
T_L > T_{\text{Phase}} \quad g_{\text{cuti10}} = g_{\text{cuti20}} \frac{T_{L} - T_{\text{Phase}}}{10},
\]

where $g_{\text{stom}}$ is the stomatal conductance taking into account the dependence of $g_{\text{stom}}$ on light, temperature and CO$_2$ concentration, as well as water status:

\[
g_{\text{stom}} = \gamma \cdot g_{\text{stom, max}},
\]

where $g_{\text{stom, max}}$ is the stomatal conductance without water stress and is determined as a function of light, temperature and CO$_2$ concentration following Jarvis (1976). $\gamma$ is a regulation factor that varies between 0 and 1 to represent stomatal closure according to $\psi_{\text{LSym}}$ and an empirical sigmoid function depending on the potential at 50% of stomatal closure ($\psi_{\text{gs50}}$) and a shape parameter (slope$_{gs}$) describing the rate of decrease in stomatal conductance per unit of water potential drop.

\[
\gamma = 1 - \frac{1}{1 + e^{-\frac{\psi_{\text{LSym}} - \psi_{\text{gs50}}}{\text{slope}_{gs}}}}
\]

2.4.2 Soil evaporation

$E_{\text{soil}}$ depends on the maximum soil conductance ($g_{\text{soil10}}$) and the REW of the first soil layer as follows:

\[
E_{\text{soil}} = g_{\text{soil10}} \frac{\text{REW}}{P_{\text{Atm}}}
\]

2.5 Capacitances

As described in Sect. 2.1, the link between $Q$ and $\psi$ are not represented in the same way for the soil and plant compartments. The notion of capacitance is used for the plants, while water retention curves are used for the soil.

2.5.1 Plant compartments

The contribution of capacitance ($C$) to the plant compartment water balance is related to the saturated (or initial) water quantity ($Q$) in that compartment. Symplasmic and apoplastic capacitances are not modeled in the same way, but both require the water volume at saturation ($Q_{\text{Sat}}$) of the considered reservoir. For the leaves, the volume of symplasmic and apoplastic reservoirs at saturation ($Q_{\text{Sat}}$ and $Q_{\text{Sat}}^{\text{Sat}}$, respectively) are defined as follows:

\[
Q_{\text{Sat}}^\text{LSym} = (1 - \alpha_{\text{LSym}}) Q_{\text{Sat}}^\text{LSym},
\]

\[
Q_{\text{Sat}}^\text{LApO} = \alpha_{\text{LApO}} Q_{\text{Sat}}^\text{LApO},
\]

with

\[
Q_{\text{Sat}}^\text{LSym} = \frac{1}{\text{DM}} = \text{succulence},
\]

where DM is the dry matter per unit of leaf area. The leaf dry matter content (LDMC), fraction of apoplastic tissue in the leaves ($\alpha_{\text{LApO}}$) and leaf mass per area (LMA) are all input parameters.

The apoplastic and symplasmic water quantities of the stem at saturation ($Q_{\text{Sat}}^{\text{SSym}}$ and $Q_{\text{Sat}}^{\text{SApO}}$, respectively) includes the volume of the roots, trunk and branches. They are computed based on the volume of the woody compartment and the water fraction of this volume as follows:

\[
Q_{\text{Sat}}^{\text{SSym}} = \frac{V_S}{M_{\text{H}_2\text{O}}} \cdot \alpha_{\text{Water}} \cdot \alpha_{\text{SSym}}
\]

\[
Q_{\text{Sat}}^{\text{SApO}} = \frac{V_S}{M_{\text{H}_2\text{O}}} \cdot \alpha_{\text{Water}} \cdot \alpha_{\text{SApO}}
\]

where $V_S$ is the volume of tissue of the stem compartment (including the root, trunk and branches), $M_{\text{H}_2\text{O}}$ is the water molar mass, $\alpha_{\text{Water}}$ is the proportion of water in this volume, and $\alpha_{\text{SApO}}$ and $\alpha_{\text{SSym}}$ are the apoplastic and symplasmic fraction of this water volume, respectively.

Symplasmic reservoirs behave as variable plant capacitances related to the pressure volume curve, which corresponds to the water quantity changes in symplasmic cells ($\frac{dQ}{dt}$). Symplasmic conductances are functions of the $Q_{\text{Sat}}^{\text{LSym}}$ and the temporal change in the symplasmic relative water content (RWC) (illustrated here for the leaf, but similar equations apply for the trunk):

\[
d\psi_{\text{LSym}} = \frac{dQ_{\text{Sat}}^{\text{LSym}}}{d\psi_{\text{LSym}}} = \frac{dQ_{\text{Sat}}^{\text{LSym}}}{d\psi_{\text{LSym}}} \frac{d\psi_{\text{LSym}}}{dt}
\]

\[
C_{\text{LSym}} = \psi_{\text{LSym}}^{\text{RWC}},
\]

with this formulation the capacitance of the leaf symplasm ($C_{\text{LSym}}$) can be written as follows:

\[
C_{\text{LSym}} = Q_{\text{Sat}}^{\text{LSym}} \text{RWC}.
\]

where RWC’ is the derivative of the RWC with respect to $\psi_{\text{LSym}}$, derived from pressure–volume curves (Tyree and Hammel, 1972; Bartlett et al., 2012).

\[
\psi_{\text{LSym}} = \left\{ \begin{array}{ll} \pi_0 - \varepsilon (1 - \text{RWC}) + \frac{\pi_0}{\text{RWC}}, & \text{RWC} \geq \frac{\pi_0}{\varepsilon} + 1 \\ \frac{\pi_0}{\text{RWC}}, & \text{RWC} \leq \frac{\pi_0}{\varepsilon} + 1 \end{array} \right.
\]
We used the following formulation for RWC’ (see the justification below for the expression above $\psi_{tp}$):

$$RWC' = \begin{cases} RWC, & \psi_{LSym} \geq \psi_{tp} = \frac{1}{\frac{1}{\psi_{Sym}} - 2RWC}, \\ \frac{\pi_0 \epsilon + \psi_{LSym}}{\psi_{LSym}}, & \psi_{LSym} < \psi_{tp} \end{cases}$$

with

$$RWC = \pi_0 + \epsilon + \psi_{LSym} + \sqrt{(\pi_0 + \epsilon + \psi_{LSym})^2 - 4\epsilon \pi_0}$$

when $\psi_{LSym} \geq \psi_{tp} = \frac{1}{\frac{1}{\psi_{Sym}} - 2RWC}$.

In the above equation the formulation for $\psi_{LSym} < \psi_{tp}$ simply results from the fact that $RWC = \frac{\pi_0}{\psi_{LSym}}$.

The case $\psi_{LSym} \geq \psi_{tp}$ was obtained from basic manipulations of the derivation of the following form of the pressure–volume curve:

$$\psi_{LSym}RWC + \pi_0RWC + \epsilon (1 - RWC)RWC - \pi_0 = 0,$$

which with a derivative with respect to $\psi_{LSym}$ becomes

$$\psi_{LSym}RWC' + RWC + \pi_0RWC' + \epsilon (1 - RWC)RWC' - \epsilon RWC'RWC' = 0,$$

and thus $RWC' = \frac{\pi_0 - \psi_{LSym} - 2RWC}{\psi_{LSym}} + \frac{\pi_0}{\psi_{LSym}}$.

Apoplastic capacitance is constant and is computed as the product between $Q_{Apo}^{Sat}$ and the specific apoplastic capacitance ($C_{Apo}$). Note that, given the very low elasticity of the xylem, this contribution is very weak.

### 2.5.2 Soil compartments

Capacitances for soil are not explicitly computed in SurEau-Ecos. Rather, soil water potentials for the different soil layers ($\psi_{soil}$, MPa) are directly computed according to the van Genuchten parametric formulation (van Genuchten, 1980):

$$\psi_{soil} = \left(\frac{1}{REW} \left[\frac{1}{\alpha} - 1\right]\right)^\frac{1}{n},$$

where $m$, $n$ and $\alpha$ are empirical parameters describing the typical sigmoidal shape of the function and $REW$ is the relative extractable water (see Eq. 21).

### 2.6 Numerical resolution

#### 2.6.1 Plant compartments

The resolution of the plant hydraulic part of SurEau-Ecos is to solve the water balance for the four hydraulic compartments (i.e., nodes in Fig. 1c), whose equation are presented in Sect. 2.2.1. Three different numerical resolution schemes were implemented to solve water balances of plant compartments. For these three schemes, water potentials were discretized between two consecutive time steps, $\psi^n$ and $\psi^{n+1}$, separated by $\delta t$. Thanks to cautious hypotheses, these equations were linearized at the first order in $\psi$, to lead to a four-equation linear system. Specifically, we neglected all variations of capacitances and conductances during a given time step ($C \approx C^n$ and $K \approx K^n$), as these variations are expected to be marginal with respect to weather changes, stomatal regulation or water release by cavitation.

The simpler explicit scheme, also implemented in SurEau, assumes that water fluxes can be expressed from the current time step $n$ (see Appendix B1 for details). From the generic water balance Eq. (4), it leads to

$$C \frac{\psi^{n+1} - \psi^n}{\delta t} + \sum_j K_j (\psi^n - \psi_j^n) + S^n = 0.$$  (49)

Rearranging this equation, the potential at the next time step $\psi^{n+1}$ can be simply computed as follows:

$$\psi^{n+1} = \psi^n + \frac{\delta t}{C} \left(\sum_j K_j (\psi_j^n - \psi^n) - S^n\right).$$  (50)

While the implementation of the explicit time integration scheme is undoubtedly the most straightforward numerical solution, it suffers from a well-known numerical constraint referred to as the CFL, which imposes very small time steps ($\delta t$) to avoid numerical instabilities:

$$\delta t \leq \frac{C}{2 \max (K_j)}.$$  (51)

This constraint implies that the smaller the $C$, the smaller the $\delta t$. An intuitive interpretation of this limitation is that the time step needs to be small enough to avoid water movements between non-adjacent cells. This constraint is particularly strong in plant xylem that is inelastic (i.e., $C$ is very small) such that apoplastic compartments cannot absorb water fluxes from their adjacent compartments when the time step is too large. This typically imposes $\delta t$ to be smaller than 10 ms (CPRM21).

A common option to avoid these numerical instabilities is to use an implicit scheme, where fluxes are estimated from the values of $\psi$ at time $n+1$ ($\psi^{n+1}$) as follows:

$$C \frac{\psi^{n+1} - \psi^n}{\delta t} + \sum_j K_j (\psi_j^{n+1} - \psi_j^n) + S^n = 0.$$  (52)

This numerical scheme is unconditionally stable, meaning that an increase in $\delta t$ will not induce numerical instabilities but might induce a loss of numerical accuracy. One very important limitation of this scheme is that the equations of the different compartments now correspond to a system of four equations that are coupled. Such a system can be linearized...
(by pieces to account for thresholds such as cavitation) and solved. In general, it implies the inversion of the matrix of the linear system, but the resolution can also be done analytically when the equations are not too many, as it is the case with SurEau-Ecos (see details in Appendix B2).

An alternative scheme, based on a semi-implicit approach, has also been recently proposed to solve water balances in plant hydraulic models while overcoming the numerical instabilities associated with an explicit formulation (Xu et al., 2016; Tuzet et al., 2017; Li et al., 2021; De Kauwe et al., 2020). Although not usual in numerical resolution approaches, this scheme has been shown to have great performance and has led to convergence in simulations with time steps on the order of 10 min (Xu et al., 2016).

This approach consists of solving the differential equation of each compartment assuming that \( \psi_j \) and \( S \) remain constant (respectively equals to \( \psi_j^n \) and \( S^n \)) as follows:

\[
C \frac{d\psi}{dt} + \sum_j K_j \left( \psi_i - \psi_j \right) + S^n = 0. \tag{53}
\]

After linearization of the coefficient, this ordinary differential equation has the following solution:

\[
\psi(u) = \psi^n e^{-\sum_j \frac{K_j}{\delta t}} u + \left( 1 - e^{-\frac{\psi^n}{\sum_j K_j}} \right) \sum_j K_j \psi_j^n - S^n \sum_j K_j. \tag{54}
\]

Therefore, \( \psi^{n+1} \) can be estimated by its value at \( u = \delta t \):

\[
\psi^{n+1} = \psi(\delta t), \tag{55}
\]

which implies that

\[
\psi^{n+1} = \eta \psi^n + (1 - \eta) \tilde{\psi}, \tag{56}
\]

with

\[
\eta = e^{-\frac{\sum_j K_j}{\delta t}}. \tag{57}
\]

and

\[
\tilde{\psi} = \frac{\sum_j K_j \psi_j^n - S^n}{\sum_j K_j}. \tag{58}
\]

One can notice here that \( \tilde{\psi} \) is the steady-state solution of the equation, typically valid when \( C = 0 \) (fully elastic media).

In practice, this formulation is equivalent to the corresponding numerical scheme (provided that \( \delta t \) is very small):

\[
C \frac{\psi^{n+1} - \psi^n}{\delta t} + \sum_j K_j \left( \psi^{n+1}_j - \psi^n_j \right) + S^n = 0. \tag{59}
\]

This formulation allows for comparing this scheme to the explicit and implicit schemes proposed above. This scheme uses \( \psi^{n+1} \) as a value for \( \psi \) (so that it remains stable) and \( \psi^n_j \) as a value of \( \psi_j \) (so that the equations of the four compartments are decoupled) and can be seen as an intermediate between the explicit and the implicit scheme. For that reason, it will be referred to as semi-implicit (Appendix B3). In theory, the water fluxes computed from values of water potentials evaluated at different time steps should be less accurate than the implicit scheme, especially when water potential changes are fast. It is thus expected that simulations require a larger time step to converge than the implicit scheme.

For the three different numerical schemes, we assume that soil potentials were estimated at the current time step \( n \) (i.e., \( \psi_{S_i} \approx \psi_j^n \)) as in the explicit formulation (instead of \( n + 1 \), as normally expected in an implicit scheme). This assumption is supported by the very small variations in soil potentials occurring during a single time and avoids the linearization of soil potential equations, which would have required unnecessary complex developments.

Source and sink fluxes \( S^{n+\frac{1}{2}} \) are computed for the climate at the middle of the time step (mid-climate between the current and next time step at \( n + \frac{1}{2} \)). For the implicit scheme, to account for the quick adjustment of stomatal regulation to climate variations, \( E^{n+\frac{1}{2}} \) accounts for linear variations in water potential \( \psi_{LSym} \) over the time step, thanks to the derivative of transpiration function \( S^{n+\frac{1}{2}} \left( \psi_{LSym}^n \right) \), also estimated for the mid-climate, but the current regulation of \( \psi_{LSym}^n \) is as follows:

\[
S^{n+\frac{1}{2}} \left( \psi_{LSym}^{n+\frac{1}{2}} \right) \approx S^{n+\frac{1}{2}} \left( \psi_{LSym}^n \right) + \frac{S^{n+\frac{1}{2}}}{2} \left( \psi_{LSym}^{n+1} - \psi_{LSym}^n \right). \tag{60}
\]

### 2.6.2 Soil compartments

Soil water balance in SurEau-Ecos is solved for each soil layer (Sect. 2.2.2) following a simple explicit scheme assuming that water fluxes can be expressed from the current time step \( n \). From the generic soil water balance Eq. (5), it leads to

\[
\frac{Q_{n+1} - Q^n}{\delta t} + K_{soil_{LSym}} \left( \psi^n - \psi_{SApo} \right) + S^n = 0. \tag{61}
\]

### 3 Impacts of numerical schemes on simulations and computation times

In this section, we explore the benefits and limitations of the three numerical schemes implemented in SurEau-Ecos to solve water fluxes, namely an explicit, semi-implicit and implicit scheme. As mentioned above, the minimal time step
required for accurate simulations is determined by computational limitations that depend on the chosen scheme. First, unlike the implicit and semi-implicit scheme, the explicit scheme is limited by the CFL, which causes numerical instabilities. We explored how much computation time can be gained by using implicit or semi-implicit schemes compared to the explicit scheme. In addition, in the case of the implicit and semi-implicit scheme, reducing the temporal resolution (i.e., increasing the time step) can also limit the accuracy of the simulation. The magnitude of corresponding errors then depends on the physiological processes at play in the plant and on the precision of the numerical scheme. We also assessed the sensitivity of model outputs to the temporal resolution (time step \( \delta t \)) for the implicit and semi-implicit schemes.

For these simulations, all inputs were set identical to those used in the section dedicated to the evaluation of SurEau-Ecos (see Sect. 4). Daily weather was kept constant, without precipitation, and simulations were run until total hydraulic failure of the plant. To compare the explicit scheme with the two other schemes, we made two slight simplifications to the model. First, we neglected the cavitation term in Eqs. (6) and (7). Indeed, the explicit numerical scheme of SurEau-Ecos cannot account for the flux term associated with water released by cavitation. This is due to the direct dependence of \( K_S^{cav} \) and \( K_S^{cav} \) on \( \delta t \) (Sect. 2.3.2) that prevents the CFL from being satisfied at any time step. Second, the values for stem and leaf of apoplastic capacitances (\( C_{SApo} \) and \( C_{LApo} \)) were increased (from about 1 \( \times 10^{-3} \) to 10 mmol m\(^{-2}\) MPa\(^{-1}\)) to decrease computational costs and ease the comparison between the numerical schemes. The CFL constraint imposed very small time steps (on the order of 1 \( \times 10^{-5} \) s) with the original values of plant apoplastic capacitance, which caused unaffordable computation times under most CPUs. Preliminary analyses showed that the impact of \( C_{SApo} \) and \( C_{LApo} \) were negligible on simulation results for values up to 50–100 mmol m\(^{-2}\) MPa.

When using the implicit or semi-implicit schemes with a relatively small time steps (\( \delta t = 10 \) s) our results show that these schemes yielded identical plant dynamics to those obtained with the explicit mode (Fig. 2). However, the gains in computation time were considerable. Computation time was divided by about 10 for the implicit and semi-implicit scheme compared to the explicit scheme. This is because \( \delta t \) had to be set to 1 s for the explicit scheme because of the CFL. Any attempt to set a \( \delta t \) above this 1 s threshold caused (as expected by the CFL) critical numerical instabilities (Fig. B1). Since some modifications to the model had to be performed for this comparison, the differences in computation times were solely indicative and were reported to illustrate the benefits of the semi-implicit and implicit schemes compared to the explicit scheme. Our results showed that the semi-implicit scheme was less accurate than the implicit scheme. Smaller time steps were required for the convergence of the model. Numerical explorations show that the semi-implicit scheme requires time steps on the order of 1 min (which is slightly slower than described in Xu et al. , 2016, which stated that 10 min was enough), whereas the time step can be generally larger than 30 min with the implicit scheme (Figs. B2 and B3).

For the implicit and semi-implicit schemes, two adaptive time steps were further implemented to reduce computation times. This improvement was based upon the assumption that smaller time steps were only required when changes in two critical processes, stomatal regulation and cavitation, were the highest. In a “normal” mode, the base time step is at 10 min but is automatically and gradually refined up to 1 min in periods of intense regulation changes, based on a criterion aimed at preventing variation in stomatal regulation and cavitation of more than 1 % between two consecutive time steps. In a “fast” mode, the base time step is at 1 h and refined up to 10 min. The implementation of adaptive time steps allowed for further increasing this gain in computing time (Table 2) without affecting plant dynamics.

### 4 Model parametrization

Due to the reduction of plant compartments, SurEau-Ecos requires fewer parameters than SurEau. However, the parametrization of plant hydraulics models can be problematic, especially for large-scale applications (i.e., for many species and stands). In order to facilitate the parametrization of SurEau-Ecos, we provided a table where we listed the most important parameters and where to find relevant datasets (Table 3). We also proposed some procedures to estimate the value of the parameters not directly available in current databases. We distinguished four different types of parameters: (i) the species-specific parameters, (ii) the plant (or stand) morphological parameters, (iii) the soil parameters and (iv) the parameters linked to hydraulic conductance.

Species-specific parameters (leaf, stomatal and hydraulic traits) can be derived from direct ecophysiological measurements or traits databases. This includes the parameters related to stomatal conductance, now available in several databases (Klein, 2014; Lin et al., 2015), and parameters of the \( p-v \) curves and vulnerability curves to cavitation both for the leaves and stems. The \( p-v \) curves are generally available for leaves (Bartlett et al., 2016), but very few data are available for the stem (but see Tyree and Yang, 1990; Meinzer et al., 2008). Until the release of additional datasets for these traits, we recommend to use the same value for the leaf and the stem symplasm. Vulnerability curves to cavitation are increasingly available at the branch and leaf level. In cases where it would be difficult to find the data for either the stem or the leaves, some hypotheses regarding the level of segmentation can be made. However, for vulnerability curves to cavitation, we recommend paying attention to the method that has been used to build the curves, as many artifacts are known to influence these values depending on the tree species (Sergent et al., 2020). Cuticular conductance at...
Table 2. Comparison of computation times between the three resolution schemes implemented in SurEau-Ecos.

| Resolution scheme          | Time step | Computational time (s) |
|----------------------------|-----------|-----------------------|
| Explicit                   | 1 s       | 1403.98*              |
| Implicit or semi-implicit   | 10 s      | 138.29                |
|                            | 1 min     | 21.42                 |
| Adaptive “normal” (10–1 min)|          | 4.78                  |
| Adaptive “fast” (60–10 min)|          | 1.49                  |

* This computational time is for indicative purposes only as several changes had to be made to the model to run it with the explicit scheme (see details in main text).

Figure 2. Comparison of the three numerical schemes implemented in SurEau-Ecos to solve water balances. Computation times for each scheme are given in Table 2.

a reference temperature ($g_{cuti20}$) and its dependence on temperature ($Q_{10a}$, $Q_{10b}$, and $T_{Phase}$) are increasingly recognized as a key trait for survival time during drought (Duursma et al., 2019) and heat waves (Cochard, 2021). $g_{cuti20}$ is increasingly available in species trait databases, but the parameters driving $g_{cut}$ dependence to temperature are far less measured (Riederer and Schreiber, 2001). Recent methodological innovations should allow a greater acquisition of this trait (see Billon et al., 2020).

The plant (or stand) morphological parameters that determine the overall leaf area index (LAI) and the plant internal water stores can be derived from forest inventory and species-specific allometries. LAI can also be derived from vegetation remote sensing data.

The soil parameters determine the total soil available water for plant (TAW, see equation 23), which depends on the volume of soil explored by roots on the one hand (i.e., a function rock fragment content and rooting depth) and the water retention curve on the other hand (i.e., the relationship between

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water potential with soil water content). Such parameters can primarily be derived from soil databases. Note, however, that such databases generally provide only pedo-physical information (textures, organic matter content, rock fragment, depth) so that it will be needed to apply pedotransfer function to compute the parameters (Tóth et al., 2017). Pedotransfer function can also be used to compute the soil hydraulic conductance \(K_{\text{sat}}\), although \(K_{\text{sat}}\) global databases are also available (Gupta et al., 2021).

Finally, the hydraulic conductance of the soil-to-leaf pathway and its repartition within the plant is rarely available (Mencuccini et al., 2019). The easiest way to obtain some values for these parameters is to compute the total maximal plant hydraulic conductance by using flux data (derived from sap flow, leaf gas exchange or remote sensing) and in situ water potential data (Mencuccini et al., 2019). The distribution between compartment can then be done by using average hydraulic architecture maps (Tyree and Ewers, 1991; Cruiziat et al., 2002). Alternatively, this can be computed from the elementary conductivity of plant organs taken from databases and plants sizes derived from inventory (De Cáceres et al., 2021).

### 5 Comparison between SurEau-Ecos and SurEau

SurEau-Ecos relies on the same biological and physical principles of SurEau (CPRM21). The soil–plant–atmosphere system is segmented and described as compartments linked together and exchanging water fluxes according to the gradients of water potential and hydraulic conductances. However, significant disparities between the implementation, parametrization and resolution of water fluxes between the two models led to some major differences in plant architecture and representation of water fluxes. It was therefore essential to confirm that both models provide comparable dynamics of the main state variables under similar conditions. This comparison of model outputs also consists at as an indirect evaluation effort of SurEau-Ecos since SurEau has been evaluated against field data (see details in CPRM21).

We identified three major differences in plant architecture and representation of hydraulic processes within the models. First, plant architecture is simpler in SurEau-Ecos than in SurEau. SurEau-Ecos represents the plant as two leaf cells (leaf apoplast and leaf symplasm) and two stem compartments that include the woody volume of branches, trunk and roots. In contrast, SurEau offers a detailed plant organ discretization (including roots, trunk, branches, leaves and buds). Second, while both models represent the belowground stems by three roots in parallel, the resistance to water flow linked to the root endoderm (a symplasmic root resistance) is not explicitly included in SurEau-Ecos contrary to SurEau. Instead, only one resistance per root, from the root entry to the stem, is accounted to mimic all possible resistance (root symplasm and apoplasm). Finally, in SurEau-Ecos, all leaf level fluxes to the atmosphere – i.e., the stomatal and the cuticular fluxes – pass through the symplasm, whereas in SurEau stomatal fluxes pass through the apoplasm and cuticular fluxes.

To compare model outputs, we performed an equivalent parameterization of the two models (see details in Fig. B4) and ran simulations until total hydraulic failure of the plant. We started the comparison with a typical plant fully described in CPRM21 whose parameters are given for each organ in Table B2. We then aggregated the values of SurEau parameters to match the following input parameters of SurEau-Ecos: water quantities of the leaf and stem compartments \(Q_{\text{LAp}}\), \(Q_{\text{LSym}}\), \(Q_{\text{SA}}\), and \(Q_{\text{SS}}\), the symplasmic conductance of the stem \(k_{\text{SS}}\), the apoplastic root-to-stem conductance \(k_{\text{R-AP}}\) and the apoplastic stem-to-leaf conductance \(k_{\text{SA-PL}}\). We also set the cuticular conductance of non-leaf organs to 0 in both models. All other sub-models, parameters and environmental forcing (weather and soil) were also set equal, including stomatal, boundary layer and crown conductance, linear approximation for the leaf energy balance, soil parameters, and hourly climatic inputs. This ensured that any divergence between models could only come from either the numerical scheme or plant hydraulic architecture.

Figure 3 shows the dynamics of water potentials, leaf transpiration and percent loss of conductance obtained when simulations were run from a wet soil profile until hydraulic failure is reached. Note that for this comparison the output of the trunk in SurEau was compared to the stem in SurEau-Ecos. For both models, at the beginning of the simulations when the soil was wet, leaf and stem water potentials followed the hourly variations in meteorological conditions, thereby reflecting the response of stomata to light and response of plant transpiration to \(g_{\text{st}}\) and VPD. As the soil reservoir emptied, stomata progressively closed according to the intensity of foliar water potential. After about 65 d for both models, the stomata permanently closed and transpiration was limited to cuticular losses that gradually accentuated the drought stress of the plant (decreased plant water potentials). Simultaneously, cavitation increased in the different organs, inducing water release from the apoplasm which partly dampened the decrease in plant water potentials. These results show that SurEau-Ecos and SurEau yielded very similar results when parameterized in such a way that plant organs had similar conductances and water reservoirs.

Despite similar dynamics, we also identified some differences in infra-daily water potentials between the two models. As a result, the time to leaf hydraulic failure was underestimated by 3 d (out of 90 d) in SurEau-Ecos compared to SurEau. These slight differences can be linked to the presence of the higher number of compartments in SurEau that increase the seasonal dampening effect of water potential compared to SurEau-Ecos where a lower number of compartments are represented. Notably, we observed some differences between the short-term (infra-daily) variations in
Table 3. Parametrization of SurEau-Ecos. Note that p–v stands for pressure–volume curve.

| Parameter | Organization Level | Importance* | Direct availability | Source | Protocol | Comments |
|-----------|--------------------|-------------|---------------------|--------|----------|----------|
| LAI\(\text{max}\) | Stand | High | Yes (remote sensing, inventory and allometries) | – | – | Dynamic parameters, can also be related to growth or photosynthesis modules |
| \(V_L\) and \(V_S\) | Leaf and stem | Intermediate | No | – | Computed from inventories or remote sensing |
| \(\text{rf}_L\) | Soil layer | High | Yes (from soil databases) | Hengl et al. (2017) | – | – |
| \(d_f\) | Soil layer | High | Partial (from soil database) | Hengl et al. (2017) | – | Not available for forest root depth |
| \(\theta_s\) | Soil layer | High | No (but can derived from soil database) | Hengl et al. (2017) | – | Derived from soil texture with pedotransfer functions |
| \(\theta_r\) | Soil layer | High | No (but can derived from soil database) | Hengl et al. (2017) | – | Derived from soil texture with pedotransfer functions |
| \(\alpha\) | Soil layer | High | No (but can derived from soil database) | Hengl et al. (2017) | – | Derived from soil texture with pedotransfer functions |
| \(n\) | Soil layer | High | No (but can derived from soil database) | Hengl et al. (2017) | – | Derived from soil texture with pedotransfer functions |
| \(I\) | Soil layer | High | No (but can derived from soil database) | Hengl et al. (2017) | – | Derived from soil texture with pedotransfer functions |
| \(k_{\text{sat}}\) | Soil layer | High | No (but can derived from soil database) | Hengl et al. (2017) | – | Derived from soil texture with pedotransfer functions |
| \(\varepsilon_L\), \(\varepsilon_S\) | Leaf and stem (symplasm) | Intermediate | Yes, for leaf (p–v curves) | Bartlett et al. (2016, 2012), Martin-StPaul et al. (2017), Guillemot et al. (2022) | – | Rarely available for stem (use leaf values instead). Note this parameter can be used to inform the stomatal conductance regulation model |
| \(\pi_{0_L}, \pi_{0_S}\) | Leaf and stem (symplasm) | Intermediate | Yes, for leaf (p–v curves) | Bartlett et al. (2016, 2012), Martin-StPaul et al. (2017), Guillemot et al. (2022) | – | Rarely available for stem (use leaf values instead). Note this parameter can be used to inform the stomatal conductance regulation model |
| \(\alpha_{\text{LApo}}, \alpha_{\text{SApo}}\) | Leaf and stem | Intermediate | Yes, for leaf (p–v curves) | Bartlett et al. (2016, 2012), Martin-StPaul et al. (2017), Guillemot et al. (2022) | – | Rarely available for stem (use leaf values instead). Note this parameter can be used to inform the stomatal conductance regulation model |
| \(g_{\text{stom,max}}\) | Leaf and stem cuticle | High | Yes (stomatal response curves) | Kattge et al. (2011) | – | – |
| \(\psi_{g_s,50}\) | Leaf stomata (symplasm) | High | Yes (stomatal response curves) | Martin-StPaul et al. (2017), Klein (2014) | – | – |
| \(\text{slope}_{gs}\) | Leaf stomata (symplasm) | Low | Yes (stomatal response curves) | Martin-StPaul et al. (2017), Klein (2014) | – | – |
| \(g_{\text{cuti20}}\) | Leaf and stem cuticle | High | Yes (Vulnerability curve) | Choat et al. (2012), Lens et al. (2016), Martin-StPaul et al. (2017) | – | Take care of segmentation and methods |
| \(Q_{10a}\) | Leaf and stem cuticle | Low | Partial (very few data) | Duursma et al. (2019) | – | – |
| \(Q_{10b}\) | Leaf and stem cuticle | Low | Partial (very few data) | Billon et al. (2020) | – | – |
| \(T_{\text{Phase}}\) | Leaf and stem cuticle | Low | Partial (very few data) | Billon et al. (2020) | – | – |
| \(P_{50}\) | Leaf and stem | High | Yes (Vulnerability curve) | Choat et al. (2012), Lens et al. (2016), Martin-StPaul et al. (2017) | – | Take care of segmentation and methods |
| \(\text{slope}\) | Leaf and stem | Low | Yes (Vulnerability curve) | Choat et al. (2012), Lens et al. (2016), Martin-StPaul et al. (2017) | – | Take care of segmentation and methods |
| \(K_{\text{Plant}}\) | Plant | High | No | Mencuccini et al. (2019) | – | Can be computed from \(K_{\text{Plant}}\) and hypothesis on resistance distribution within the plant |
| \(K_{\text{R-SApo,max}}\) | Plant | – | No | – | – | – |
| \(K_{\text{SApo-LApo,max}}\) | Plant | – | No | – | – | – |
| \(K_{\text{SSym}}\) | Plant | – | Yes | Bartlett et al. (2016) | – | At the biome scale, probably dynamics |
| \(K_{\text{LSym}}\) | Plant or soil | Low | Yes | Jackson et al. (1996) | – | – |

* The importance of each parameter was determined from preliminary analyses and the results of sensitivity analyses (Sect. 6) for reference only as it might vary depending on the species or the climate conditions. The description and unit of each parameter is given in Table 1.

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the water potential dynamics of the trunk symplasmic compartment of SurEau and the stem compartment of SurEau-Ecos (including the volume of roots, trunk and branches; see Table B2). The daily magnitude of the fluctuation in SurEau-Ecos appeared more dampened (Figs. 3, B5 and B6). The most plausible explanation for this difference is that the volume of the stem compartment in SurEau-Ecos is greater than the volume of the trunk compartment in SurEau. This is likely to lead to greater water discharge and lower water potential fluctuations in SurEau-Ecos (Fig. B6). Ongoing developments of a modular version of SurEau within the Capsis modeling platform (Dufour-Kowalski et al., 2012) will allow us to more deeply evaluate the effects of plant hydraulic architecture on the dynamics of plant desiccation.

6 Sensitivity experiments

6.1 Model sensitivity to input parameters

We carried out a variance-based sensitivity analysis to gain insights into the species traits that influence plant water dynamics in SurEau-Ecos and explore the main drivers of tree response to extreme drought. Variance-based approaches can measure sensitivity across the whole input space (i.e., it is a global method) and quantify the effect of interactions that can be unnoticed on a local sensitivity analysis approach (i.e., when moving one parameter at a time). Here, we used the Sobol’s sensitivity analysis method (Sobol, 2001) and reported “total order indices” that quantify the contribution of each parameter to the variance of the model output.

Two different physiological phases control the dynamics of plant desiccation under extreme drought, according to whether \( \psi_L^{LSym} \) is above or below the point of stomatal closure (Fig. 1a). Three time-based metrics were therefore considered to explore the sensitivity of plant desiccation to input parameters: (i) the time to hydraulic failure, (ii) the time to stomatal closure, and (iii) the survival time, defined as the time difference between hydraulic failure and stomatal closure (see an illustration in Fig. 4). We performed a sensitivity analysis for three different tree species with contrasting ecology and which exhibited various combinations of input parameters (Table 4). For each parameter, we randomly sampled a value within a range of \( \pm 20\% \) of the observed value. Starting from a wet soil, and without further precipitation, we ran simulations until hydraulic failure of the plant, defined as the moment when leaves reach 99\% loss of hydraulic conductivity (PLC\(_L\) \( \geq 99\% \)). This threshold guarantees that plant water pools were almost empty and that no other water reservoirs are available for the plant. The water content of plant tissues is probably a better indicator of plant mortality than the percent loss of conductivity (Martinez-Vilalta et al., 2019; Mantova et al., 2021). However, an accurate prediction of moisture content would require the integration of carbon metabolism (Martinez-Vilalta et al., 2019) that is currently not implemented in SurEau-Ecos. Daily climate inputs were set constant according to the simulations shown in Sect. 4. In total, we ran 700 000 simulations in the sensitivity experiment.

We based our selection of parameters used in the sensitivity analysis on the results from preliminary analyses and from the findings by CPRM21. To ease the interpretation of the results, we grouped the parameters according to several families, representing different processes: “water use” (LAImax, KPlant, TAW and \( \psi_{\text{gs},50} \)), “regulation” (\( \psi_{\text{gs},50} \)), “water leaks” (\( g_{\text{cuti}20} \), \( Q_{10a} \)), “safety” (\( P_{50} \)) and “plant internal stores” (\( V_S \)) (see definition in Table 1). The total available water (TAW) for the plant is not an input parameter in SurEau-Ecos, but it is an integrative index resulting from the interaction between soil characteristics and rooting depth. TAW is determined as the difference between the water quantity at field capacity and the water quantity at residual water content cumulated over the three soil layers. To make TAW vary in simulations without affecting soil physical properties, we adjusted rooting depth to match the targeted TAW.

Our results showed that a few parameters explained most of the variability in the response of trees to extreme drought (Fig. 4), although their importance largely depended on the physiological phase under study. The parameters related to “water use” (LAImax, TAW and \( K_{\text{Plant}} \)) and “regulation” (\( \psi_{\text{gs},50} \)) mainly explained the variance in time to stomatal closure, i.e., the first physiological phase. It suggests that, in this phase, interactions between how much water is available in the soil (TAW) and how fast plant transpiration will empty that reservoir (LAImax, \( \psi_{\text{gs},50} \) and \( K_{\text{Plant}} \)) determine the time to stomatal closure. The surprisingly relative low influence of \( \psi_{\text{gs},50} \) on the time to stomatal closure could be explained by the fact that, with that set of parameters and environmental conditions, \( K_{\text{Plant}} \) has a more limiting impact on plant transpiration than \( \psi_{\text{gs},50} \). In the second phase (after stomatal closure), survival time was mostly driven by parameters related “water use” (LAI, \( P_{50} \)), “water leaks” (\( g_{\text{cuti}20} \)), “safety” (\( P_{50} \)) and “plant internal stores” (\( V_S \)). In that phase, the importance of TAW and \( \psi_{\text{gs},50} \) decreased to the benefit of traits related to the rate of water losses through cuticular transpiration (\( g_{\text{cuti}20} \) and \( Q_{10a} \)); the volume of water reservoirs in the root, trunk, and branches (\( V_S \)); and plant resistance to cavitation (\( P_{50} \)). When both phases were considered jointly, we observed that the variability in the time to hydraulic failure was mainly associated with stand parameters (LAI and TAW) and to a lesser extent with \( \psi_{\text{gs},50} \) and \( g_{\text{cuti}20} \).

We also observed that the patterns described here above were almost identical regardless of the vegetation type under study. In particular, the parameters controlling “time to hydraulic failure” and “survival time” were similar among the three studied vegetation types, suggesting a similarity of plant adaptation strategies to avoid hydraulic failure in a changing climate. The one exception to this pattern is the importance of varying plant resistance to cavitation (\( P_{50} \)) in survival time. The influence of \( P_{50} \) ranged from low for Quer-
Figure 3. Comparison of the dynamics of plant water status between SurEau-Ecos and SurEau.

cus ilex (about 0.05) to very important for Quercus petraea (about 0.37). This observation suggests that less drought-resistant species (with higher $P_{50}$) receive a more direct benefit when lowering their $P_{50}$ to increase their survival time than drought-resistant species (with lower $P_{50}$). This might be due to the nonlinear response of water potential to soil and plant water content, which implies that the rate of change of plant water potential increases as soil and plant water content decreases.

Our results shed some light on our understanding of plant functioning under extreme drought. We highlighted the prominent role of stand traits, namely LAI$_{\text{max}}$, TAW and $\psi_{50}$, in determining the time needed to reach stomatal closure. In contrast, physiological variables, namely $g_{\text{cut20}}$, $Q_{10a}$, $V_s$ and $\psi_{50.L}$, played a more important role in determining “survival time”. Two improvements to the present analyses may strengthen these findings. First, numerous correlations exist between those traits, reflecting trade-offs and plant functioning strategies (Christoffersen et al., 2016; Martin-StPaul et al., 2017) that we did not take into account. Similarly, it has been shown that LAI$_{\text{max}}$ and TAW covary because trees with a higher amount of available water tend to develop a higher leaf surface value (Hoff and Rambal, 2003). Second, the relative importance of input parameters is likely to be influenced by climate. For instance, we would expect the influence of $g_{\text{min20}}$, $Q_{10a}$ and $Q_{10b}$ on survival time to increase when temperature increases, following previous results showing the vulnerability of trees during heat waves (Cochard, 2021). Integrating these potential improvements in future simulations may further help to elucidate the specific spatial and temporal patterns of drought-induced mortality (Meir et al., 2015).

6.2 Model sensitivity to the inclusion of symplasmic and apoplasmic capacitances

Whether or not plant hydraulic capacitances are explicitly taken into account is one of the key distinctions between current large-scale plant hydraulic models. Some models represent trees as single- or multiple-resistance organisms (e.g., Kennedy et al., 2019), while others like SurEau and SurEau-Ecos also include one or several hydraulic capacitances. SurEau and SurEau-Ecos describe the soil–plant–atmosphere system as a network of resistances and capacitances while introducing a novel distinction between the symplasmic and apoplasmic capacitances. This approach is beneficial for model parametrization and to derive values such as water content, as has already been discussed in Sect. 2. However, the role and importance of both the symplasmic and apoplasmic capacitances for plant survival and water dynamics have not yet been studied.

To further understand the role hydraulic capacitances on plant water dynamics, we conducted sensitivity experiments where capacitances were successively set to zero: first apoplasmic capacitance (leaf and stem), then symplasmic capacitance (leaf and stem), and finally both apoplasmic and symplasmic capacitances. These simulations applied the same experiment settings as the model comparison experiment (see Sect. 4), i.e., similar plant parameters, soil and climate conditions.

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Figure 4. Global sensitivity analysis of plant desiccation dynamics to the main hydraulic traits and stand parameters in SurEau-Ecos, shown for three different tree species with contrasting ecology and which exhibited various combinations of input parameters. We explored the sensitivity of three physiological time-based metrics to input parameters: time to stomatal closure, time to hydraulic failure and survival time. These three metrics describe the two different physiological phases controlling the dynamics of plant desiccation under extreme drought, according to whether $\psi_{LSym}$ is above or below the point of stomatal closure (Fig. 1a). All traits varied $\pm 20\%$ around their original value. TAW is the total available water for the plant.

Figure 5 shows the results of the simulations of the sensitivity experiment. Overall, hydraulic capacitances induced significant differences in both the dynamics of plant water potentials and the time to hydraulic failure. More specifically, we observed that symplasmic capacitance can buffer short-term variations in plant water potentials and therefore induced fewer negative values at midday. Apoplastic capacitances played a major role in both delaying the time to hydraulic failure and buffering daily variations in plant water potentials by providing water when cavitation occurs. The importance of this effect increases with decreasing water potentials (increasing drought). Our results therefore suggest that the representation of plant water storage greatly affects the simulations of plant water dynamics. Further studies aimed at measuring plant water content will help to validate and affine the role of plant water storage for tree response to extreme drought.

7 Regional prediction of climate-change impacts on tree mortality

In this section, we aimed to illustrate the potentialities offered by SurEau-Ecos for improving our understanding of forest response to drought. We explored whether the probability of plant hydraulic failure simulated by SurEau-Ecos was related to the distribution of two tree species at their southern distribution margin, and we then used this information to identify future areas at risk of drought-induced tree mortality. Specifically, we hypothesized that hydraulic failure was a significant constraint to tree distribution at the regional level.

We quantified the probability of hydraulic failure over France ($544,000\,km^2$) for two different species chosen for their contrasted functioning strategies: an evergreen Mediterranean oak ($Quercus ilex$) and a temperate deciduous European beech ($Fagus sylvatica$) (see main parameters in Table 3). $Quercus ilex$ is a drought-resistant species with low LAI, $P_{50}$, and deep root systems to extract water from cracks in the bedrock during drought. In contrast, $Fagus sylvatica$...
Figure 5. Model sensitivity to the inclusion of symplasmic and apoplastic capacitances in SurEau-Ecos.

Table 4. Main parameter values used in the model simulations whose results are shown in Figs. 4 and 5. Parameters derived from pressure–volume curves and PLC curves were set equal for the leaf and stem ($\varepsilon_L = \varepsilon_L$, $\pi_0_L = \pi_0_L$, $\text{slope}_{gs} = \text{slope}_{L}$, $\psi_{50,L} = \psi_{50,L}$). The description and unit of each parameter is given in Table 1.

| Parameter | Quercus ilex | Fagus sylvatica | Quercus petraea |
|-----------|--------------|----------------|----------------|
| LAI$_{max}$ | 3 | 5 | 6 |
| $\varepsilon_L$ | 10 | 10 | 10 |
| $\pi_0_L$ | -2.1 | -1.9 | -2.1 |
| $\psi_{gs,50}$ | -2.8 | -1.8 | -2.34 |
| slope$_{gs}$ | 44 | 130 | 92 |
| $g_{cut20}$ | 3 | 4 | 3 |
| $Q_{10a}$ | 1.2 | 1.2 | 1.2 |
| $Q_{10b}$ | 4.8 | 4.8 | 4.8 |
| $T_{Phase}$ | 37.5 | 39 | 42 |
| $\psi_{50,L}$ | -7 | -3.15 | -3.4 |
| slope$_{L}$ | 30 | 40 | 60 |
| K$_{Plant,max}$ | 0.62 | 1 | 1 |
| K$_{SSym}$ | 0.26 | 0.26 | 0.26 |
| $g_{stom,max}$ | 200 | 200 | 200 |
| $\beta$ | 0.97 | 0.98 | 0.97 |
| VS | 20 | 33 | 40 |
| TAW | 160 | 160 | 160 |

is characterized by a higher vulnerability to drought (higher $P_{50}$) and higher LAI values. As in Sect. 5, we defined hydraulic failure as the point when leaves reach 99% loss of hydraulic conductivity ($\text{PLC}_L \geq 99\%$). For each period investigated, we reported the probability of hydraulic failure as the frequency of years during which $\text{PLC}_L \geq 99\%$.

We ran simulations for present (1991–2020) and future (2071–2100) periods at an 8 km$^2$ resolution over France for both species. Climate data for the present period (1970–2020) were extracted from the SAFRAN climate reanalysis database (Vidal et al., 2010), which covers France at an 8 km$^2$ resolution. Projections of climate variables for the future climate period (2071–2100) were obtained from a climate simulation program involved in the fifth phase of the Coupled Model Intercomparison Project (CMIP5) and produced as part of the EURO-CORDEX initiative (Kotlarski et al., 2014). One single global circulation model–regional climate model (GCM–RCM) couple was extracted for these analyses (i.e., MPI-ESM-REMO2009), which was chosen because of its averaged climate trajectory over France when compared to an ensemble of GCM–RCM couples (Fargeon et al., 2020; Ruffault et al., 2020). Data were extracted at a 0.44° spatial resolution for the historical (1990–2005) and future (2006–2099) periods. Model outputs were bias-corrected and down-scaled at the 8 km$^2$ resolution using a quantile–quantile correction approach (Ruffault et al., 2014).
To apply the model at the landscape scale, we made several simplifying assumptions. First, we assumed that each 8 km² grid cell was covered by trees of the same species and that LAI was set to a constant value representative of observed values for the considered species (Table 3). Second, soil characteristics were also set constant over the territory. Both assumptions are unrealistic because stand characteristics vary at the local scale and have a primordial role in the probability of hydraulic failure (Sect. 5). However, as we aimed to assess the regional (rather than the local) vulnerability of tree species to changes in climate, we did not expect this to be a main limitation, provided that the results of these simulations be interpreted accordingly to these assumptions. To assess whether the probability of hydraulic failure was a good proxy of the current southern range of tree species distribution, we compared the results of our simulations with presence and absence data for each species. Tree species data were extracted from the national forest inventory database (available at http://www.ifn.fr, last access: 12 May 2022) and aggregated to obtain presence–absence on the 8 km studied grid following Cheaib et al. (2012).

Maps of probability of hydraulic failure (probability of reaching PLC ≥ 99 %) are shown in Fig. 6. We observed contrasting regional patterns according to the species under study. We observed a higher probability of mortality in southeastern France for both species, but the probability of hydraulic failure was higher for the European beech than for the holm oak. In the rest of the country, the probability of hydraulic failure was almost 0 for the holm oak. In contrast, we observed probabilities up to 50 % for the European beech in the western part and middle of the country, where the climate is temperate. When comparing these results with the maps of current species distribution, we observed a reasonable degree of spatial agreement between our simulations and presence/absence data. European beech was predominantly present in areas where our simulations indicated a probability of drought-induced mortality equal to 0 %. However, we could not interpret the results for the holm oak in the same way since the current distribution of this species indicates that the southern climate margin is not reached in the present climate. In the parts of the country where summer drought is less intense, several other factors might explain why Quercus ilex is currently not observed, including competition from more productive species, cold resistance or even forest management policies.

Our projections for the end of the century showed a future increase in the areas characterized by a high risk of hydraulic failure over France. For Fagus sylvatica, the areas characterized by a high risk of hydraulic failure will extend towards the northeast and west of the country (i.e., over the major part of the territory). For Quercus ilex, our simulations indicated that the probability of hydraulic failure should significantly increase in southeastern France, where this species is currently widespread.

Altogether, these results indicate that future climate conditions might overcome the capacity of the two studied tree species to face drought over France, which might increase the likelihood of tree mortality and wildfires in the future. Adding information about the LAI and soil physical properties might further refine our simulation results. LAI can be estimated from remote sensing indices (see for instance (De Kauwe et al., 2020)). However, TAW estimations are more problematic because information about root depth is rarely available (Ruffault et al., 2013; Venturas et al., 2020).

8 Limitations and future developments

SurEau-Ecos can already be applied as a standalone model to understand plant water dynamics and can be used in a wide of research applications, from stand-scale estimations of water fluxes to regional predictions of drought-induced mortality (see Sect. 7). In addition, the specific distinction between the symplasmic and apoplastic compartments implemented in SurEau-Ecos provides a solid foundation for predicting and monitoring water storage in the plant, a key factor in ecosystem disturbances such as mortality (Martinez-Vilalta et al., 2019) and wildfires (Ruffault et al., 2018b; Pimont et al., 2019).

The development of several supplementary key processes also warrants future consideration to extend the range of research questions and applications that SurEau-Ecos would be able to address. First, SurEau-Ecos currently simulates plant water dynamics for a single tree species for a homogeneous forest stand, and it therefore neglects the effects of species interactions on tree response to drought. This would, however, require us to affine the current representation of water competition between trees and microclimatic effects. Such developments would not only provide a mechanistic basis for multi-species modeling but could also help us to better understand the processes driving heterogenous mortality in the canopy and integrate the effects of forest management on stand structure microclimatic conditions. Another important limitation of SurEau-Ecos is that it does not simulate the processes related to photosynthesis, respiration, growth and carbon allocation. Future developments will aim at integrating SurEau-Ecos with other forest models that are designed to represent the carbon cycle and vegetation dynamics, including the forest growth models CASTANEA (Dufrène et al., 2005) and GO+ (Moreaux et al., 2020), as well as the gap model ForCEEPS (Morin et al., 2021) under the Capsis platform (Dufour-Kowalski et al., 2012). These future research projects and developments will also be an opportunity to further evaluate the feedbacks between carbon balance, growth metabolism and hydraulic properties, including the impacts of post-drought growth on the recovery of hydraulic properties and therefore on tree vulnerability to water stress in the long run (Arend et al., 2022).
Figure 6. Probability of hydraulic failure (%) over the past (1991–2020) and future (2017–2100) period for two tree species in France simulated with SurEau-Ecos. The current distribution is shown for comparison with the simulated risk of hydraulic failure.

9 Conclusion

Drought is arguably one of the most important natural disturbances threatening forest ecosystems in a number of regions worldwide (Allen et al., 2015). The challenges facing our understanding of the role of plant hydraulics in vegetation dynamics are numerous (McDowell et al., 2019), with one being the ability of current vegetation models, including those based on plant hydraulics, to predict plant desiccation dynamics at regional scales (Venturas et al., 2020; De Kauwe et al., 2020; Rowland et al., 2021; Trugman et al., 2021). Here, we presented SurEau-Ecos, a new plant hydraulic SPA model aimed at predicting plant water status and drought-induced mortality at scales from stand to region. SurEau-Ecos was designed to simulate the plant water status of the different plant’s compartments, while at the same time balancing for the needs of input parameters and computational requirements. SurEau-Ecos simulates key mechanisms associated with plant desiccation during drought and heat waves, including the dynamics of plant’s water status beyond the point of stomatal closure via residual transpiration flow, plant cavitation and the solicitation of plants’ water reservoirs. We showed that SurEau-Ecos was able to provide accurate estimations of plant water status dynamics compared to the SurEau model, despite the latter representing plant hydraulics mechanisms in more detail. This confirms that, for large-scale applications, the changes we implemented in SurEau-Ecos largely outweigh a potential loss of accuracy associated with the simplification of plant architecture and hydraulic processes. SurEau-Ecos provides the capability for us to better understand the role of plant hydraulics in vegetation dynamics under climate-change conditions characterized by increased drought frequency.

Appendix A: Leaf phenology module in SurEau-Ecos

Leaf area index (LAI) of the stand is updated daily. Species can have either evergreen or winter deciduous phenology. Evergreen species are assumed to maintain a constant LAI throughout the year. LAI values of deciduous plants are adjusted as a function of leaf phenology (∅) and the maximum of the stand (LAI_max) as follows:

\[
LAI = \varnothing \cdot LAI_{\text{max}}.
\]

∅ is set to 0 until budburst occurs. Budburst is assumed to be driven by the cumulative effect of forcing temperatures (R_f).
on bud development (Chuine and Cour, 1999) as follows:

\[
\sum_{t_0}^{t_f} R_f(T_d) \geq F^*, \tag{A2}
\]

where \(t_0\) is a parameter defining the initial date of the forcing period, \(t_f\) the budburst date and \(F^*\) is a parameter defining the amount of forcing temperature to reach budburst. Once budburst is reached, \(\varnothing\) increases from 0 to 1 at a rate specified by a parameter describing the LAI growth rate per day (\(R_{LAI}\)). In autumn, leaf fall occurs (\(\varnothing\) starts to decline) when the average daily temperature falls below 5 °C (Sitch et al., 2003; De Cáceres et al., 2015) and then \(\varnothing\) declines at a similar rate to LAI growth in spring.

Appendix B: Additional tables and figures

Table B1. Daily climate input variables.

| Symbol | Unit          | Description               |
|--------|---------------|---------------------------|
| \(T_{\text{mean}}\) | °C            | Mean temperature          |
| \(T_{\text{min}}\) | °C            | Minimum temperature       |
| \(T_{\text{max}}\) | °C            | Maximum temperature       |
| \(R_{\text{global}}\) | MJ m\(^{-2}\) | Global radiation          |
| \(\text{ppt}\) | mm            | Precipitation             |
| \(R_{H,\text{mean}}\) | %             | Mean relative humidity    |
| \(R_{H,\text{min}}\) | %             | Minimum relative humidity |
| \(R_{H,\text{max}}\) | %             | Maximum relative humidity |
| \(u\) | m s\(^{-1}\) | Mean wind speed           |

Figure B1. Illustration of the constraint on \(\delta t\) due to the Courant–Friedrichs–Lewy (CFL) condition in SurEau-Ecos. Numerical instabilities are observed when \(\delta t > 2 \cdot K/C\).
Table B2. The main physiological parameters of plant compartments used for the comparison between SurEau and SurEau-Ecos.

| Parameters | Leaf | Branches | Trunk | Root |
|------------|------|----------|-------|------|
| **SurEau “trunk-only”** | | | | |
| **Symplasm** | | | | |
| $\pi_0$ (MPa) | 0 | 0 | 0 | 0 |
| $\varepsilon$ (MPa$^{-1}$) | 10 | 10 | 10 | 10 |
| $K$ (mmol s$^{-1}$ MPa$^{-1}$ m$^{-2}$) | 1.80 | 0.55 | 0.26 | 7.22 |
| $Q_{sat}$ (mol m$^{-2}$) | 43.75 | 91.12 | 355.73 | 377.77 |
| Surface (m$^2$) | 10.5 | 5.8 | 2.7 | 54.1 |
| **Apoplasm** | | | | |
| $P_{50}$ (MPa) | 0 | 0 | 0 | 0 |
| Slope (% MPa$^{-1}$) | 60 | 60 | 60 | 60 |
| $K$ (mmol s$^{-1}$ MPa$^{-1}$ m$^{-2}$) | 1.32 | 6.50 | 7.16 | 2.03 |
| $Q_{sat}$ (mol m$^{-2}$) | 14.58 | 182.25 | 711.46 | 658.8 |
| **SurEau-Ecos** | | | | |
| **Symplasm** | | | | |
| $\pi_0$ (MPa) | 0 | 0 | 0 | 0 |
| $\varepsilon$ (MPa$^{-1}$) | 10 | 10 | 10 | 10 |
| $K$ (mmol s$^{-1}$ MPa$^{-1}$ m$^{-2}$) | 1.80 | 0.84 | 0.84 | 2.03 |
| $Q_{sat}$ (mol m$^{-2}$) | 4.16 | 78.53 | 78.53 | 78.53 |
| Surface (m$^2$) | 10.5 | 62.6 (All wood) | 62.6 (All wood) | 62.6 (All wood) |
| **Apoplasm** | | | | |
| $P_{50}$ (MPa) | 0 | 0 | 0 | 0 |
| Slope (% MPa$^{-1}$) | 60 | 60 | 60 | 60 |
| $K$ (mmol s$^{-1}$ MPa$^{-1}$ m$^{-2}$) | 1.32 | 3.4 | 3.4 | 3.4 |
| $Q_{sat}$ (mol m$^{-2}$) | 1.4 | 148 | 148 | 148 |
Figure B2. Impact of the time step ($\delta t$) on simulation results with the implicit resolution scheme.

Figure B3. Impact of the time step ($\delta t$) on simulation results with the semi-implicit resolution scheme.
Figure B4. Comparison between the plant architecture in SurEau and SurEau-Ecos. $Q$ indicates the water quantities of the compartments, $P$ the water potentials, $k$ the hydraulic conductances, $gs$ the gaseous stomatal conductances and $g_{\text{cuti}}$ the gaseous cuticular conductances. The subscripts “Apo”, “Sym” and “Endo” indicate the apoplasm, symplasm and endoderm compartments, respectively. The subscripts L, S, B, T and evap stand for leaf, stem, branch, trunk, root and evaporative sites, respectively.
Figure B5. Comparison of hourly outputs between SurEau and SurEau-Ecos for the first day of simulation. Climatic inputs (radiation, temperature and VPD) are shown in the upper panels.
Figure B6. Comparison of the water potential and the water discharge dynamics for the first day of simulation between the trunk compartment of SurEau and the stem compartment of SurEau-Ecos for two different parameterizations. In the upper row, SurEau-Ecos was parameterized with the stem symplasmic water volume computed as the sum of the symplasmic water volumes of the roots, trunk and branches of SurEau. In lower row, the stem symplasmic water volume of SurEau-Ecos was considered to be equivalent to the trunk water volume of SurEau (see Table B2 for details).
Appendix C: Numerical schemes

C1 Explicit scheme

Let

\[
\delta_{\text{cav}}^L = \begin{cases} 
0, & \text{if } \psi_{LApo}^n \geq \psi_{\text{mem}}^{LApo} \text{ (no cavitation event)} \\
1, & \text{if } \psi_{LApo}^n < \psi_{\text{mem}}^{LApo} \text{ (cavitation event)}
\end{cases},
\]

and let

\[
\delta_{\text{cav}}^S = \begin{cases} 
0, & \text{if } \psi_{SApo}^n \geq \psi_{\text{mem}}^{SApo} \text{ (no cavitation event)} \\
1, & \text{if } \psi_{SApo}^n < \psi_{\text{mem}}^{SApo} \text{ (cavitation event)}
\end{cases}.
\]

Applying the explicit scheme (Eq. 48 in main text) to the four water balance equations (Eqs. 6 to 9 in main text) gives the following equations.

Equation (6) can be rearranged to determine \( \psi_{LApo}^{n+1} \):

\[
\psi_{LApo}^{n+1} = \psi_{LApo}^n + \delta \frac{t}{C_{LApo}} \left( \frac{K_{LApo}}{C_{LApo}} \left( \psi_{LApo}^n - \psi_{SApo}^n \right) + K_{LSym} \left( \psi_{LApo}^n - \psi_{LApo}^n \right) \right) + K_{LSym} \left( \psi_{SApo}^n - \psi_{LApo}^n \right).
\]

Similarly, Eq. (7) gives \( \psi_{SApo}^{n+1} \):

\[
\psi_{SApo}^{n+1} = \psi_{SApo}^n + \delta \frac{t}{C_{SApo}} \left( K_{SLSym} \left( \psi_{SApo}^n - \psi_{SApo}^n \right) - K_{LSym} \left( \psi_{SApo}^n - \psi_{SApo}^n \right) \right) + \delta \frac{t}{C_{LSym}} \left( \psi_{SApo}^n - \psi_{SApo}^n \right).
\]

Equations. (8) and (9) give \( \psi_{LSym}^{n+1} \) and \( \psi_{TSym}^{n+1} \):

\[
\psi_{LSym}^{n+1} = \psi_{LSym}^n + \delta \frac{t}{C_{LSym}} \left( K_{LSym} \left( \psi_{LApo}^n - \psi_{SApo}^n \right) - E_{\text{stom}}^{n+1} \right),
\]

\[
\psi_{SSym}^{n+1} = \psi_{SSym}^n + \delta \frac{t}{C_{LSym}} \left( K_{LSym} \left( \psi_{SApo}^n - \psi_{SSym}^n \right) - E_{\text{cutis}}^{n+1} \right).
\]

C2 Implicit scheme

By combining Eqs. (6), (7), (8) and (9) with the implicit discretization (Eq. 50), it is possible to analytically compute the unknown water potentials of each compartment at time \( n + 1 \).

First, we eliminate \( \psi_{LSym}^{n+1} \) in Eqs. (6) and (8) by summing (6) + (8) \times \frac{K_{LSym}}{K_{LSym} + \frac{C_{LSym}}{\delta t} + \frac{E_{\text{stom}}^{n+1}}{2}} \) and re-organizing the result as follows:

\[
\frac{C_{LApo}}{\delta t} \left( \psi_{LApo}^{n+1} - \psi_{LApo}^n \right) + K_{SLSym} \left( \psi_{LApo}^{n+1} - \psi_{SApo}^n \right) + K_{LSym} \left( \psi_{SApo}^n - \psi_{LApo}^n \right) + \psi_{SApo}^n - \psi_{LApo}^n + \frac{\delta_{\text{cav}}^c}{C_{LSym}} \left( \psi_{SApo}^n - \psi_{SApo}^n \right) = 0,
\]

with \( \delta_{\text{cav}}^c = \begin{cases} 
0, & \text{if } \psi_{LApo}^{n+1} \geq \psi_{\text{mem}}^{SApo} \text{ (no new cavitation event)} \\
1, & \text{if } \psi_{LApo}^{n+1} < \psi_{\text{mem}}^{SApo} \text{ (new cavitation event)}
\end{cases}.

Next, let us define intermediate variables to ease the resolution with the following equations:

\[
\tilde{K}_L = \frac{C_{LApo}}{\delta t} + \frac{K_{LSym} \left( \frac{C_{LSym}}{\delta t} + \frac{E_{\text{stom}}^{n+1}}{2} \right)}{K_{LSym} + \frac{C_{LSym}}{\delta t} + \frac{E_{\text{stom}}^{n+1}}{2}} + \delta_{\text{cav}}^c K_{cav}^L,
\]

\[
\tilde{\psi}_L = \frac{C_{LApo} \frac{K_{LSym}}{\delta t} + \frac{K_{LSym} \left( \frac{C_{LSym}}{\delta t} + \frac{E_{\text{stom}}^{n+1}}{2} \right)}{K_{LSym} + \frac{C_{LSym}}{\delta t} + \frac{E_{\text{stom}}^{n+1}}{2}} + \delta_{\text{cav}}^c K_{cav}^L \psi_{LApo}^{n+1}}{K_L},
\]

\[
\tilde{E}_L = \frac{K_{LSym}}{K_{LSym} + \frac{C_{LSym}}{\delta t} + \frac{E_{\text{stom}}^{n+1}}{2}} \left( \frac{E_{\text{stom}}^{n+1} + E_{\text{cutis}}^{n+1}}{2} \right),
\]

Now, Eq. (C5) can be rewritten as follows:

\[
\tilde{K}_L \left( \psi_{LApo}^{n+1} - \tilde{\psi}_L \right) + K_{SLSym} \left( \psi_{LApo}^{n+1} - \psi_{SApo}^n \right) + \tilde{E}_L = 0.
\]
These equations can be combined to determine $\psi_{SApo}^{n+1}$, $\psi_{LSym}^{n+1}$, $\psi_{Lapo}^{n+1}$, and $\psi_{SSym}^{n+1}$.

We can now rearrange this to determine $\psi_{Lapo}^{n+1}$:

$$\psi_{Lapo}^{n+1} = \frac{K_{SLapo} \tilde{K}_L \psi_{SApo} + K_{SLapo} \tilde{K}_L \psi_{LSym} - \left[ E_L + \tilde{E}_L \right]}{K_{SLapo} + \tilde{K}_L},$$  \hspace{1cm} (C15)

Knowing $\psi_{SApo}^{n+1}$, we can determine $\psi_{SSym}^{n+1}$ from Eq. (C9):

$$\psi_{SSym}^{n+1} = \frac{K_{SSym} \psi_{SApo}^{n+1} + C_{SSym} \psi_{LSym}^{n+1} - E_{cutis}}{K_{SSym} + C_{SSym} \psi_{SApo}^{n+1}}.$$  \hspace{1cm} (C18)

C3 Semi-implicit scheme

Let

$$\delta_{L_{cav}} = \begin{cases} 0, & \text{if } \psi_{Lapo}^{n+1} > \psi_{mem}^{Lapo} \text{ (no cavitation event)} \\ 1, & \text{if } \psi_{Lapo}^{n+1} < \psi_{mem}^{Lapo} \text{ (cavitation event)} \end{cases},$$

$$\delta_{L_{cav}} = \begin{cases} 0, & \text{if } \psi_{SApo}^{n+1} \geq \psi_{SApo}^{cav} \text{ (no cavitation event)} \\ 1, & \text{if } \psi_{SApo}^{n+1} < \psi_{SApo}^{cav} \text{ (cavitation event)} \end{cases},$$

By combining Eqs. (6)–(9) with the semi-implicit Eq. (57), it leads to the following equations.

For $\psi_{Lapo}^{n+1}$:

$$\psi_{Lapo}^{n+1} = \alpha_{Lapo} \psi_{Lapo}^{n} + \left( 1 - \alpha_{Lapo} \right) \tilde{\psi}_{Lapo},$$

where

$$\alpha_{Lapo} = e^{-\frac{\psi_{Lapo}^{n} - \psi_{SApo}^{cav}}{\epsilon_{Lapo}}}.$$

and

$$\tilde{\psi}_{Lapo} = \frac{K_{SLapo} \psi_{SApo}^{n+1} + K_{LSym} \psi_{LSym}^{n+1} + \delta_{L_{cav}} \tilde{K}_L \psi_{Lapo}^{cav}}{K_{SLapo} + K_{LSym} + \delta_{L_{cav}} \tilde{K}_L \psi_{Lapo}^{cav}}.$$
the manuscript. All authors read and approved nary version of the code. FP designed the numerical resolutions of code and conducted the experiments. NMS developed a prelimi-
nary input from all authors. JR and NMS coordinated the project. HC

Author contributions. JR led the writing of the manuscript with
input from all authors. JR and NMS coordinated the project. HC and JLD supervised the project. JR, NMS and FP developed the code and conducted the experiments. NMS developed a preliminary version of the code. FP designed the numerical resolutions of the model with inputs from NMS. All authors read and approved the manuscript.

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