Do stomata optimize turgor-driven growth? A new framework for integrating stomata response with whole-plant hydraulics and carbon balance

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Summary

- Every existing optimal stomatal model uses photosynthetic carbon assimilation as a proxy for plant evolutionary fitness. However, assimilation and growth are often decoupled, making assimilation less ideal for representing fitness when optimizing stomatal conductance to water vapor and carbon dioxide. Instead, growth should be considered a closer proxy for fitness.
- We hypothesize stomata have evolved to maximize turgor-driven growth, instead of assimilation, over entire plants' lifetimes, improving their abilities to compete and reproduce. We develop a stomata model that dynamically maximizes whole-stem growth following principles from turgor-driven growth models. Stomata open to assimilate carbohydrates that supply growth and osmotically generate turgor, while stomata close to prevent losses of turgor and growth due to negative water potentials.
- In steady state, the growth optimization model captures realistic stomatal, growth, and carbohydrate responses to environmental cues, reconciles conflicting interpretations within existing stomatal optimization theories, and explains patterns of carbohydrate storage and xylem conductance observed during and after drought.
- Our growth optimization hypothesis introduces a new paradigm for stomatal optimization models, elevates the role of whole-plant carbon use and carbon storage in stomatal functioning, and has the potential to simultaneously predict gross productivity, net productivity, and plant mortality through a single, consistent modeling framework.

Introduction

Stomata regulate the terrestrial water and carbon cycles through transpiration and photosynthesis (Hetherington & Woodward, 2003), influencing climate through feedback with soil-moisture storage and runoff (Leipprand & Gerten, 2006; Betts et al., 2007), rainfall cycling (Van der Ent & Savenije, 2011; Zemp et al., 2017), atmospheric boundary layer development (Siqueira et al., 2009), atmospheric CO₂ concentrations (c_a), and air temperatures (T_a) (Betts et al., 2000; Cox et al., 2000) among others. Predictions of future climate traditionally employ empirical relationships between stomatal conductance (g_w) , photosynthetic carbon assimilation (A_n) , vapor-pressure deficit (VPD; D_L), and c_a (e.g. Ball et al., 1987; Leuning, 1995), lacking rigorous plant physiological mechanisms and requiring parameters with little physical meaning. While process-based models of stomatal guard cell dynamics exist (e.g. Buckley et al., 2003, 2012), they remain an open research frontier. Instead, optimization models are replacing empirical formulations for gw at large scales (Medlyn et al., 2011; De Kauwe et al., 2015; Eller et al., 2020; Sabot et al., 2020). These optimization models are based on the theory

distribution and reproduction in any medium, provided the original work is properly cited.

that plants have evolved through natural selection to maximize their evolutionary fitness within the limits of genotypic variation and physiological constraints (Mäkelä *et al.*, 2002; Franklin *et al.*, 2020).

Theoretically, optimization models should maximize a plant's reproductive success as the true measure of evolutionary fitness (Mäkelä et al., 2002; Table 1). In practice, however, optimization models for g_w traditionally hypothesize stomata maximize A_n as a proxy for fitness (e.g. Cowan, 1977; Cowan & Farquhar, 1977), which we call assimilation optimization hypotheses (AOHs; see Table 2 for terminology). Maximizing A_n instantaneously would unrealistically predict stomata always maximally open (unless including nonstomatal limitations to photosynthesis). Thus, AOHs introduce additional costs to penalize opening and to predict realistic gw (Wolf et al., 2016; Wang et al., 2020). These costs and constraints have been associated with water-saving strategies to avoid future water stress (solved either over short durations: Cowan, 1977; Cowan & Farquhar, 1977; Hari et al., 1986; Katul et al., 2010; Medlyn et al., 2011; or dynamically: Cowan, 1982; Mäkelä et al., 1996; Manzoni et al., 2013; Lu et al., 2016; Mrad et al., 2019), xylem embolism (e.g.

Table 1 Proxies for evolutionary fitness in stomatal optimization.

Rank (from best to worst)	Plant processes/ proxy for evolutionary fitness	Could this process be satisfactorily modeled at the genesis of stomatal optimization (i.e. Cowan, 1977; Cowan & Farquhar, 1977)?	Can this process be satisfactorily modeled now?	Selected studies of how to model these processes	Selected studies that apply this proxy for fitness to model stomatal conductance
1 2	Reproduction Growth (G)	× ×¹	× ,		– Present study
3	Photosynthetic net carbon assimilation (A _n)		*	Farquhar et al. (1980); Farquhar & Wong (1984); Collatz et al. (1991); Evans & Farquhar (1991); De Pury & Farquhar (1997)	Cowan & Farquhar (1977); Cowan (1982); Givnish (1986); Hari et al. (1986); Mäkelä et al. (1996); Katul et al. (2010); Medlyn et al. (2011); Manzoni et al. (2013); Prentice et al. (2014); Lu et al. (2016, 2020); Wolf et al. (2016); Hölttä et al. (2017); Sperry et al. (2017); Anderegg et al. (2018); Dewar et al. (2018, 2022); Eller et al. (2018); Bartlett et al. (2019); Deans et al. (2020); Wang et al. (2020)

A historic inability to satisfactorily mathematically model better proxies of evolutionary fitness, such as plant reproduction and growth, leads to an emphasis on plant processes that could be modeled at the genesis of stomatal optimization, specifically photosynthetic net carbon assimilation. Recent improvements in plant growth modeling may now advance stomatal optimization.

¹Early work did not consider the role of nonstructural carbohydrate (NSC) storage and thus assumed that an increase in photosynthetic carbon assimilation will always increase growth, which is not necessarily true. Hence, the choice of photosynthetic carbon assimilation as the proxy for fitness was intended to reflect growth optimization.

Wolf et al., 2016; Sperry et al., 2017), and nonstomatal limitations to photosynthesis (decreased mesophyll conductance and/or photosynthetic capacities; e.g. Givnish, 1986; Dewar et al., 2018). Besides nonstomatal limitations (Salmon et al., 2020), these costs cannot be measured directly. It is possible to infer them from gas-exchange experiments. However, these experiments are difficult and rarely performed (Hall & Schulze, 1980; Fites & Teskey, 1988; Thomas et al., 1999). Overall, AOHs emphasize carbon acquisition, but few AOHs consider how plants actually use carbon once assimilated (Nikinmaa et al., 2013; Hölttä et al., 2017; Dewar et al., 2022). Plant survival, competition, and reproduction should be better reflected by carbon use than by assimilation.

Few AOHs explicitly discuss growth (Givnish & Vermeij, 1976; Cowan, 1982), an important aspect of carbon use. These studies, however, incorrectly assume assimilation and growth are coupled. Assimilation and growth are certainly correlated over annual or longer timescales (Von Allmen *et al.*, 2012; Smith & Sperry, 2014) but decoupled over shorter timescales (Cabon *et al.*, 2022). Over diurnal timescales, photosynthesis occurs during the day, while most radial growth typically occurs at night or early morning (Peters *et al.*, 2021; Zweifel *et al.*, 2021). The balance among photosynthesis, growth, and respiration governs the storage of carbon as nonstructural carbohydrates (NSCs), which accumulate during the day (when assimilation exceeds demand)

and deplete at night (when demand exceeds assimilation; Smith & Stitt, 2007). Similar NSC patterns occur over seasonal time-scales. Trees' NSCs often peak before or at the onset of the growing season (Martínez-Vilalta *et al.*, 2016), storing them throughout the winter and sourcing the sprouting of new stems in the following year (Epron *et al.*, 2012). In fact, trees store a seeming abundance of NSCs, enough to potentially rebuild their canopies more than four times (Hoch *et al.*, 2003). Nonetheless, a consensus on the degree to which growth is limited by NSCs is yet to be reached (Sala *et al.*, 2012; Wiley & Helliker, 2012). Are stomatal responses coordinated with growth and NSCs? Could such a potential coordination optimize plants' fitness? AOHs lack a framework to investigate these questions.

We hypothesize stomata have evolved strategies that maximize stem growth over trees' entire lifetimes, which we refer to as the growth optimization hypothesis (GOH). Integrated over time, maximizing growth further maximizes size (height, rooting depth), which directly reflects their abilities to compete for resources (light, water, nutrients; King, 1990; Franklin, 2007). In fact, maximizing height growth is an evolutionarily stable strategy for trees competing for light (Mäkelä, 1985; King, 1990). Size also impacts reproductive success, since most plants reproduce only after achieving a minimum size (Obeso, 2002). Any optimization requires a tradeoff. According to the GOH, stomata open to assimilate carbon to supply growth (i.e. the benefit); however, opening reduces water

Table 2 Glossary of terminology.

Term	Meaning
Assimilation optimization hypothesis (AOH)	A hypothesis that states that stomata optimize a trade-off between photosynthetic net carbon assimilation (A _n) and some additional cost that penalizes excessive stomatal opening
Growth optimization hypothesis (GOH)	The hypothesis that stomata optimize turgor-driven growth
Growth-optimizing stomata model (GOSM)	A model that predicts the stomatal conductance (g_c) that optimizes cambial growth
Nonstructural carbohydrates (NSCs)	Large organic macromolecules that provide both the material and energy required for biological chemical reactions, the synthesis of other organic compounds, and the growth of new biomass. Internal plant storage of NSCs buffers the asynchrony of supply and demand over diurnal, seasonal, and decadal timescales and across plant organs
NSC-use efficiency (NSCUE)	Cost of depleting NSCs that is associated with the gain of growing new biomass
Water-use efficiency (WUE)	Cost of evaporative water loss that is associated with photosynthetic carbon gain
Instantaneous optimal solution	Solutions of g_c , marginal WUE (λ), and growth rate (G) that optimize growth given prescribed values of the current NSC storage (C) and marginal NSCUE (η)
Steady-state optimal solution	Solutions of g_c , λ , G , and C that optimize growth once both C and η cease changing ($dC/dt = 0$ and $d\eta/dt = 0$) under constant environmental conditions. Steady-state solutions represent the final product of the feedback between C and η once they have ceased changing under constant environmental conditions
Transient optimal solution	Dynamic solutions of g_c , λ , G , and C that optimize growth over a specified duration, which we refer to as the <i>timescale</i> of optimization. Transient solutions allow C and η to dynamically feedback and evolve in coordination with environmental conditions throughout the timescale of optimization. Environmental conditions may be constant or vary during the timescale of optimization in transient solutions
Long-term acclimation of leaf traits to eCO_2	Any solutions under elevated atmospheric CO ₂ concentrations (eCO ₂) that reflect leaf-level adaptations to eCO ₂ that arise slowly over seasonal to annual timescales, including increased leaf areas and reduced photosynthetic capacities. Relevant to Supporting Information Figs S13 and S14
Acclimation to permanent soil water stress	Steady-state optimal solution to permanent soil water stress that reflects a plant's long-term ability to survive and function in different environments of differing soil water availability. Relevant to Figs 4 and 5
Recovery after transient drought stress	Steady-state optimal solution with plentiful soil water and with reduced xylem conductance due to a past transient period of intense soil water stress. This past drought is assumed to have been long enough to reduce xylem conductance while short enough for the simulated plant to still survive. These simulations represent <i>drought legacy</i> effects to varying drought intensities. Relevant to Figs 4 and 5

potentials (ψ) throughout the plant, turgor pressure (P) in the cambium and inner bark, and thus growth (i.e. the cost; Figs 1, 2; Sperry et al., 1998; Hölttä et al., 2010). Turgor is an established physiological control on plant growth, including cell expansion (Lockhart, 1965) and division (Kirkham et al., 1972). Turgor also drives the transport of the building blocks of structural biomass (carbohydrates) from where they are produced (leaves) or stored through the phloem to growth sites (Münch, 1930). Additionally, permanent turgor loss leads to mortality of individual cells (i.e. protoplast detachment from the cell wall) and whole plants (through tissue water content loss; Sapes et al., 2019; Preisler et al., 2021; Sapes & Sala, 2021). In fact, an inability to grow often precedes death (Das et al., 2016; Cailleret et al., 2017; DeSoto et al., 2020), potentially due to turgor loss. By maximizing growth and turgor and meristem water content, the GOH reflects a strategy that minimizes desiccation-induced mortality (Supporting Information Notes \$1; Section \$2).

In this paper, we develop a growth-optimizing stomata model (GOSM) following our GOH that dynamically maximizes the whole-stem growth of an individual tree. We model growth following turgor-driven stem growth models (e.g. Steppe *et al.*, 2006; De Schepper & Steppe, 2010) and source-sink models that consider NSC limitations (e.g. Schiestl-Aalto *et al.*, 2015; Hayat *et al.*, 2017). We compare our GOSM to AOH models as a first-order test. We show the GOH agrees with our current understanding of stomatal regulation to environmental forcing. Finally,

we explore our GOH's potential to improve our understanding and ability to predict responses of stomata, growth, and NSC storage, particularly during and after drought.

Materials and Methods

Growth-optimizing stomata model

This section summarizes the core assumptions, structure, and key variables of our GOSM (Fig. 1; Table S1). We do not use GOH and GOSM interchangeably. Growth optimization hypothesis refers to a general modeling framework, while GOSM refers to the specific solution to the GOH presented here. Our GOH assumes only that stomata maximize growth, while our GOSM makes many other assumptions to reach a tractable solution. Growth-optimizing stomata model details and assumptions are fully explained in the Supporting Information (Notes S1; Fig. S1), including its limitations (Section S10). The GOH's overarching principle is stomata maximize an individual tree's whole-stem growth over its lifetime. Mathematically, this principle is stated as:

$$\max_{g_c} \int G(C, E) \, \mathrm{d}t$$
 Eqn 1

where g_c is the total conductance to CO_2 , t is time, and G is the whole-tree growth rate expressed in carbon equivalents per unit

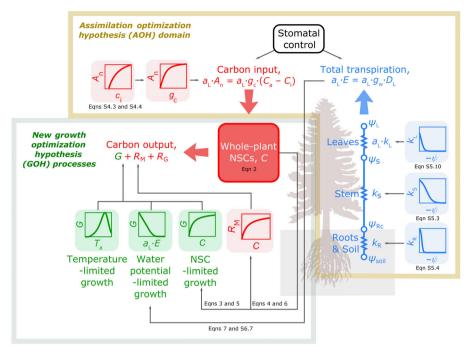


Fig. 1 Conceptual diagram for our growth-optimizing stomata model (GOSM; see Table S1 for symbols used). Stomata optimize growth by balancing a trade-off between accumulation of nonstructural carbohydrates (NSCs, C; red box) and preventing excessively negative water potentials that hamper cell expansion and division. Stomata open to accumulate NSCs through total photosynthetic carbon assimilation (a_L · A_n , where a_L is the total leaf area, and A_n is the leaf area-specific carbon assimilation). These NSCs are required to osmotically generate turgor and supply growth (G; NSC-limited growth), maintenance respiration (R_M), and construction respiration (R_G , where $R_G \propto G$). NSCs accumulate when carbon inputs (a_L · A_n) exceed carbon outputs ($R_M + R_G + G$) and deplete when outputs exceed inputs. In steady state, carbon inputs and outputs are equal, and NSC reserves are stable. Meanwhile, stomata close to prevent cambial water potentials from reaching a threshold, past which growth ceases, by limiting total transpiration (a_L ·E, where E is the leaf area-specific transpiration; water potential-limited growth). The total transpiration is conducted through soil, roots, stems, and then leaves, which are given by their conductances (k_R , k_S , and a_L · k_L , respectively, where k_R is the conductance of the soil and roots in series, and k_L is the leaf conductance per unit leaf area), and soil, root collar xylem, stem apex xylem, and leaf xylem potentials (ψ_{soil} , ψ_{RC} , ψ_S , and ψ_L , respectively). These conductances decline as a result of xylem embolism and loss of soil conductance due to the unsaturation of soil void space (blue subplots on right). Additionally, growth may be limited by cold or hot temperatures (temperature-limited growth). Through the diagram, processes are distinguished by color. Red reflects carbon and water, while thin black arrows represent mathematical relationships between model variables.

time. G depends on other state variables and environmental conditions. The state variables are the leaf area-specific transpiration rate (E) and the NSC storage (C), while environmental conditions are soil water potential (ψ_{soil}) , air temperature (T_{a}) , the air's relative humidity (RH), and insolation (I_{s}) .

We first examine the steady-state GOSM solution to show there exists a g_c that maximizes G (Fig. 2) for the environmental conditions used in Table 3 and later mathematically define the steady-state solution (Eqns 10, 11). Opening stomata releases water vapor as transpiration (E), reducing leaf temperature ($T_{\rm L}$) and leaf-to-air VPD (D_L) . Opening stomata enables photosynthetic assimilation of CO_2 (A_n), which depends on T_L by the kinetics of enzymes involved in photosynthesis. Hence, opening stomata supplies carbon for growth. Meanwhile, opening 'pulls' xylem sap upward, reducing plant water potentials (ψ_L , ψ_S , and ψ_{Rc} in Fig. 2), turgor (P), and the carbon demand for growth (G_0 in Fig. 2). As stomata close, supply diminishes, while demand raises, shrinking carbon storage in the form of NSCs (C). Conversely as stomata open, *supply* increases, and *demand* decreases, raising NSCs. In steady state, the actual growth (G) balances carbon *supply* with *demand*, and *G* peaks at an intermediate stomatal

aperture (G_{max} in Fig. 2). These G_0 , G, C, A_n , E, T_L , D_L , ψ_L , ψ_S , and ψ_{Rc} curves shift as environmental conditions vary, repositioning the optimal g_c (Fig. S2). For example, under soil drought, water potential curves decline, reducing G₀ and G curves, elevating the C curve, and shifting G_{max} to a smaller conductance (Fig. S2), thereby changing g_c through changes in *demand* curves. Similarly, elevated atmospheric CO₂ concentration shifts G_{max} to a smaller conductance, however, through changes in *supply* curves by elevating A_n , C, and G curves (Fig. S2). Reducing RH shifts G_{max} to a smaller conductance through changes in the *demand* (G_0) curve, particularly by increasing its sensitivity to g_c (larger $\partial G_0/\partial g_c$ due to higher VPD). Increasing air temperatures has a small effect on the *supply* (A_n) curve; however, changes in *demand* are dominant, closing stomata by increasing the sensitivity of G_0 to g_c (i.e. a VPD response) and increasing the maximum value of G_0 when stomata are fully closed (a direct growth stimulation by warmer temperatures; Figs S2, S3).

We apply the calculus of variations (Witelski & Bowen, 2015) to solve Eqn 1 for the optimal g_c . Several AOH studies have applied the calculus of variations to investigate soil water-saving strategies (Cowan, 1982; Mäkelä *et al.*, 1996; Manzoni *et al.*, 2013;

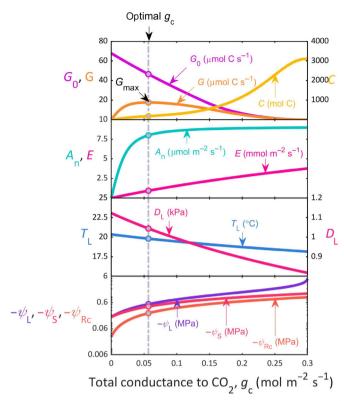


Fig. 2 Stomatal response model based on the steady-state optimization of turgor-limited growth, showing predictions for photosynthetic C assimilation (A_n , teal line), transpiration (E, pink line), leaf temperature (T_L , blue line), leaf-to-air vapor-pressure deficit (VPD; D_I, magenta line), leaf, stem, and root collar water potentials $(-\psi_1, -\psi_S, \text{ and } -\psi_{RC}, \text{ and, dark purple,}$ red, and dark orange lines, respectively), potential growth rate (G_0 , light purple line), actualized growth rate (G, light orange line), and nonstructural carbohydrate (NSC) storage (C, yellow line) for potential values of total conductance, gc. Here, we present results from the steady-state version of our model for which both NSC storage and the NSC-use efficiency (NSCUE, η) are constant (C = 0 and $\dot{\eta} = 0$) for the environmental conditions and physiological parameters listed in Table 3. The optimal g_c occurs where G is maximum (G_{max} , vertical dot-dashed gray line) and coincides with optimal values of A_n , E, T_L , D_L , ψ_L , ψ_S , ψ_{RC} , G_0 , and C (circles). The upper limit of the x-axis ($g_c = 0.3 \text{ mol m}^{-2} \text{ s}^{-1}$) approximately coincides with the maximum potential E, above which ψ_1 rapidly approaches negative infinity; thus, the x-axis represents the hydraulically viable range of potential g_{cr} which are not prone to runaway embolism.

Mrad *et al.*, 2019). Here, we define whole-plant NSC storage (*C*) as the limiting dynamic resource that trees must conserve to support future growth and respiration. To apply the calculus of variations, we must define the dynamics of *C*:

$$\dot{C} = \frac{\mathrm{d}C}{\mathrm{d}t} = a_{\mathrm{L}}A_{\mathrm{n}} - R_{\mathrm{M}} - R_{\mathrm{G}} - G$$
 Eqn 2

where a_L is the total leaf area, A_n is the leaf area-specific average carbon assimilation rate, R_M is the maintenance respiration of stems and roots, and R_G is the whole-tree growth respiration (Eqns S1.1, S1.2). A_n is modeled following a big-leaf version of the Farquhar *et al.* (1980) model, based on electron transport and carboxylation capacities and their dependence on leaf temperature, the calculation of which considers evaporative leaf

cooling (Notes S1; Sections S3, S4). Future GOSM iterations should consider nonstomatal limitations to photosynthesis based on their compatibility with traditional AOH theories (Zhou *et al.*, 2013; Novick *et al.*, 2016).

 $R_{\rm G}$ is proportional to G such that their sum equals $G/(1-f_{\rm c})$, where $f_{\rm c}$ is a constant less than one, equaling $R_{\rm G}/(a_{\rm L}\cdot A_{\rm n}-R_{\rm M})$ when C=0 (Eqn S1.6). Both G and $R_{\rm M}$ are potentially limited by NSCs (Thornley, 1970, 1971; Eqns S1.3, S1.7; $R_{\rm G}$ may also be NSC-limited, since $R_{\rm G}\propto G$):

$$G = \sigma_{g}(C)G_{0}$$
 Eqn 3

$$R_{\rm M} = \sigma_{\rm r}(C)R_{\rm M,0}$$
 Eqn 4

where G_0 and $R_{\rm M,0}$ are maximum potential values of G and $R_{\rm M}$ if NSCs were nonlimiting, and $\sigma_{\rm g}$ and $\sigma_{\rm r}$ are fractions that span between zero (when C is small) and unity (when C is large). We model $\sigma_{\rm g}$ and $\sigma_{\rm r}$ similarly to Jones *et al.* (2020) (Eqns S1.4, S1.8):

$$\sigma_{\rm g}(C) = \frac{C}{C + \gamma_{\rm o} C_{\rm struct}}$$
 Eqn 5

$$\sigma_{\rm r}(C) = \frac{C}{C + \gamma_{\rm r} C_{\rm struct}}$$
 Eqn 6

where C_{struct} is the total dry biomass in carbon equivalents, and $\gamma_{\rm g}$ and $\gamma_{\rm r}$ are unitless parameters. We include $C_{\rm struct}$ in Eqns 5 and 6 to signify tree size should influence potential NSC limitations and so $\sigma_{\rm g}$ and $\sigma_{\rm r}$ can be related to NSC concentrations rather than whole-tree NSC reserves and are thus presumably less sensitive to tree size. For simplicity, we here model a large, mature tree, enabling us to treat C_{struct} as effectively constant throughout simulations. However, C_{struct} would change dynamically due to G in a full dynamic growth model, especially for young, small trees. This simplification is reasonable if simulations are short or if relative growth rates are small, the latter of which holds for trees that are mature in stature (Valentine & Mäkelä, 2012; Forrester, 2021). Thus, in this study, $\gamma_g \cdot C_{\text{struct}}$ and $\gamma_{\rm r} \cdot C_{\rm struct}$ are more important quantities than $\gamma_{\rm g}$ and $\gamma_{\rm r}$, respectively, both of which we indirectly estimated for model performance given our value of C_{struct} (see 'Parameterization' in the Materials and Methods section; Notes \$1; Section \$8).

 $R_{\rm M,0}$ is temperature-dependent and modeled by Q₁₀-type equations (Eqn S1.5). G_0 is modeled following Potkay *et al.* (2021a), including growth limitations imposed by turgor (Lockhart, 1965; Steppe *et al.*, 2006) and temperature (Parent *et al.*, 2010; Cabon *et al.*, 2020). Their formulation is equivalent to the Lockhart (1965) equation for turgor-driven cell expansion integrated axially along the whole stem within an allometric configuration:

$$G_0 = \widetilde{\phi} \frac{C_{\mathbf{W}}}{u_{\mathbf{S}}} \int_0^1 \max(P_0(\widetilde{z}) - \Gamma, 0) d\widetilde{z}$$
 Eqn 7

where ϕ is an 'effective' whole-tree extensibility that is proportional to the true cell wall extensibility (Potkay *et al.*, 2021a), $C_{\rm W}$

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 Table 3
 Default environmental conditions and physiological parameters for our growth-optimizing stomata model (GOSM) and their source.

Meaning

Value

Symbol

Source

															Estimated assuming a leaf area index of 0.4 (LAI = $a_1.\varphi^{-1}.W^{-2}$) based on Scots pine trees at Poblet nature reserve (Prades Mountains, northeast Spain; Poyatos $et aI$,, 2013); we further assume a ratio of total to projected leaf area, ω , equal to 3.34 for conical pines (Buckley & Roberts, 2006a)	Estimated from C _w assuming a root-to-stem mass ratio of c. 0.3 based on Poorter <i>et al.</i> (2012) for the given tree size; leaf biomass was estimated assuming a specific leaf area of 0.08 m ² (mol C) ⁻¹ based on Marfinez-Vilalta <i>et al.</i> (2009) for Scots pine and that needle dry matter of Scots pine is 50% carbon based on Janssens <i>et al.</i> (1999)	Estimated from H based on allometric scaling equations with parameters from Potkay et al. (2021b) for Scots pine; allometric equation assumes an empirical form factor of 0.5 (Schiestl-Aalto et al. 2015) and a stem carbon density of 1.4×10^4 mol C m ⁻³ (Buckley & Roberts 2006a)	Estimated from H based on allometric scaling equations with parameters from Potkay et al. (2021b) for Socis pine; allometric equation is originally based on data from Mencuccini & Grace (1996); Hari & Kulmala (2005); and Rever et al. (2020)	Representative height of Scots pine trees at Poblet nature reserve (Prades Mountains, northeast	Spain; Poyatos <i>et al.</i> ., 2013) Estimated from <i>H</i> based on allometric scaling equations with parameters from Potkay <i>et al.</i> (2021b)	for Scots pine; allometric equation is originally based on data from Pretzsch (2014)	Approximate mean rooung depth for pines from Fan et al. (2017)	Based on Kolari <i>et al.</i> (2014) for Scots pine, the temperature dependence of the maximum rate of electron transport. Jmar. is modeled following Harley & Baldocchi (1995)	Hölttä <i>et al.</i> (2017) for Scots pine	Hölttä <i>et al.</i> (2017) for Scots pine	Based on Kolari <i>et al.</i> (2014) for Scots pine, the temperature dependence of the maximum carboxylation rate, V _{c,max} , is modeled following Harley & Baldocchi (1995)
Specific heat capacity of dry air at	constant pressure Acceleration due to gravity	Universal gas constant	Density of water	Stefan–Boltzmann constant	Atmospheric CO ₂ concentration	expressed as a mole fraction	Boundary layer conductance to vapor	Atmospharic O. concentration	expressed as a mole fraction	Atmospheric pressure	Atmospheric relative humidity	Air temperature	Solar elevation below the zenith	Soil water potential	Leaf area	Whole-tree structural biomass in carbon equivalents	Total aboveground stem (wood) biomass in carbon equivalents	Diameter at breast height	Stem height	Canopy width		Kooting depth	Maximum rate of electron transport at 17°C	Michaelis-Menten constant for	carboxylation Michaelis—Menten constant for	oxygenation Maximum carboxylation rate at 17°C
J mol ⁻¹ K ⁻¹	m s ⁻²	J mol ⁻¹ K ⁻¹		W m ⁻² K ⁻⁴	mol mol ⁻¹	,	mol m ⁻² s ⁻¹	wo III	5	kPa	I	Ç	rad	MPa	m ²	mol C	mol C	ш	Ш	E	;	m osynthesis	mol m ⁻² s ⁻¹	mol mol ⁻¹	mol mol ⁻¹	mol m ⁻² s ⁻¹
stants 29.2	28	8.314	866	σ $5.67 imes10^{-8}$ V Default environmental drivers	4.10×10^{-4}		2.4	$\frac{500}{207} \times 10^{-1}$	\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \	101.325	0.4	25	0 (0 netrics	6.9	1.13×10^3	8.3×10^{2}	9.2	41	6.1	ſ	ے Light interception and photosynthesis	1.1×10^{-4}	2.75×10^{-4}	4.20×10^{-1}	6.1×10^{-5}
Physical constants C _P 29.2	b	s ex	б	σ Default envi	Ca		Sp '	~ c	o g	$ ho_{ m atm}$	RH	\mathcal{T}_{a}	Φ	ψ_{soil} 0 Structural metrics	a_{L}	Cstruct	O _w	Q	Н	8	1	ے Light interce	Jmax,17°C	$K_{\rm c}$	×°	V _{c,} max,17°C

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Symbol	Value	Units	Meaning	Source
* &	3.60×10^{-5} mol mol ⁻¹ 0.97	mol mol ⁻¹ -	CO ₂ compensation point Leaf emissivity	
θ_{c}	0.98	ı	nyperbolic minimum of tion-limited assimilation	Sperry <i>et al.</i> (2017)
			rate, A_c , and the electron transport-limited assimilation rate, A_j ;	
$\theta_{ m j}$	06.0	I	$ heta_{c}A_{n}^{2}-(A_{c}+A_{j})A_{n}+A_{c}A_{j}=0$ Coefficient in hyperbolic minimum of	Sperry <i>et al.</i> (2017)
			maximum electron transport rate,	
			electron transport, J_1 ; $\theta_1 J^2 - (J_{\text{max}} + I_{\text{max}} + I_{\text{max}})$	
×	$6.90 \times 10^{-7} \text{ mol J}^{-1}$	mol J ⁻¹	nstant between the	Calculated for a quantum yield of 0.3 (mol photon).(mol e ⁻) ⁻¹ and an average PAR wavelength of
			light-limited rate of electron transport, <i>I</i> 1, and absorbed radiation,	550 nm
			$l; J_1 = \kappa \cdot l$	
κL	0.5	$m^{2} m^{-2}$	Canopy light extinction coefficient	Murty et al. (1996); Buckley & Roberts (2006a) for lodgepole pine
Whole-tree carbon use	carbon use			
$f_{\rm c}$	0.28	1	Fixed fraction of GPP diverted to	Chung & Barnes (1977) for loblolly pine
			growth respiration, $R_{ m G}$	a a
RM,0,25°C	$R_{\text{M,0,25}^{\circ}\text{C}} = 5.0 \times 10^{-5} \text{mol C s}^{-1}$	mol C s ⁻¹	: maintenance	For various combinations of γ_g and γ_r , we calculated $R_{M,0,25^{\circ}C}$ (Fig. S4), $\phi_{25^{\circ}C}$ (Fig. S5), and the
			respiration rate at 25°C	steady-state NSC reserve (C, Fig. 56) that would satisfy two constraints:

The target value for $R_{\rm M}$ was determined given:

et al., 2020) (NPP: GPP = $G/(R_M + G/(1 - f_c))$) in steady state at the default environmental condi-

mol C s⁻¹ at the default environmental conditions (i.e. values of c_a , l_s , o_a , ρ_{atm} , RH, T_a , Φ , and ψ_{soil} • the net-to-gross primary productivity (NPP: GPP) ratio equals 0.45 (Waring et al., 1998; Collalti

from this table)

ullet the stem and root maintenance respiration rate, R_{M} , equals a target value of 1.5 imes 10⁻⁵

• a carbon pool-specific root maintenance respiration rate at 15°C, r_{M,R,15°C}, equal to

 $3.1 \times 10^{-8} \, \text{s}^{-1}$ based on Schiestl-Aalto et al. (2015) for Scots pine

 $\bullet \;\; a$ temperature dependence for root maintenance respiration following Q_{10} relationship with

 $Q_{10} = 1.98$ based on Marshall & Waring (1985) for Scots pine

• a carbon pool-specific stem maintenance respiration rate at 15°C, r_{M,W,15°C}, equal to 6.6×10^{-11} s⁻¹ based on Schiestl-Aalto et al. (2015) for Scots pine

 $\bullet \;$ a temperature dependence for stem maintenance respiration following Q_{10} relationship with $Q_{10} = 1.8$ based on Zha et al. (2004) for Scots pine

Given the calculated \overline{C} at ambient and elevated atmospheric CO₂ concentrations for various combinations of γ_g and γ_r (Figs S6, S7), we chose values of γ_g , γ_r , and their corresponding • allometric relationships for Scots pine's carbon pools given by Potkay et al. (2021b)

• $\overline{C} \approx 175 \text{ mol under ambient atmospheric CO}_2 \text{ concentrations, based on Schiestl-Aalto}$ R_{M,0,25°C}, φ_{25°C} values that would satisfy two criteria:

• the percent increase in \overline{C} due to elevated atmospheric CO_2 concentrations was >0% but did not exceed c. 20%, based on Li et al. (2018) et al. (2019)

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Tab

Symbol	Value	Units	Meaning	Source
Ys	0.26	ı	Shape parameter for $\sigma_{\rm g}$, the NSC substrate limitation function for whole-tree growth G	
γr	0.38	I	Shape parameter for σ_n , the NSC substrate limitation function for stem and root maintenance respiration, $R_{\rm M}$	
Turgor-lim	Turgor-limited growth		-	
C _{m,1}	0.48	$^{ m mol}{ m kg}^{-1}$	Intercept of the empirical relationship between phloem sap molality, $m_{\rm p}$, and stem water potential, $\frac{1}{W_{\rm S}}$	Paljakka e <i>t al</i> . (2017) for Scots pine
C _{m,2}	0.13	mol kg ⁻¹ MPa ⁻¹	$(m_p = C_{m,1} - C_{m,2} \cdot \psi_s)$ Slope (negative) of the empirical relationship between phloem sap	Paljakka <i>et al.</i> (2017) for Scots pine
Сп,1	866.0	ı	molality, $m_{\rm p}$, and stem water potential, $\psi_{\rm S}$ ($m_{\rm p} = C_{\rm m,1} - C_{\rm m,2} \cdot \psi_{\rm S}$) Linear coefficient between osmotic potential, π , and phloem sap molality,	$(\pi=-\overline{\omega}\cdot ho\cdot R\cdot T_{ m a}\cdot\left(c_{\Pi,1}\cdot m_{ m p}+c_{\Pi,2}\cdot m_{ m p}^2 ight)$, where $\overline{\omega}=10^{-6}$ MPa Pa $^{-1}$)
		Michel (1972); Thompson & Holbrook (2003a) for	т _р	
Сп,2	0.089	kg mol ⁻¹	Quadratic coefficient between osmotic potential, π , and phloem sap molality, $m_{\rm p}$	$(\pi=-\overline{\omega}\cdot ho\cdot R\cdot T_{\rm a}\cdot\left(c_{\Pi,1}\cdot m_{ m p}+c_{\Pi,2}\cdot m_{ m p}^2)$, where $\overline{\omega}=10^{-6}{ m MPa}{ m Pa}^{-1})$
		Michel (1972); Thompson & Holbrook (2003a) for sucrose	ı.	
US	0.25	ı	Fraction of new growth allocated to stems	Xia <i>et al</i> . (2019); Potkay <i>et al</i> . (2021b) for mature tree
Ĺ	0.75	МРа	Threshold that turgor, P, must exceed for stem expansion	Potkay <i>et al.</i> (2021a)
$\widetilde{\phi}_{25^{\circ}\text{C}}$	4.6 × 10 ⁻⁸	MPa ⁻¹ s ⁻¹	Effective whole-tree extensibility at 25°C	See the Source section in this table for $R_{M,0.25^{\circ}C}$, γ_{g_i} and γ_r parameters. The temperature dependence of the effective whole-tree extensibility, $\tilde{\phi}$, is modeled following an equation similar to Johnson et al. (1942) and Parent et al. (2010) using the parameters reported by Cabon et al. (2020) and Peters et al. (2021). Unlike Johnson et al. (1942) and Parent et al. (2010), $\tilde{\phi}$ here continuously approaches zero as temperatures approach $\tilde{5}^{\circ}$ C, the growth threshold suggested by Könner (2008) (Fig. S3)
Xylem hydraulics k∟ 1.6	raulics 1.6 × 10 ⁻²	mol H ₂ O m ⁻² s ⁻¹ MPa ⁻¹	Maximum leaf xylem conductance per leaf area	Buckley & Roberts (2006a) for loblolly pine

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Symbol	Value	Units	Meaning	Source
, k	2.30 × 10 ⁻²	mol s ⁻¹ MPa ⁻¹	Maximum soil-root conductance	Estimated given the values of H , Z , and a_L from this table (H and a_L are based on Poyatos $etaI_L$, 2013) assuming: • belowground resistance is 45% of total resistance based on Martínez-Vilalta $etaI_L$ (2007) for Scots pine • a midday leaf area-specific transpiration, E , equal to 1.6×10^{-3} mol m ⁻² s ⁻¹ coincides with predawn and midday leaf water potentials, ψ_{pd} and ψ_{md} , equal to -0.72 and -1.50 MPa, respectively, based on Poyatos $etaI_L$ (2013) • roots operate at 80% of their maximum conductance based on Lintunen $etaI_L$ (2020) for Scots
ks	9.76×10^{-2}	mol s ⁻¹ MPa ⁻¹	Maximum whole-stem xylem conductance	pane Estimated from H based on allometric scaling equations with parameters from Potkay et al. (2021b) for Scots pine; allometric equation is originally based on data from Mencuccini & Grace (1996) and Hölttä et al. (2013); we further assumed in this calculation that the sapwood fraction of the total wood is 94% for Scots pine based on Mencircini & Grace (1996)
α^{Γ}	5: 1	MPa ⁻¹	Shape parameter for describing the loss of leaf conductance under increasingly negative water notentials	Estimated by fitting exponential sigmoidal curves to needle conductance measurements by Domec et al. (2009) for loblolly pine
α_{R}	3.5	MPa ⁻¹	Shape parameter for describing the loss of belowground conductance under increasingly negative water	Estimated by fitting exponential sigmoidal curves to belowground conductance measurements by Poyatos <i>et al.</i> (2018) for Scots pine
α_{S}	0.8	MPa ⁻¹	Shape parameter for describing the loss of stem conductance under inconsistent and an activity in the conductance under the conductance under the conductance under the conductance and th	Estimated by fitting exponential sigmoidal curves to branch conductance measurements by Torres-Ruiz $etal.$ (2016) for Scots pine
eta_{L}	-0.75	МРа	Increasingly regauve water poternias Water potential at which half of the leaf conductance is lost (i.e. P50 for	Estimated by fitting exponential sigmoidal curves to needle conductance measurements by Domec $etal.(2009)$ for loblolly pine
$eta_{ t R}$	1.1	МРа	leaves) Water potential at which half of the Belowground conductance is lost (i.e. P50 for the root-cal exctem)	Estimated by fitting exponential sigmoidal curves to belowground conductance measurements by Poyatos <i>et al.</i> (2018) for Scots pine
$eta_{ m S}$	-3.3	МРа	Water potential at which half of the stem conductance is lost (i.e. P50 for stems)	Estimated by fitting exponential sigmoidal curves to branch conductance measurements by Torres- Ruiz <i>et al.</i> (2016) for Scots pine
Parameters μ _w	Parameters for other models $\mu_{\rm w}$ 1 × 10 ⁻³	s mol mol ^{–1}	Constant marginal carbon cost of water; used in the Cowan &	Chosen to agree with the other models under the default environmental conditions
β	145	I	Farquhar (1977) model Ratio of costs associated with carboxylation capacity, V _{c,max} , and transpiration, E, used in the Prentice	Smith <i>et al.</i> (2019); Lavergne <i>et al.</i> (2020b)
β1	4 × 10 ⁻⁶	mol m ⁻² s ⁻¹ MPa ⁻²	et al. (2014) model Quadratic coefficient between hydraulic risk, Θ , and leaf water potential, ψ_L ; used in the Anderegg et al. (2018) model; $\Theta = \beta_1 \cdot \psi_L^2 + \beta_2 \cdot \psi_L + \beta_3$	Zenes et al. (2020) for Ponderosa pine

Symbol	Value	Units	Meaning	Source
β_2	0	mol m ⁻² s ⁻¹ MPa ⁻¹	Linear coefficient between hydraulic risk, Θ, and leaf water potential, ψ _L ; used in the Anderegg <i>et al.</i> (2018)	Set to zero since Zenes $etaI.$ (2020) found that eta_1 and eta_2 covaried
β_3	0	mol m ⁻² s ⁻¹	model; $\Theta = \beta_1 \cdot \psi_1^L + \beta_2 \cdot \psi_1 + \beta_3$ Minimum hydraulic risk, Θ ; used in the Anderegg <i>et al.</i> (2018) model;	Set to zero to compare with other models, all of which have zero hydraulic risk at their maximum leaf water potentials and minimum transpiration rates
ψ L,crit	-2	MPa	$\Theta = \beta_1 \cdot \psi_1^{\ell} + \beta_2 \cdot \psi_1 + \beta_3$ Critical leaf water potential, ψ_1 , at which the carboxylation capacity,	Dewar <i>et al.</i> (2018)
			V _{c.max} , equals zero; used in the Dewar <i>et al</i> . (2018) model	

is the stem biomass in carbon equivalents, \tilde{z} is the axial distance along the stem from the stem apex normalized by the tree height, P_0 is axially variable potential turgor if NSCs were not limiting, which may exceed the true turgor limited by NSCs, P, and Γ is the threshold turgor must exceed before growth occurs (Eqn S6.1). Hence, G_0 depends on the average of $\max(P_0(z)-\Gamma,0)$ over the tree height between the stem apex ($\tilde{z}=0$) and the root collar ($\tilde{z}=1$). We model the temperature dependence of $\tilde{\phi}$ following Parent *et al.* (2010) using parameters provided by Cabon *et al.* (2020) and Peters *et al.* (2021) (Fig. S3). Our approach to estimate P_0 (Notes S1; Section S6) and its limitations and future improvements (Notes S1; Sections S10.1, S10.2) are discussed in the Supporting Information.

Upon applying the calculus of variations (Witelski & Bowen, 2015) to solve Eqn 1 for the g_c that maximizes growth (Notes S1; Section S2), the solution is defined in terms of the *marginal carbon cost of water*, χ_w . When stomata behave optimally, χ_w equals the *marginal carbon profit of water*, dA_n/dE , which we denote by λ as shorthand based on similar usage in *Gain-Risk* AOHs (Wolf *et al.*, 2016; Wang *et al.*, 2020; Figs 3a,b, S1), for which an alternate expression can be independently formulated from photosynthesis and gas exchange following Buckley *et al.* (2002, 2017) (Eqn S4.11). Simply, when water is more costly (larger χ_w), stomata close (smaller g_c). The same principle may be expressed in terms of hydraulic cost, $\Theta = \int \chi_w dE$, through which *Gain-Risk* AOHs are often framed (Wang *et al.*, 2020; Fig. 3c,d; Eqn S2.5). That is, stomata close as Θ increases. The GOH solution for χ_w is:

$$\chi_{\rm w} \equiv \left(\frac{1}{1-f_{\rm c}} - \frac{1}{\eta}\right) \frac{1}{a_{\rm L}} \frac{\partial G}{\partial E} = \left(\frac{1}{\eta} - \frac{1}{1-f_{\rm c}}\right) \frac{1}{a_{\rm L}} \left|\frac{\partial G}{\partial E}\right|$$
 Eqn 8(a)

$$\chi_{\rm w} \equiv \left(\frac{1}{1 - f_{\rm c}} - \frac{1}{\eta}\right) \frac{\sigma_{\rm g}}{a_{\rm L}} \frac{\partial G_0}{\partial E} = \left(\frac{1}{\eta} - \frac{1}{1 - f_{\rm c}}\right) \frac{\sigma_{\rm g}}{a_{\rm L}} \left|\frac{\partial G_0}{\partial E}\right| \quad \text{Eqn 8(b)}$$

$$\chi_{\rm w} = \lambda$$
 Eqn 8(c)

We list the identity of $\chi_{\rm w}$ twice to distinguish between a general identity (Eqn 8a) and an identity specific to how we model G here (Eqn 8b). Importantly, Eqn 8(a) holds for any growth-optimizing framework regardless of how growth is modeled (as long as G can be formulated as a function of E), while Eqn 8(b) is particular to our current GOSM. $\partial G_0/\partial E$ in Eqn 8(b) is solved by differentiating Eqn 7 with respect to E (Eqns S6.7, S6.11; $\partial G_0/\partial E \leq 0$). We informally refer to the solution of the Lagrange multiplier, η , as the marginal NSC-use efficiency (NSCUE; Notes S1; Section S2) $(0 \le \eta < 1 - f_c)$, which is analogous to the inverse of the marginal carbon cost of carbon use (Buckley & Roberts, 2006b) and represents the cost of spending NSCs on growth and respiration. Large η describes a strategy in which NSCs are highly valued, in which stomata open (small χ_w ; large g_c ; Fig. 3a) to increase A_n , decrease G, and maintain or even elevate NSCs. When η is small, NSCs are 'cheap', prompting plants to spend their NSCs by closing stomata (large χ_w and Θ ; small g; Fig. 3a,c) to increase G. Fig. 3(a) shows potential

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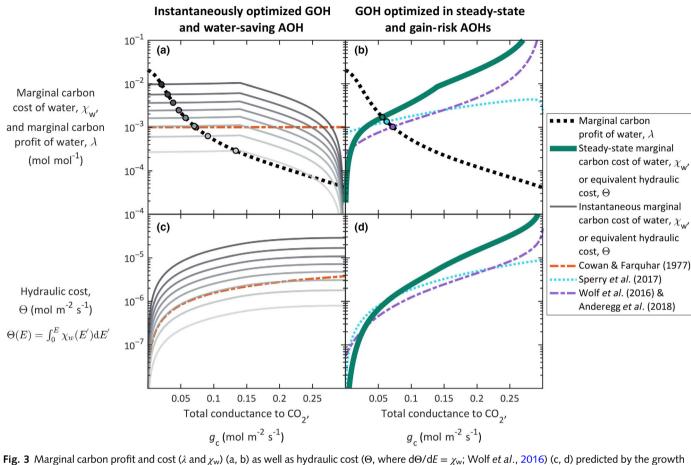


Fig. 3 Marginal carbon profit and cost (λ and χ_w) (a, b) as well as hydraulic cost (Θ, where dΘ/dE = χ_w ; Wolf et al., 2016) (c, d) predicted by the growth optimization hypothesis (GOH) and several assimilation optimization hypotheses (AOHs) for potential values of total conductance values, g_c , for the environmental conditions and physiological parameters listed in Table 3. (a) Instantaneous marginal costs, χ_w , calculated for different NSCUEs (η) according to the GOH, with η ranging from 20% to 90% of its maximum value (1 - f_c) in 10% increments (darkest to lightest). Instantaneous χ_w was calculated for an NSC storage based on measurements of Scots pine of similar size as our simulated tree by Schiestl-Aalto et al. (2019) (C = 175 mol). In (a), note that $\chi_w = 0$ for $g_c > c$. 0.3 mol m⁻² s⁻¹ (coinciding with where there is no growth), regardless of η . (b) Steady-state marginal cost, χ_w , calculated when both NSC storage and NSCUE have reached constant state (C = 0 and $\dot{\eta} = 0$) according to the GOH. Intersection of λ and λ or $\bar{\chi}$ curves (circles in a and b) coincides with the optimal g_c . When multiple intersections occur for instantaneous optimizations in (a), we select the lesser g_c , which coincides with higher growth rates; the greater g_c typically coincides with insignificant growth. (c) Hydraulic cost from instantaneous GOH corresponding to (a). (d) Hydraulic cost from steady-state GOH corresponding to (b).

 $\chi_{\rm w}$ and λ calculated across potential $g_{\rm c}$ for a range of η using the environmental conditions and physiological parameters listed in Table 3. The $\chi_{\rm w}$ and λ curves intersect at the optimal $g_{\rm c}$ (circles in Fig. 3a). Indeed, the optimal $g_{\rm c}$ increases as η increases. In the range of $g_{\rm c}$ values that $\chi_{\rm w}$ and λ typically intersect, $\chi_{\rm w}$ is near constant due to near-constant $\partial G_0/\partial E$. At larger $g_{\rm c}$, $\chi_{\rm w}$ declines to zero as E increases, ψ becomes more negative, turgor declines, and G_0 , $\partial G_0/\partial E$, and $\chi_{\rm w}$ all approach zero. Hence, multiple intersection points may exist in the instantaneous optimization. In this case, we select the lesser $g_{\rm c}$, which coincides with the higher G, since greater $g_{\rm c}$ reduces G (Notes S1; Section S2).

The calculus of variations suggests η is not constant and changes dynamically in response to environmental conditions, carbon use, and NSC storage (Eqn S2.6):

$$\begin{split} \dot{\eta} &= \frac{d\eta}{dt} = \eta \frac{\partial R_{\rm M}}{\partial C} + \left(\frac{\eta}{1 - f_{\rm c}} - 1\right) \frac{\partial G}{\partial C} \\ &= \eta \frac{\partial \sigma_{\rm r}}{\partial C} R_{\rm M,0} + \left(\frac{\eta}{1 - f_{\rm c}} - 1\right) \frac{\partial \sigma_{\rm g}}{\partial C} G_0 \end{split}$$
 Eqn 9

According to Eqn 9, $\partial R_{\rm M}/\partial C > 0$. This is because we do not expect excess NSCs to penalize growth (i.e. $\partial G/\partial C \geq 0$), and $0 \leq \eta < 1 - f_c$, which means that η would eventually become negative (i.e. $\dot{\eta} < 0$ for all times) if maintenance respiration either were independent of NSCs ($\partial R_{\rm M}/\partial C = 0$) or declined under elevated NSC storage ($\partial R_{\rm M}/\partial C < 0$). However, η must be non-negative. Thus, the existence of an optimal solution requires $\partial R_{\rm M}/\partial C > 0$, which indeed has empirical support (Sevanto *et al.*, 2014; Collins *et al.*, 2021) and thus justifies the functional forms of Eqns 4 and 6.

To test our GOH under varied environmental conditions, we consider solutions in steady state when carbon assimilation and use are in balance ($\dot{C}=0$) and the marginal NSCUE has ceased changing ($\dot{\eta}=0$). Steady state is only an approximation of how plants behave in reality; nonetheless, it provides insights into the optimal stomatal behavior to environmental cues. Unlike instantaneous solutions, it reflects stomata responses after they have equilibrated to environmental variations if those conditions were sustained constantly over an extended period of time (Feng *et al.*, 2015). We denote steady-state values by symbols with a vinculum (e.g. $\overline{\eta}$, \overline{C}). Setting Eqn 9 to zero and rearranging gives the steady-state marginal NSCUE (Eqn S2.7):

$$\begin{split} \overline{\eta} &= \left(1 - f_{c}\right) \frac{\frac{\partial \overline{G}}{\partial \overline{C}}}{\frac{\partial \overline{G}}{\partial \overline{C}} + \left(1 - f_{c}\right) \frac{\partial \overline{R}_{M}}{\partial \overline{C}}} \\ &= \left(1 - f_{c}\right) \frac{\frac{\partial \overline{\sigma}_{k}}{\partial \overline{C}} \overline{G}_{0}}{\frac{\partial \overline{\sigma}_{k}}{\partial \overline{C}} \overline{G}_{0} + \left(1 - f_{c}\right) \frac{\partial \overline{\sigma}_{t}}{\partial \overline{C}} R_{\mathrm{M},0}} \end{split}$$
 Eqn 10

Setting Eqns 8 and 10 equal and rearranging defines the steady-state marginal cost of water (Eqn S2.8).

$$\overline{\chi}_{\mathrm{w}} = -\frac{1}{a_{\mathrm{L}}} \frac{\partial \overline{G}}{\partial \overline{E}} \frac{\frac{\partial \overline{R}_{\mathrm{M}}}{\partial \overline{C}}}{\frac{\partial \overline{G}}{\partial \overline{C}}} = -\frac{\overline{\sigma}_{\mathrm{g}}}{a_{\mathrm{L}}} \frac{\partial \overline{G}_{0}}{\partial \overline{E}} \frac{R_{\mathrm{M},0}}{\overline{G}_{0}} \frac{\frac{\partial \overline{\sigma}_{\mathrm{t}}}{\partial \overline{C}}}{\frac{\partial \overline{\sigma}_{\mathrm{g}}}{\partial \overline{C}}} = \frac{\overline{\sigma}_{\mathrm{g}}}{a_{\mathrm{L}}} \left| \frac{\partial \overline{G}_{0}}{\partial \overline{E}} \right| \frac{R_{\mathrm{M},0}}{\overline{G}_{0}} \frac{\frac{\partial \overline{\sigma}_{\mathrm{t}}}{\partial \overline{C}}}{\frac{\partial \overline{\sigma}_{\mathrm{g}}}{\partial \overline{C}}} = \frac{\overline{\sigma}_{\mathrm{g}}}{a_{\mathrm{L}}} \left| \frac{\partial \overline{G}_{0}}{\partial \overline{E}} \right| \frac{R_{\mathrm{M},0}}{\overline{G}_{0}} \frac{\frac{\partial \overline{\sigma}_{\mathrm{t}}}{\partial \overline{C}}}{\frac{\partial \overline{\sigma}_{\mathrm{g}}}{\partial \overline{C}}} = \frac{\overline{\sigma}_{\mathrm{g}}}{a_{\mathrm{L}}} \left| \frac{\partial \overline{G}_{0}}{\partial \overline{E}} \right| \frac{R_{\mathrm{M},0}}{\overline{G}_{0}} \frac{\frac{\partial \overline{\sigma}_{\mathrm{t}}}{\partial \overline{C}}}{\frac{\partial \overline{\sigma}_{\mathrm{g}}}{\partial \overline{C}}} = \frac{\overline{\sigma}_{\mathrm{g}}}{a_{\mathrm{L}}} \left| \frac{\partial \overline{G}_{0}}{\partial \overline{E}} \right| \frac{R_{\mathrm{M},0}}{\overline{G}_{0}} \frac{\frac{\partial \overline{\sigma}_{\mathrm{t}}}{\partial \overline{C}}}{\frac{\partial \overline{\sigma}_{\mathrm{g}}}{\partial \overline{C}}} = \frac{\overline{\sigma}_{\mathrm{g}}}{a_{\mathrm{L}}} \left| \frac{\partial \overline{G}_{0}}{\partial \overline{E}} \right| \frac{R_{\mathrm{M},0}}{\overline{G}_{0}} \frac{\frac{\partial \overline{\sigma}_{\mathrm{t}}}{\partial \overline{C}}}{\frac{\partial \overline{\sigma}_{\mathrm{g}}}{\partial \overline{C}}} = \frac{\overline{\sigma}_{\mathrm{g}}}{a_{\mathrm{L}}} \left| \frac{\partial \overline{G}_{0}}{\partial \overline{C}} \right| \frac{R_{\mathrm{M},0}}{\overline{G}_{0}} \frac{\partial \overline{G}_{0}}{\partial \overline{C}}$$

Contrary to the instantaneous solution, the steady-state solution given by Eqn 11 has mathematically unique solution with a single g_c at which $\lambda = \overline{\chi}_w$ (Fig. 3b). This uniqueness results from $\overline{\chi}_w$ being a positive, increasing function of g_c and from $\overline{\chi}_w < \lambda$ at $g_c = 0$ (Wang *et al.*, 2020). In fact, $\overline{\chi}_w$ must equal zero at $g_c = 0$ due to the $\overline{\sigma}_g$ term in Eqn 11, which equals zero when $g_c = 0$ resulting from insignificant NSC storage and carbon *sup-ply* limitations (Fig. 2). $\overline{\chi}_w$ is an increasing function of g_c , because of the \overline{G}_0 term in the denominator of Eqn 11, which approaches zero faster than the decline in the other terms in the numerator.

Parameterization

We parameterized our GOSM with values reported or estimated from the existing literature to represent an evergreen conifer, many of which were synthesized by Potkay et al. (2021b) (Table 3). The majority of the parameters originate from studies of Scots pine (*Pinus sylvestris*); however, several parameters represent loblolly pine (*Pinus taeda*). No parameters were finely tuned to match observations, and we focus on the model's ability to capture fundamental stomatal responses by using parameters that are as physically based as possible. Four parameters were estimated indirectly ($R_{\rm M,0}$, $\widetilde{\phi}$, $\gamma_{\rm g}$, and $\gamma_{\rm r}$) to satisfy a net-to-gross primary productivity ratio of 0.45 (Waring et al., 1998; Collalti et al., 2020) and realistic NSC reserves under ambient and elevated $c_{\rm a}$ (Figs S4–S7; Notes S1; Section S8).

Responses to the environment

To test our GOSM to environmental cues, we consider both instantaneous and steady-state solutions. We determine instantaneous solutions for known η and C (Fig. 4; Eqn 10), assuming an NSC storage based on measurements of Scots pine of similar size as our simulated tree by Schiestl-Aalto et al. (2019) (C = 175 mol). Steady-state solutions require no assumptions about $\overline{\eta}$ or \overline{C} . Their values emerge from the optimization (Eqns 10, 11). We test GOSM predictions to VPD (by varying atmospheric relative humidity, RH; 0.1-0.8), soil water potential $(\psi_{\text{soil}}; -0.8 \text{ to } 0 \text{ MPa})$, atmospheric CO₂ concentration $(c_a;$ 250-650 ppm), and the loss of hydraulic conductance due to past drought. We quantify past drought severity by the minimum experienced soil water potential causing conductance loss ($\psi_{\text{soil}}^{\text{min}}$; -2.4 to -0.02 MPa) and present results from two approaches for conductance hysteresis (Notes S1; Section S9). One hysteresis approach considers the true loss of conductance (e.g. Mackay et al., 2015), and the alternative approach postulates plants maintain the optimal $\psi_{\rm L}$ as if embolism were impermanent (Venturas et al., 2018). We hold the other environmental conditions constant at the default values listed in Table 3. We consider how the environmental cues influence total conductance (g_c), the marginal WUE (λ), growth (G), and NSC storage (C).

Furthermore, we consider the possibility of long-term acclimation of leaf traits (leaf area, $a_{\rm L}$, and photosynthetic capacities) to elevated $c_{\rm a}$ (eCO₂) when testing our GOSM to increased $c_{\rm a}$. In acclimated simulations, we increased $a_{\rm L}$ by a factor of 1.25 based on Lauriks *et al.* (2021), reduced the maximum carboxylation capacity ($V_{\rm c,max}$) by a factor of 0.90, and reduced the maximum electron transport capacity ($J_{\rm max}$) by a factor of 0.95 based on Ainsworth & Rogers (2007).

Crucially, instantaneous optimization solutions may be neither realistic nor meaningful. There exists a temporal mismatch between key model variables required to solve the instantaneous optimization (η, C) and environmental conditions, since instantaneous optimizations do not consider their key feedback (Eqn 11). Instantaneous results are appropriate when stomatal responses due solely to environmental conditions (i.e. at constant η and C) are far faster than those due to dynamics in η and C that arise from environmental changes. These scenarios apply when testing environmental factors that vary significantly over hourly or shorter timescales in real environments (e.g. D_L) but are inappropriate for more slowly changing factors (e.g. ψ_{soil} , $\psi_{\text{soil}}^{\text{min}}$, and ca). Our steady-state results, nevertheless, capture the key feedback that ensures realistic responses of η , C, and g_c for slowly changing conditions, because steady-state predictions represent responses that have equilibrated to environmental conditions.

Additionally, we compare steady-state solutions to predictions by several extensively tested AOH models (Table S2), including those by Cowan & Farquhar (1977), Prentice *et al.* (2014), Wolf *et al.* (2016), Anderegg *et al.* (2018), Sperry *et al.* (2017), Eller *et al.* (2018), Dewar *et al.* (2018), and Wang *et al.* (2020). This comparison provides a first-order, quantitative test for the GOSM's predicted g_c , while previous tests checked for realistic qualitative trends. The comparison also assesses whether GOH

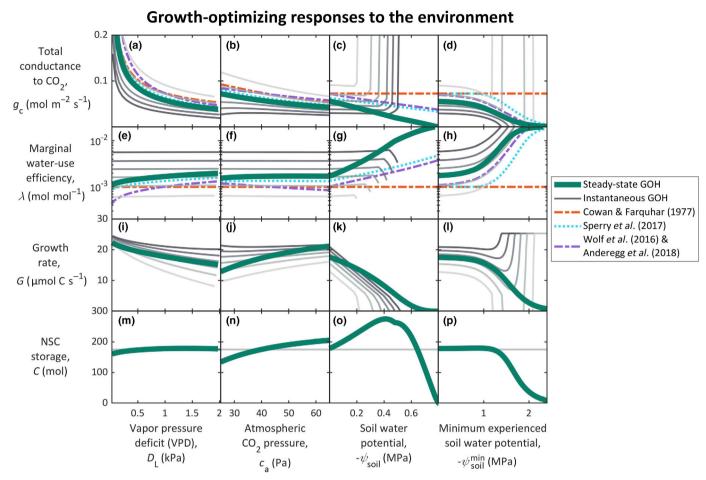


Fig. 4 Instantaneous and steady-state responses, according to the growth optimization hypothesis (GOH), to varied vapor-pressure deficit (VPD; D_L) (a, e, i, m), atmospheric CO₂ pressure (c_a) (b, f, j, n), soil water potential ($-\psi_{soil}$) (c, g, k, o), and losses in hydraulic conductance (d, h, l, p). Predictions include the optimal total conductance (g_c) (a–d), the optimal marginal water-use efficiency (λ) (e–h), the coinciding growth rate (G) (i–l), and the coinciding non-structural carbohydrate (NSC) storage (C) (m–p). Instantaneous optimizations were calculated for different NSCUEs (η) according to the GOH, with η ranging from 30% to 80% of its maximum value (1 – f_c) in 10% increments (darkest to lightest), and were calculated for an NSC storage based on measurements of Scots pine of similar size as our simulated tree by Schiestl-Aalto *et al.* (2019) (C = 175 mol). In (d, h, l, p), conductance was lost as if exposed to a previous drought with magnitude denoted by the minimum experienced soil water potential (ψ_{soil}^{min}), assuming that this conductance loss is irrecoverable. Despite the loss of conductance due to past drought (i.e. negative ψ_{soil}^{min}), results in (d, h, l, p) represent a tree that is currently well-watered (i.e. ψ_{soil} = 0), and thus these results represent postdrought recovery. Sharp increases in the instantaneous g_c under current or past soil water stress in (c, d) (i.e. at more negative ψ_{soil} and ψ_{soil}^{min}) results from coinciding instantaneous λ = 0 in (g_c , h) and thus also instantaneous G_c = 0 in (g_c , l). Predictions made by three selected AOHs are shown for comparison.

and AOH predictions are consistent despite differences in their proxies for evolutionary fitness. Mathematical details of the AOHs are presented in the Supporting Information (Notes S1; Section S7), and their parameters are listed in Table 3.

Results

Both instantaneous and steady-state optimizations predict stomata close as VPD increases (Fig. 4a). Predicted sensitivities of g_w to VPD ($m = -dg_w/dlog_e(D_L)$) from instantaneous and steady-state optimizations agree with reported ranges ($m \approx 0.5-0.6$; Oren *et al.*, 1999; Fig. S8). In instantaneous optimizations with large η , VPD sensitivities are smaller ($m \approx 0.47$) but still realistic (Katul *et al.*, 2009). The instantaneous marginal WUE (λ) is nearly independent of VPD, except at large η , when λ increases

slightly with VPD (Fig. 4e). In steady-state optimizations, increasing VPD resulted in larger $\bar{\lambda}$, slightly more NSCs, and slower growth (Fig. 4e,i,m). Steady-state predictions of g_c , $\bar{\lambda}$, and their VPD trends agree well with AOH models (Figs 4a,e, S9a,e), which are all instantaneous in nature. The Prentice *et al.* (2014) model is the only model that predicts λ declines slightly with VPD (Fig. S9e).

Both instantaneous and steady-state optimizations predict stomatal closure under eCO₂ (Fig. 4b), agreeing with AOH models, except the model by Wolf *et al.* (2016) and Anderegg *et al.* (2018), which is the only model to predict a negative relationship between λ and c_a (Figs 4f, S9f; Wang *et al.*, 2020). In instantaneous and steady-state optimizations, λ is nearly independent of c_a (Fig. 4f), while some AOH models predict λ increases (Figs 4f, S9f). The GOSM can also predict greater increases in λ

under eCO₂ for particular combinations of γ_r and γ_g (Fig. S10). These combinations of γ_r and γ_g , however, do not necessarily coincide with NSC reserves under ambient and elevated c_a (Figs S6, S7). Regardless of γ_r and γ_g , eCO₂ always promotes growth (Fig. S11) and stomatal closure (Fig. S12). Predictions with additional acclimation of a_L , $V_{c,max}$, and J_{max} to eCO₂ always show an increase in λ under eCO₂ regardless of γ_r and γ_g (Fig. S13). Strictly speaking, acclimation of a_L , $V_{c,max}$, and J_{max} is an optimal strategy only if it further enhances growth, which is indeed the case (Fig. S14).

Instantaneous solutions for g_c and λ are generally independent of ψ_{soil} , though they respectively increase and decrease with η (Fig. 4c). Beyond a threshold in ψ_{soil} , instantaneous g_c dramatically increases and λ approaches zero as turgor becomes too small for growth (Fig. 4g,k). Conversely, steady-state optimization predicts g_c and G decrease and $\bar{\lambda}$ increases as soils dry (Fig. 4g). Nonstructural carbohydrates are maximal under moderate soil water stress, because reductions in A_n are smaller than the reduction in G, thus elevating NSCs (Fig. 40). Under extreme stress, NSCs deplete due to strong stomatal limitations to A_n (Fig. 40). Steady-state predictions of g_c and $\bar{\lambda}$ to soil water stress agree with AOH models (Figs 4c,g, S9c,g), except the Cowan & Farquhar (1977) and Prentice et al. (2014) models, both of which require additional parameterizations to describe drought responses (Manzoni et al., 2011; Lavergne et al., 2020a). Steadystate g_c is more sensitive to declines in ψ_{soil} than instantaneous AOH models (Figs 4c, S9c), representing a more conservative water-use strategy in response to soil water stress.

Drought-induced xylem conductance loss induced stomatal closure and increased λ during postdrought recovery in both instantaneous and steady-state optimizations (Fig. 4d,h). Near-identical steady-state results are found when conductance hysteresis is modeled by Venturas *et al.*'s (2018) approach, which

considers plant hydraulics but not NSCs (Fig. S15). Predictions agree with AOH models regardless of how we modeled conductance hysteresis (Figs 4d,h, S15). The Cowan & Farquhar (1977) and Prentice *et al.* (2014) models, however, consider leaf-level processes only and thus cannot respond to extrafoliar hydraulics (Figs 4d,h, S9d,h). When applying Venturas *et al.*'s (2018) approach to conductance hysteresis, all models behave similarly, even the Cowan & Farquhar (1977) and Prentice *et al.* (2014) models (Fig. S16).

Postdrought growth and whole-tree conductance progressively decline after worsening past droughts (Figs 4l, 5). After moderate droughts ($\psi_{\text{soil}}^{\min} > c$. -1.2 MPa; PLC < c. 78%), NSCs can recover to predrought levels (Figs 4p, 5). After severe droughts ($\psi_{\text{soil}}^{\min} < c$. -1.44 MPa; PLC > c. 85%), NSCs deplete, approaching starvation for $\psi_{\text{soil}}^{\min} < -2.4$ MPa (Figs 4p, 5). Near-identical postdrought results for growth, conductance, and NSCs are found by Venturas *et al.*'s (2018) approach for conductance hysteresis (Figs S15, S17). These postdrought responses differ remarkably from predictions under permanent water stress (Fig. 5). Under permanent water stress, acclimated NSCs accumulate with PLC between 0% and c. 38%, past which NSCs rapidly deplete, approaching NSC depletion when PLC $\approx 47\%$ and soil water stress is only moderate ($\psi_{\text{soil}} = -0.8$ MPa; Figs 40, 5).

Discussion

Comparing GOH and AOHs

Here, we hypothesized stomata follow strategies that maximize stem growth over a plant's entire lifetime, which we call the GOH. Integrated over time, maximizing growth further maximizes size, enhancing their ability to compete for resources (King, 1990; Franklin, 2007) and reproduce (Greene & Johnson, 1994;

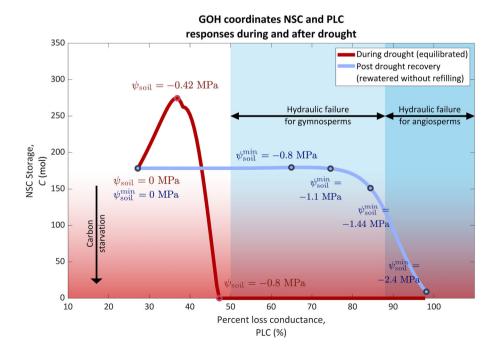


Fig. 5 Phase diagram of nonstructural carbohydrates (NSCs) and percent loss of whole-plant conductance (PLC) for our simulated Scots pine in equilibrium with varied ψ_{soil} (during drought) and $\psi_{\text{soil}}^{\text{mir}}$ (postdrought recovery; rewatered without refilling of embolisms at $\psi_{\text{soil}} = 0$) and other environmental forcing listed in Table 3. Coinciding values of ψ_{soil} and ψ_{soil}^{min} are denoted (circles) to signify the magnitude of soil water stress. Equilibration to ψ_{soil} represents acclimation to permanent soil water stress, while equilibration to $\psi_{\text{soil}}^{\text{min}}$ represents postdrought recovery, particularly after a transient period of intense soil water stress that leads to irrecoverable embolism. Hydraulic failure is typically associated with a PLC of at least 50% for gymnosperms (Brodribb & Cochard, 2009) and 88% for angiosperms (Resco et al., 2009; Urli et al., 2013).

Obeso, 2002; Minor & Kobe, 2019). We have shown the GOH agrees with our current understanding of stomatal regulation in relation to environmental forcings (Figs 4, S9). We compared these results to the historically prevailing assumption that stomata optimize photosynthetic carbon assimilation in relation to external water availability or internal water stress (e.g. Cowan & Farquhar, 1977; Wolf *et al.*, 2016), which we call AOHs.

All optimization theories assume plants have developed stomatal strategies that maximize evolutionary fitness. This assumption is reasonable, since it springs from natural selection, a core principle of biology (Mäkelä et al., 2002; Franklin et al., 2020). For biologically realistic solutions, AOHs must make a secondary assumption about additional costs or constraints that penalize stomatal opening; otherwise, AOHs would always predict maximal conductance, since A_n is a positive monotonic function of g_c (in the absence of nonstomatal limitations to photosynthesis). These costs and constraints have been associated with watersaving strategies (solved either over short durations; Cowan, 1977; Cowan & Farquhar, 1977; Hari et al., 1986; or dynamically; e.g. Mäkelä et al., 1996; Manzoni et al., 2013), xylem embolism (e.g. Wolf et al., 2016; Sperry et al., 2017), and nonstomatal limitations (e.g. Dewar et al., 2018). Few experiments have attempted to determine the true nature of these costs (Hall & Schulze, 1980; Fites & Teskey, 1988; Thomas et al., 1999; Manzoni et al., 2011). That is, costs are assumed a priori, and the validity of those assumptions relies on how well predictions of g_c match observations. The fact that a wide range of assumed costs all converge toward similar responses to environmental cues (Figs 4, S9) demonstrates the difficulty in inferring the true cost from g_c responses. The GOH, on the contrary, does not require assuming additional costs, since the cost of opening stomata is implicit in how plants grow. While opening stomata assimilate carbon as NSCs, which build tissue and generate turgor (Sapes et al., 2021), opening reduces ψ (Sperry et al., 1998) and thus also cambial turgor pressures (Hölttä et al., 2010), thereby hindering the division and expansion of cells (Lockhart, 1965; Kirkham et al., 1972). Hence, the GOH advances turgor-limited growth as a simpler proxy for evolutionary fitness.

Many AOHs maximize their assimilation instantaneously for the immediate environmental conditions without consideration of future resource availability, potentially leading to overly aggressive resource use, hampering future fitness (Lu et al., 2020; Feng et al., 2022). Here, we compared GOH predictions to several instantaneous AOHs (Wolf et al., 2016; Wang et al., 2020), which generally predict biologically correct responses of g_c to environmental stimuli (Wang et al., 2020; Table S2). They compared well to the steady-state predictions of our GOSM, both qualitatively and quantitatively (Figs 4, S9). However, whether instantaneously AOHs truly reflect current conditions or whether they compensate for delayed acclimation to future conditions through their assumed costs (λ) remains unclear. For example, some AOHs predict an immediate increase in λ as soon as exposed to eCO₂ (Figs 4f, S9f). However, little is known about how quickly λ changes under eCO₂ in reality and whether these changes solely reflect stomatal movements. While experiments suggest λ indeed increases under eCO₂ over multiple years (e.g.

Katul *et al.*, 2010), increases in λ may reflect other physiological acclimations to eCO₂ over longer timescales (Buckley & Schymanski, 2014), including acclimation of photosynthetic capacities, leaf area, and NSC reserves (Ainsworth & Rogers, 2007; Dietze *et al.*, 2014; Lauriks *et al.*, 2021; Fig. S13). Hence, instantaneous AOHs are inappropriate for examining responses to long-term environmental change (Medlyn *et al.*, 2013). If stomata truly maximize growth, then our results would suggest instantaneous AOHs (e.g. Wolf *et al.*, 2016) may predict realistic g_c by approximating foliar acclimations that have not yet occurred.

Our steady-state solution represents the coordination among g_c, G_c, and C upon complete acclimation to constant environmental conditions, while the instantaneous solution represents a snapshot in time as g_c, G, and C transition dynamically from one state to another. The dynamic state represented by the instantaneous solution depends on η and C (Figs 3, 4), their temporal evolution (Eqns 2, 9), and the variation in environmental conditions over small timescales. Analogously, our steady-state solution is comparable to observations from long experiments with small variations in the environment and other physiological properties (a_L , $V_{c,max}$) J_{max} , xylem conductance, allometry, etc.). Our instantaneous solution is comparable to observations from short experiments. However, properly comparing instantaneous solutions to observations would require explicitly modeling the dynamics of η and C. In steady state, the GOH consistently captured stomatal responses to $D_{\rm L}$, $c_{\rm a}$, $\psi_{\rm soil}$, and $\psi_{\rm soil}^{\rm min}$ (Fig. 4a-d), while instantaneous solutions explained only responses to D_L (Fig. 4a), which changes quickly. However, $c_{\rm a}$, $\psi_{\rm soil}$, and $\psi_{\rm soil}^{\rm min}$ change slowly in reality and likely at rates that are similar to or slower than changes in η and/or C, which are assumed constant in our instantaneous predictions (Fig. 4). Future work should explore how quickly η evolves. Thus, steady state is better than the instantaneous solution for predicting stomatal behavior to slowly changing c_a , ψ_{soil} , and $\psi_{\text{soil}}^{\text{min}}$, since steady state approximates the key dynamic feedback between η , C, and the environment.

Interestingly, the GOH predicts the instantaneous marginal WUE (χ_w) is generally near constant over short timescales with respect to potential g_c (Fig. 3a) and over broad environmental conditions (Fig. 4e-g), resulting from growth being a near-linear function of transpiration and g_c ($\partial^2 G_0/\partial E^2 \approx 0$; Fig. 2). Conversely, the GOH predicts the steady-state marginal WUE $(\overline{\chi}_w)$ varies strongly with potential g_c (Fig. 3b) with environmental conditions over longer timescales once both η and C have equilibrated (Fig. 4e-h). Thus, in the short term, the GOH behaves like the earliest water-saving AOH which specifies a constant χ_{w} (Cowan, 1977; Cowan & Farquhar, 1977; Figs 3a, 4), which much evidence supports (Medlyn et al., 2011; Buckley et al., 2017). In the long term, the GOH behaves like recent Gain-Risk AOHs, which specify χ_w (or its related hydraulic cost, Θ ; $d\Theta$ / $dE = \chi_w$) depends on environmental conditions and physiological variables (e.g. Wolf et al., 2016; Sperry et al., 2017; Figs 3b-d, 4). The GOH reconciles differences between early and recent AOHs.

Additionally, the GOSM suggests $\chi_{\rm w}=0$ when growth cannot occur (G=0) in instantaneous solutions (Fig. 3a), since $\chi_{\rm w} \propto \partial G_0/\partial E$ (Eqn 8b) and $\partial G_0/\partial E=0$ when $G_0=0$ (Fig. 2).

That is, our instantaneous GOH is maximizing solely A_n without shadow costs ($\Theta=0$) when plants cannot grow. Nonstomatal limitations should be incorporated into the GOH for these instances and improve predictions when plants cannot grow (Fig. 4g,h) following aspects of AOHs that maximize A_n subject to nonstomatal limitations, since these AOHs implicitly assume $\chi_w=0$ (Hölttä *et al.*, 2017; Dewar *et al.*, 2018, 2022).

Xylem embolism and mortality

Stomata have long been theorized to close early under water stress to prevent xylem embolism and conductance loss, thereby inspiring many models which postulate embolism and conductance loss as dominant controls on stomata (e.g. Jones & Sutherland, 1991; Sperry & Love, 2015; Sperry et al., 2016, 2017; Wolf et al., 2016; Eller et al., 2018; Carminati & Javaux, 2020). Under water stress, our GOSM predicts similar results as embolism- and conductance-centric models (Figs 4, S9), though without explicitly minimizing embolism or maximizing conductance. This agreement may not be surprising, considering numerous stomatal strategies inadvertently prevent embolism (Novick et al., 2016; Dewar et al., 2022). While plants indeed tend to operate with minimal embolism and constant xylem conductance (Carminati & Javaux, 2020), this phenomenon does not necessitate that stomata strictly close to prevent embolism and conductance loss. In fact, stomata behave more conservatively than required to prevent significant embolism (Anderegg et al., 2017; Martin-StPaul et al., 2017; Buckley, 2021), suggesting stomata close to maintain other biological functions that are more sensitive to water stress, including turgor maintenance and growth (Hsiao & Acevedo, 1974; Bartlett et al., 2016). Our GOH does not explicitly incorporate embolism prevention or conductance maintenance into its assumptions but nonetheless minimizes embolism and conductance loss. If xylem lost conductance, downstream tissues would have to operate under greater tension, thereby reducing turgor and growth. Hence, stomata close to maximize growth, avoiding conductance losses which would penalize growth.

The prevalence of the idea that stomata operate to prevent embolism and maintain xylem conductance likely originates from widely held notions of how plants die. McDowell et al. (2008, 2011) proposed a popular framework in which embolism ultimately leads to desiccation (the so-called hydraulic failure) and mortality. Thus, stomata are expected to prevent mortality by moderating embolism and conductance loss. The notion of hydraulic failure has stimulated a large number of studies on the degree of embolism past which plants die (e.g. Brodribb & Cochard, 2009; Barigah et al., 2013; Urli et al., 2013; Adams et al., 2017) and how the corresponding ψ threshold varies among species and environments (e.g. Choat et al., 2012; Oliveira et al., 2019; Rosas et al., 2019). However, several recent experimental studies have demonstrated plants can recover despite surpassing these thresholds (Li et al., 2016; Hammond et al., 2019; Mantova et al., 2021), and modeling studies could not identify thresholds to accurately predict mortality in forest stands (De Kauwe et al., 2020; Venturas et al., 2021). These studies cast doubt on embolism as a useful predictor of mortality

as well as on whether stomata strictly attempt to prevent embolism. Desiccation (loss of water content and turgor of the meristem and inner bark) is indeed mechanistically linked to mortality (Sapes et al., 2019; Preisler et al., 2021; Sapes & Sala, 2021; Mantova et al., 2022). However, embolism alone cannot describe the critical sequence of events leading up to desiccation and mortality. How quickly the meristem and inner bark desiccates depends not only on the degree of water stress (expressed by ψ or corresponding conductance), but also on the radial conductance between xylem and inner bark (Sevanto et al., 2011; Baert et al., 2015; Preisler et al., 2021), other extra-xylem resistances (e.g. Scoffoni et al., 2017), tissues' water storage capacities (Meinzer et al., 2009; Salomón et al., 2017; Preisler et al., 2022), phloem transport (Chan et al., 2016; Epron et al., 2021), and NSC storage (O'Brien et al., 2014; Sapes et al., 2021). Hence, strategies for avoiding desiccation and mortality depend on complex water and carbon transport processes throughout the plant rather than simply embolism (McDowell et al., 2022). Through the GOH, we offer a general framework to incorporate all of these processes to predict mortality directly from tissue water content as outlined in the Supporting Information (Notes S1; Section S10.2).

Our GOH shifts focus on the predictor of drought-induced mortality from xylem conductance and embolism to turgor (or water content; Notes S1; Sections S2, S10.1, S10.2). Even though it does not include many of the water and carbon transport processes required to predict the time and stress required to desiccate the meristem and kill a plant, our GOSM does represent NSC storage, a key control of survival (O'Brien et al., 2014; Sapes et al., 2021). Thus, our GOSM can predict how percent loss of conductance (PLC) and NSC storage are related for hypothetical drought scenarios (Fig. 5). Though conceptually interdependent (McDowell et al., 2011, 2022), PLC and NSCs are considered two distinct axes for mortality in practice (e.g. Anderegg et al., 2012; Adams et al., 2017). Notably, the GOH provides a link between PLC and NSCs, though the exact relationship differs between duringdrought and postdrought conditions (Fig. 5). Conversely, AOHs do not consider NSCs and thus cannot predict such coordination. Though recent studies suggest PLC is not the best predictor of mortality (De Kauwe et al., 2020; Venturas et al., 2021), we nevertheless demonstrate the GOSM's ability to capture fundamental processes and trends between PLC and NSCs (Fig. 5), since a large body of empirical studies of PLC exists to test model predictions. However, we do not focus on the exact values of this PLC-NSC coordination (Fig. 5), which are species- and climate-specific, since we have parameterized the GOSM for a mature Scots pine at a single site (Table 3).

In all cases, plants increasingly embolize as drought worsens, but NSC trends depend on the severity of past and current water stress (Fig. 5). During persistent drought (i.e. long enough for plants to equilibrate; dC/dt = 0; $d\eta/dt = 0$), NSCs accumulate under moderate water stress ($\psi_{soil} > -0.42$ MPa) and deplete under severe water stress ($\psi_{soil} < -0.42$ MPa; Fig. 5), explaining conflicting NSC patterns from drought studies (e.g. Galiano *et al.*, 2011; Galvez *et al.*, 2011; Piper, 2011) and observations of NSCs first elevating before later depleting during progressively worsening droughts (e.g. Adams *et al.*, 2013; Mitchell *et al.*, 2014). Our simulated tree

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would certainly be incapable of inhabiting environments where ψ_{soil} never exceeded -0.8 MPa due to the inability to maintain minimum NSCs (the so-called carbon starvation; Fig. 5). The exact minimum ψ_{soil} trees that could indefinitely tolerate should correspond to their minimum-tolerable tissue water content or turgor, which integrate both PLC and NSCs (Martínez-Vilalta et al., 2019; Sapes et al., 2019). Our GOSM predicts potential turgor if NSCs were nonlimiting, P_0 , rather than explicitly modeling the true turgor, P (Notes S1; Sections S6, S10.1), and thus we cannot define the exact minimum ψ_{soil} that could be tolerated indefinitely, though it would certainly be more negative than -0.42 MPa (Fig. 5). Improvements to the GOSM would enable explicit prediction of P, water content, and thus mortality (Notes S1; Section \$10.2). These predictions under long-term water stress may be interpreted as death by carbon starvation without hydraulic failure, considering hydraulic failure would require far more embolism (PLC > 50% for conifers; Brodribb & Cochard, 2009). However, plants rarely die by carbon starvation alone (Adams et al., 2017), suggesting steady-state predictions (i.e. dC/dt = 0; $d\eta/dt = 0$) of during-drought responses are not appropriate for interpreting mortality.

We also considered how plants operate postdrought with replenished soil water, having lost significant conductance (Anderegg et al., 2012), since mortality often does not occur during the drought itself but lags by several years (Trugman et al., 2018). Hence, postdrought simulations of plants with significant conductance loss (Fig. 5) represent legacy effects remaining after drought (Kannenberg et al., 2020). Again, we cannot predict the exact 'point of no return'. Nevertheless, assuming our simulated tree survived the initial drought, the tree can completely recover its NSC reserves even after severe drought ($\psi_{\text{soil}}^{\text{min}} > -1.2 \text{ MPa}$), despite suffering PLCs as large as c. 80% (Fig. 5). If mortality eventually occurred in this scenario, the death mechanism would be hydraulic failure without carbon starvation, consistent with some studies (e.g. Anderegg et al., 2012; Garcia-Forner et al., 2017; Kannenberg & Phillips, 2020). After droughts in which $\psi_{\text{soil}}^{\text{min}} < -1.44$ MPa, simulated trees were incapable of fully recovering their NSC reserves with whole-tree PLCs in excess of 85% regardless of embolismhysteresis approaches (Figs 4p, 5). Interestingly, this simulated threshold ($\psi_{\text{soil}}^{\text{min}} = -1.44$) nearly equals the permanent wilting point (classically, $\psi_{\text{soil}} = -1.5$ MPa in soil sciences; da Silva *et al.*, 1994). However, this degree of soil water stress may not always challenge plants, since the actual soil wilting point is soil-, plant-, and climate-specific (Torres et al., 2021). For trees eventually dying in this second scenario, the death mechanisms are hydraulic failure and carbon starvation. Here, PLC and NSCs are negatively correlated, which evidence supports (Adams et al., 2017; Tomasella et al., 2020), signifying the mechanisms' interdependence. These postdrought predictions may explain why hydraulic failure is widely observed at death, while dead plants may or may not suffer carbon starvation (Adams et al., 2017).

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Competing interests

None declared.

Author contributions

AP and XF designed the model and simulation experiments. AP coded and ran the model, analyzed simulated data, and led the writing. All authors contributed to data interpretation, discussion, figure preparation, and the final version.

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Data availability

MATLAB codes for model and plotting/analyzing data, including forcing data, and simulation outputs are available in Notes S2.

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Supporting Information

Additional Supporting Information may be found online in the Supporting Information section at the end of the article.

- **Fig. S1** Flowchart and comparison of model structures from growth optimization hypothesis and assimilation optimization hypotheses.
- **Fig. S2** Steady-state growth optimization hypothesis responses of key model variables for potential values of stomatal conductance under varied environmental forcings.
- Fig. S3 Temperature dependence of the whole-stem extensibility.
- **Fig. S4** Predictions of $R_{\rm M,0}$ that satisfy NPP : GPP = 0.45 under steady-state growth optimization.
- **Fig. S5** Predictions of ϕ that satisfy NPP : GPP = 0.45 under steady-state growth optimization.

- **Fig. S6** Predictions of NSC reserves (*C*) that satisfy NPP: GPP = 0.45 under steady-state growth optimization.
- **Fig. S7** Percent change in nonstructural carbohydrate storage between elevated and ambient atmospheric CO₂ concentrations under steady-state growth optimization.
- **Fig. S8** Sensitivity (*m*) of growth optimization hypothesis predicted stomatal conductance to vapor-pressure deficit.
- **Fig. S9** Comparison of steady-state growth optimization hypothesis predictions to all of the considered assimilation optimization hypotheses.
- **Fig. S10** Change in the marginal water-use efficiency between elevated and ambient atmospheric CO_2 concentrations under steady-state growth optimization.
- **Fig. S11** Percent change in growth rate between elevated and ambient atmospheric CO₂ concentrations under steady-state growth optimization.
- **Fig. S12** Change in stomatal conductance between elevated and ambient atmospheric CO₂ concentrations under steady-state growth optimization.
- **Fig. S13** Change in the marginal water-use efficiency between elevated and ambient atmospheric CO₂ concentrations under steady-state growth optimization if trees adapted to elevated atmospheric CO₂ concentrations have acclimated their leaf traits.
- **Fig. S14** Difference in growth rate between trees with and without acclimation of leaf traits, both of which grow under elevated atmospheric CO₂ concentrations.
- **Fig. S15** Comparison of responses of growth optimization model to losses in hydraulic conductance following the two different approaches for the calculation of how stomata respond to conductance losses.
- **Fig. S16** Comparison of predictions by growth optimization hypothesis and assimilation optimization hypothesis models under varied hydraulic conductance following the alternative approach of Venturas *et al.* (2018) to account for how permanent conductance losses impact predictions.
- **Fig. S17** Phase diagram of nonstructural carbohydrate and percent loss of conductance resulting from acclimation to permanent water stress or from postdrought recovery after transient water stress following the alternative approach of Venturas *et al.* (2018) to account for how permanent conductance losses impact predictions
- **Notes S1** Full model, parameter estimation, and simulation descriptions.

Notes S2 Matlab code for model and plotting/analyzing data, including forcing data and simulation outputs.

Table S1 Main variables and symbols in model.

Table S2 Additional stomatal optimization models and studies testing their predictions.

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