Conifers depend on established roots during drought: results from a coupled model of carbon allocation and hydraulics

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Introduction

Global climate dynamics are causing more frequent and widespread tree die-offs as a result of extreme droughts and higher temperatures (Adams et al., 2009; Williams et al., 2013; McDowell et al., 2016). Ability to access water from reliable sources, such as groundwater stored in bedrock fractures (McLaughlin et al., 2017), is one strategy some trees may use to survive in a warmer world with intense droughts. This is especially critical in situations of scarce precipitation (Tai et al., 2018). Plants must grow coarse root systems towards groundwater (Fan et al., 2017) and maintain sufficient absorbing fine roots to acquire the groundwater. Studies suggest that plants in drought-prone regions maintain roots specifically for accessing groundwater when soil layers dried out. Therefore, the modeling framework identified a critical mechanism for drought response that required trees to shift water uptake from existing roots rather than growing new roots.

Summary

- Trees may survive prolonged droughts by shifting water uptake to reliable water sources, but it is unknown if the dominant mechanism involves activating existing roots or growing new roots during drought, or some combination of the two.
- To gain mechanistic insights on this unknown, a dynamic root-hydraulic modeling framework was developed that set up a feedback between hydraulic controls over carbon allocation and the role of root growth on soil–plant hydraulics. The new model was tested using a 5 yr drought/heat field experiment on an established piñon-juniper stand with root access to bedrock groundwater.
- Owing to the high carbon cost per unit root area, modeled trees initialized without adequate bedrock groundwater access experienced potentially lethal declines in water potential, while all of the experimental trees maintained nonlethal water potentials. Simulated trees were unable to grow roots rapidly enough to mediate the hydraulic stress, particularly during warm droughts. Alternatively, modeled trees initiated with root access to bedrock groundwater matched the hydraulics of the experimental trees by increasing their water uptake from bedrock groundwater when soil layers dried out.
- Therefore, the modeling framework identified a critical mechanism for drought response that required trees to shift water uptake among existing roots rather than growing new roots.

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and the ability of some species to disconnect their fine roots from the soil under water deficit (West et al., 2007a,b; Plaut et al., 2013) suggest that fine roots act as hydraulic fuses (Venturas et al., 2017). Control of aquaporins or changes in fine root area can also cause adjustments to hydraulic conductance (Gambetta et al., 2012; Venturas et al., 2017). To account for these active roles of fine roots, models need to define fine roots in a way that is meaningful for quantifying water uptake (McCormack et al., 2015) and integrate plant hydraulic status with a C-mediated growth strategy (Fisher et al., 2018; Hartmann et al., 2018).

Two challenges in building models that integrate both active and passive root functions are the apparently contradictory root growth responses to drought and the large variability in root traits and passive root functions are the apparently contradictory root growth during seasonal or long-term periods of soil water deficit (Metcalfe et al., 2008; Barbeta et al., 2015; Doughty et al., 2014), suggesting that trees rely heavily on existing roots during drought. The role of a stratified root architecture having access to reliable water during drought provides a better explanation for tree survival than root : shoot ratios or biomass allocation (Padilla & Pugnaire, 2007; Laclau et al., 2013; Doughty et al., 2014), but just how much deep root area is needed is unknown. One observation that can be made from these studies is that the maintenance of roots near stable water sources reduces the need to grow new roots after the onset of drought.

The second challenge to building integrated models of C allocation and hydraulics is that each unit investment of C can yield different water-uptake capacities (Reich et al., 1998; Bauhus & Messier, 1999; Withington et al., 2006; Bowsher et al., 2016; Kramer-Walter et al., 2016) associated with phylogenetic variations (Comas et al., 2002, 2014; Comas & Eissenstat, 2009; Ma et al., 2018). In particular, the finest root diameters show a large variability among taxa, soil textures and climates (Eissenstat et al., 2015; Cheng et al., 2016; Liu et al., 2016). For example, first-order (or finest) roots in subtropical areas can be thicker because of favorable conditions for water uptake (Chen et al., 2013), whereas temperate trees generally have thinner first-order roots (Pittermann et al., 2012). In turn, this results in different water-uptake capacities per unit C invested. Two widely measured traits, specific root length, a measure of root length per unit C invested, and first- and second-order root diameters, are potentially useful here because together they define root tissue density (Ostonen et al., 2007; Ma et al., 2018; see also Supporting Information Fig. S1). Root tissue density informs the C allocation needed per unit increment of root volume and, by implication, of root area. Notably, evergreen needleleaf tree root tissue densities are high relative to other taxa, as they have low specific root areas across a wide range of diameters (Fig. S1). Thus, a second observation is that specific root length and root diameter can potentially be utilized along with the observation that optimal water transport is observed when root xylem and rhizospheres are colimiting (Sperry et al., 2002) to provide a robust link between C allocation and hydraulics.

Here we present a new modeling framework that addresses the first issue regarding variability in observed root growth responses to drought by allowing for transience in root growth in a feedback with the plant hydraulic system, and addresses the second issue in variation in root traits by representing the fine root system with a small number of measurable traits that can be used to translate C allocation into absorbing root area. We make no assumptions about root trait correlations, and instead allow the integrated dynamics of C allocation and hydraulic function to be an emergent response to environmental conditions. Our objective is to add mechanistic insight on fine root dynamics, associated C costs, and their integration with rhizosphere–plant hydraulics, specifically during hot and dry conditions. We tested the framework at a piñon-juniper site with empirical evidence of root water uptake from bedrock fractures and strong tree resistance to droughts, including warmer droughts (Grossiord et al., 2017a).

To focus this study, we use the new framework to test two hypotheses: (H1) trees rely on bedrock groundwater during dry periods and soil water during wetter periods to maintain their hydraulic status; and (H2) fine roots with access to bedrock groundwater must be maintained at all times rather than grown after the onset of drought.

**Materials and Methods**

**Study site description**

This study was conducted at the Los Alamos Survival-Mortality (SUMO) experiment in New Mexico, USA (35.49°N, 106.18°W, 2175 m asl). The soil is Hackroy clay loam, which is derived from volcanic tuff (Soil Survey Staff, Natural Resources Conservation Service, United States Department of Agriculture; http://websoilsurvey.nrcs.usda.gov), with an average depth of 65 cm. The volcanic tuff at the site is fractured (Trainer, 1974), which allows tree roots to grow into the bedrock (Tierney & Foxx, 1982; Newman et al., 1997). Calcite precipitation in the near-surface fractures induced by root growth and decay (Newman et al., 1997) and isotopic signatures showing SUMO trees using water from below the soil layer (Grossiord et al., 2017a) support tree root access to groundwater within its seasonal range of depths within the bedrock fractures. We do not know if the bedrock water uptake is by roots alone or if it is facilitated by mycorrhizas. The dominant tree species are piñon pine (Pinus edulis Engelm.) and one-seed juniper (Juniperus monosperma (Engelm.) Sarg.). Intercanopy spaces also contain a small amount of biomass in grass, cacti and Gambel oak. The 0.4 ha experimental site is surrounded by piñon that died during a 2002–2003 drought. Mean annual temperature is 10.1°C and mean annual precipitation is 360 mm (1987–2016 mean), with c. 50% falling during the North American Monsoon season from July to September (http://environweb.lanl.gov/weathermachine).
In June 2012, open-top chambers holding air temperature at c. 4.8°C above ambient and clear polymer troughs that excluded 45% of precipitation were installed to establish heat, drought, drought + heat, and untreated (or ambient) treatments for both piñon and juniper. In April 2016, the trough coverage of the drought and drought + heat structures was increased to 90% to simulate extreme drought conditions (Grossiord et al., 2017b). Each species treatment had five to six trees with tree ages of 56 ± 5 and 79 ± 7 yr for piñon and juniper, respectively, determined from increment cores, and heights between 1.5 and 4.5 m. Micrometeorological conditions were measured continuously and recorded by two weather stations at the site (Climatronics, Bohemia, NY, USA). Atmospheric temperature and relative humidity were measured in all chambers using C215 Campbell sensors (Campbell Scientific, Logan, UT, USA) at two positions (1 m height and two-thirds of canopy height), and used for controlling industrial-scale air-conditioning units that regulated chamber temperature. No tree mortality occurred over the full study period (2012–2016). More details on the study site are provided in Adams et al. (2015).

Model development

The new modeling framework was integrated into TREES (Mackay et al., 2015), which computes soil–plant hydraulics, photosynthesis, canopy diffusive conductance (stomatal, boundary), respiration, and nonstructural C (NSC). TREES has been well tested to show that it can predict seasonal time series of leaf water potentials, soil water content, and canopy transpiration with input of hydraulic parameters from a single day (Mackay et al., 2015; Tai et al., 2017; Johnson et al., 2018). It is forced with observed meteorological data (temperature, wind speed, vapor pressure deficit, photosynthetically active radiation, and precipitation), and constrained with measured parameters on hydraulics, gas exchange, allometry, and root structure. TREES retains a memory of cavitation in each xylem segment, which means the maximum hydraulic conductance declines with successive droughts. Hydraulic conductance values can be recomputed from soil water potential if observed plant hydraulic status (e.g. leaf water potentials) supports a post-drought recovery (Mackay et al., 2015).

In the original modeling framework, root area per unit ground area was defined in each soil-root layer as a fixed multiple of leaf area index (LAI). LAI was dynamic, based on the allocation of C to the canopy and a simple leaf area phenology routine (Savoy & Mackay, 2015), but root area was computed by multiplying LAI by a constant root-to-leaf area scalar (Savoy & Mackay, 2015). Carbon allocation throughout the plant was coordinated with whole-plant hydraulic conductance, \( k_p \). If \( k_p \) was < 50% of its saturated value, then growth allocation declined at a greater than linear rate and maintenance respiration declined at a linear rate. Here we replaced the constant root-to-leaf area scalar with transient root area computed as a function of C allocated to fine roots and the hydraulic status of each soil-root layer. Details of this new model are discussed in the following.

Root architectures are branching networks (Fig. 1a), in which the two finest root orders (i.e. first order is the finest) account for most of the absorbing root area (McCormack et al., 2015). A branching network is cumbersome to model without detailed below-ground observations, and so here we represent the root system as a series of root layers assuming well-mixed soil water content in each layer (Fig. 1b). Each layer was defined as a vector of root orders, represented as cylinders (Fig. 1c), with orders ranging in diameter from low (first order) to high (\( n^{th} \) order). The soil-root volume was organized into \( m \) layers, each with a vertical thickness given by an input root axial length, \( L_{ax} \) and horizontal extent defined by a lateral root length, \( L_{lat} \). The volume of soil occupied by the root system (Fig. 1b) was computed as:

\[
V_R = \pi \left( 1 - \frac{\rho_{bulk}}{\rho_{max}} \right) \sum_{j=1}^{m} l_{Lax}^2 L_{axj} \tag{Eqn 1}
\]

where \( \rho_{bulk} \) is soil bulk density (g cm\(^{-3} \)) and \( \rho_{max} \) is the maximum density (2.65 g cm\(^{-3} \)). For example, a soil-root layer with axial and lateral root lengths of 0.1 m and 1.0 m, respectively, and a soil bulk density of 1.325 g cm\(^{-3} \), has a pore volume of 0.157 m\(^3\). We assumed root diameter scales linearly with increasing root order between the first-order root and the root collar. Each soil-root layer contained a vector of \( n \) (= 10) root orders of diameter, \( d_{Rk} \), with a range defined between a minimum diameter parameter, \( d_{min} \) (mm) and a maximum diameter parameter, \( d_{max} \) (mm), of the root collar or tap root. The diameter of each root order was computed as:

\[
d_{Rk} = d_{min} A_{Rmult}^{k-1} \tag{Eqn 2}
\]

where

\[
A_{Rmult} = \left( \frac{d_{ax}}{d_{min}} \right)^{\frac{1}{2}}
\]

is the ratio of the diameter of the \( k^{th} \) root order to the diameter of the \( (k-1)^{th} \) root order. A root collar diameter of 40.0 mm and minimum root diameter of 0.125 mm yields \( d_{Rk} = (0.125, 0.237, \ldots , 40.0) \text{ mm} \).

Total root surface area per unit ground surface area, \( A_R \) (m\(^2\) root m\(^{-2}\) ground area), was computed from the amount of root C, \( C_R \) (g C m\(^{-2}\) ground area), through a linear relationship between root length and root C, given by specific root length, \( l_s \) (m g\(^{-1}\) C). Total root area for a plant was computed as the summation of the root areas over all layers and root orders, assuming that each \( j, k \) root had the surface area equivalent of the cylinder (Fig. 1c) defined by the root circumference times its length,

\[
A_R = \pi \sum_{j=1}^{m} \sum_{k=1}^{n} l_{tij} C_{Rk} d_{Rk} \tag{Eqn 3}
\]

where

\[
l_{tij} = \frac{l_{axj}}{d_{Rmult}}
\]

and \( l_{tij} \) were obtained empirically from fine root length divided by dry mass of typically the finest one or two root orders. Note also that root tissue density (or C cost per unit root volume) can be computed as \( (d_{Rk} l_{tij})^{-1} \) (Ostonen et al., 2007) with appropriate unit conversions. Assuming root dry biomass is 50% C, then if \( l_{tij} = 200 \text{ m g}^{-1} \text{ C}, d_{min} = 0.125 \text{ mm}, d_{max} = 40 \text{ mm}, \) and each root order is allocated 10 g C m\(^{-2}\) ground area, the root area sequence would be (0.785, 0.414,
0.218, . . . , 0.0025) m² root area m⁻² ground area. In this example, the first three root orders comprise 86% of $A_R$. The value of $d_{max}$ may not be readily available from data, and so it is useful to note that with 10 root orders the contribution to $A_R$ from the first three orders is insensitive to $d_{max}$, especially for values $> 20$ mm.

Carbon for growth and maintenance is taken from net photosynthesis loaded into a pool of NSC. Maintenance respiration is taken from NSC before calculation of growth allocation, as explained in Mackay et al. (2015). Here the response rate of maintenance respiration to temperature was linearly scaled from the root value in the first-order roots to the stem rate in the 10th order roots to yield comparatively higher maintenance costs in finer roots (Pregitzer et al., 1998; Makita et al., 2012). Root area dynamics were integrated with C allocated to roots, $C_R$, with the following simple first-order equation with respect to time $t$:

$$\left(\frac{dA_R}{dt}\right)_{jk} = \left(\frac{dC_R}{dt}\right)_{jk} \pi d_{Rj} l_{Rj}$$  \hspace{1cm} \text{Eqn 4}

The rate of change of C in each root represents a balance between the allocation of new C and root mortality. We assumed that the allocation of new C to each soil-root layer was proportional to the hydraulic health of the respective layers. This is supported by empirical evidence of declining root growth with soil water potential (Teskey & Hinckley, 1981), patchy root growth associated with areas of higher soil water (Hendrick & Pregitzer, 1996), and root growth dynamics associated with soil wetting and drying cycles (Joslin et al., 2000). Here plant access to water was limited by xylem hydraulic conductance ($k_R$) (mmol m⁻² MPa⁻¹) of the absorbing root of each soil-root layer, and roots growing in layers with $k_R$ values closer to their maximum (or saturated) values were assumed to be preferentially allocated C:

$$\left(\frac{dC_R}{dt}\right)_{jk} = \frac{dC_{Rj}}{dt} f_k(k_R) r_k - C_{RDk}$$  \hspace{1cm} \text{Eqn 5(a)}

where

$$f_k(k_R) = \frac{(k_R/k_{Rsat})_j}{\sum_{k=1}^{10} (k_R/k_{Rsat})_k}$$  \hspace{1cm} \text{Eqn 5(b)}

is relative hydraulic conductance, $k_{Rsat}$ is the maximum xylem hydraulic conductance of the root layer, and $r_k = 0.19 - 0.02 (k - 1)$, where $\sum_{k=1}^{10} r_k = 1$. The constants, 0.19 and 0.02, yield a linear decline in C allocation with increasing root order. TREES computes unique $k_{Rsat}$ values for each root segment using observed well-watered hydraulic properties (i.e. transpiration, predawn and midday water potentials) for the whole plant (Mackay et al., 2015). Use of $r_k$ means that C is allocated proportionally to each root order such that finer roots cost more C, reflective of their faster turnover rate (Joslin et al., 2006; McCormack et al., 2013; Adams et al., 2013). $C_{RD}$ is root C loss to mortality, computed as follows:

$$C_{RDk} = f_T(T_R) \frac{C_{Rj}}{\tau_{min}^\beta} k_T, T_R > 5$$  \hspace{1cm} \text{Eqn 5(c)}

where $f_T(T_R) = \frac{T_R - 5}{5}$, $T_R$ is root temperature (°C), $\tau_{min}$ is minimum root life span, and $\beta$ is the life span rate increase with each increment in root order. This results in faster turnover in lower root orders compared with higher root orders (Joslin et al., 2006; McCormack et al., 2013; Adams et al., 2013). $\beta$ could be derived empirically, but it was not available for this study and so we set $\beta = 1.25$, a value comparable to those found empirically (e.g. McCormack et al., 2013). Root mortality was assumed to be zero until root temperature exceeds 5°C (Kitajima et al., 2010), thereafter increasing linearly (McCormack & Guo, 2014) until root temperature reached 25°C (Kitajima et al., 2010). For $\tau_{min} = 0.75$ yr and $T_R = 25^\circ$ the life span for the second-order root is 0.94 yr.

Simulations were run for both juniper and piñon at 30 min time steps from 1 January 2012 to 31 December 2016, using meteorological forcing developed for each treatment (ambient,
drought, heat, drought + heat) as follows. For drought treatments (i.e. drought and drought + heat) we reduced precipitation by 45% of ambient beginning in the middle of 2012 and then by 90% of ambient in year 2016 to match the SUMO field experiment (Fig. S2). For heat treatments (i.e. heat and drought + heat) we increased air temperature from ambient by 4.8 °C beginning in the middle of 2012. We used measurements to increase shallow (0–15 cm) soil temperature by an average of 3.2 °C. Deep soil (15–65 cm) temperatures were computed with a 30 d moving average of shallow soil temperature. Heated treatment deep soil temperature averaged 1.2 °C above ambient. Vapor pressure deficit of the air was adjusted from ambient using the increased air temperature and the standard Clausius–Clapeyron equation.

Most model parameters were either site-specific or species-specific taken from the literature (Table S1). Observations of leaf area, leaf gas exchange and water potentials were aggregated from individual trees, weighted according to sapwood areas, in each treatment to yield treatment mean values (see Notes S1 for details on these measurements). Fine root diameters and specific root lengths were taken from Pregitzer et al. (2002). Root C was distributed among two shallow soil layers (Lshallow = 0–5 cm and 5–15 cm, Lshallow = 300 cm), a deep soil layer (Ldeep = 15–65 cm depth, \( L_{\text{deep}} = 250 \) cm), a taproot (Ltap = 65–290 cm depth, \( L_{\text{tap}} = 0 \) cm), and a bedrock layer with steady groundwater availability (Lbedrock = 290–300 cm depth, \( L_{\text{bedrock}} = 120 \) cm). The main purpose of the taproot was to define dimorphic root architectures with soil layers that were hydraulically separated from the groundwater source at the bottom of the bedrock layer, as opposed to fibrous root architectures with a continuous distribution of fine roots down to the groundwater. In each layer we initialized root C to achieve a root area:leaf area ratio, \( R_{\text{stl}} \), of 2.5 for juniper and 1.7 for piñon, consistent with a strategy of juniper preventing root cavitation by reducing the water uptake rate per unit root area (Sperry et al., 2002; West et al., 2008).

At SUMO, the empirical leaf water potential data for both juniper and piñon showed evidence of cavitation reversal during monsoons in years 2012–2015. On the basis of these observations, we forced TREES to reset the xylem water status in the simulated trees (Mackay et al., 2015) at the times of the apparent cavitation reversals. This occurred during monsoon each year, except for 2016 when there was a weak monsoon (see Table S2). Changes made to leaf and root areas since the last refilling event were accounted for in the computation of maximum hydraulic conductance (Sperry et al., 1998; Mackay et al., 2015).

### H1: Bedrock groundwater source acquisition

To test the hypothesis H1 that tree hydraulic status is maintained by taking up bedrock groundwater during dry periods and soil water during wetter periods, we used two alternative root area schemes, one with high root area in the bedrock layer and the other with low bedrock root area (Table 1). For the high bedrock root area scheme, we adjusted the C content of the bedrock layer roots to obtain sufficient root area with bedrock groundwater access to match simulated and measured predawn water potentials (Johnson et al., 2018) in the first year of simulation (2012), with the other 4 yr used as a check on the model. Thus, for the high bedrock root area, 14.4% and 11.1% of the initial root area for juniper and piñon, respectively, were supplied with bedrock groundwater. For simulations with low initial bedrock root area (Table 1), the C from the first five root orders in the 290–300 cm layer was redistributed equally among the first five root orders in the soil layers. Thus for the low bedrock root scheme, only 1% of the total initial root area had access to bedrock groundwater. In both schemes, root areas were allowed to adjust dynamically according to Eqn .

### H2: Rates of root growth

To test hypothesis H2, that fine roots must be maintained to access bedrock groundwater because they cannot grow fast enough under drought and heat stress, we quantified how simulated roots were able to grow in the low initial bedrock root area scheme relative to their ability to affect plant water relations. We also quantified the effects of plant water relations and treatment on root growth rate, noting that root growth occurs in all layers as long as their hydraulic conductance is nonzero, and growth occurs in all root orders. The link between plant water relations and root growth rate is via root C cost per unit increment of root area (Eqn ) and depends on parameters \( d_{\text{min}} \) and \( l_{\text{st}} \).

### Sensitivity analysis

The test for H2 hinges in part on understanding the sensitivity of computed root growth to \( d_{\text{min}} \) and \( l_{\text{st}} \). The species-specific parameters taken from Pregitzer et al. (2002) yield very high root tissue densities, and, in turn, high C costs to grow roots. Yet there is potentially a large variability in C costs with and between taxa (Fig. S1), including taxa from which root parameters might be obtained if species-specific ones, such as those used here from Pregitzer et al. (2002), are unavailable. Consequently, to understand how the choice of \( d_{\text{min}} \) and \( l_{\text{st}} \) affects simulated fine root growth, and place the simulations for juniper and piñon in a larger context, we computed bin averages of parameters based on diameter extremes as well as taxonomic classes (Table 2) from

### Table 1 Root area (\( A_{\text{st}} \)) distributions for the high and low bedrock root area schemes by species at initialization of the simulations.

<table>
<thead>
<tr>
<th>Species</th>
<th>Root layer</th>
<th>Depth range (cm)</th>
<th>High bedrock ( A_{\text{st}} ) (% of root area)</th>
<th>Low bedrock ( A_{\text{st}} ) (% of root area)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Juniper</td>
<td>Shallow soil</td>
<td>0–5</td>
<td>12.0</td>
<td>15.9</td>
</tr>
<tr>
<td></td>
<td>5–15</td>
<td>17.6</td>
<td>24.9</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Deep soil</td>
<td>15–65</td>
<td>55.8</td>
<td>58.0</td>
</tr>
<tr>
<td></td>
<td>Tap root</td>
<td>65–290</td>
<td>0.2</td>
<td>0.2</td>
</tr>
<tr>
<td></td>
<td>Bedrock</td>
<td>290–300</td>
<td>14.4</td>
<td>1.0</td>
</tr>
<tr>
<td>Piñon</td>
<td>Shallow soil</td>
<td>0–5</td>
<td>12.2</td>
<td>16.0</td>
</tr>
<tr>
<td></td>
<td>5–15</td>
<td>24.3</td>
<td>25.0</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Deep soil</td>
<td>15–65</td>
<td>52.2</td>
<td>57.7</td>
</tr>
<tr>
<td></td>
<td>Tap root</td>
<td>65–290</td>
<td>0.2</td>
<td>0.3</td>
</tr>
<tr>
<td></td>
<td>Bedrock</td>
<td>290–300</td>
<td>11.1</td>
<td>1.0</td>
</tr>
</tbody>
</table>
Table 2. Fine root parameters for computing root dynamics in TREES are root diameter (\(d_{\text{min}}\)), specific root length (\(l_{\text{s}}\)) and life span (\(\tau_{\text{min}}\)), representing the life span of the roots at \(d_{\text{min}}\). \(E(d_{\text{min}})\) is expected value of the minimum diameter.

<table>
<thead>
<tr>
<th>Taxonomic classification</th>
<th>(d_{\text{min}}) (mm)</th>
<th>(l_{\text{s}}) (\text{m g}^{-1} \text{C})</th>
<th>(\text{RTD (g C cm}^{-3}))</th>
<th>(\tau_{\text{min}}) (yr)</th>
<th>(n)</th>
<th>(\tau_{\text{min}}) (yr)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pinaceae</td>
<td>0.47</td>
<td>32.86</td>
<td>0.022</td>
<td>31</td>
<td>0.89</td>
<td></td>
</tr>
<tr>
<td>Cupressaceae</td>
<td>0.68</td>
<td>14.69</td>
<td>0.024</td>
<td>6</td>
<td>1.29</td>
<td></td>
</tr>
<tr>
<td>Sclerophyllus</td>
<td>0.44</td>
<td>30.00</td>
<td>0.027</td>
<td>1</td>
<td>0.84</td>
<td></td>
</tr>
<tr>
<td>Pinus edulis</td>
<td>0.40</td>
<td>22.00</td>
<td>0.045</td>
<td>1</td>
<td>0.76</td>
<td></td>
</tr>
<tr>
<td>Juniperus monosperma</td>
<td>0.39</td>
<td>23.00</td>
<td>0.045</td>
<td>1</td>
<td>0.74</td>
<td></td>
</tr>
</tbody>
</table>

The parameters are derived by taking the means from taxonomic classes in a database of c. 900 entries, which was assembled from the literature for this study. (Note that there is also a published database on fine root traits, FRED (https://roots.orni.gov/). The number (\(n\)) of samples from the database is shown. Minimum life span was computed as a linear function of RTD, \(\tau_{\text{min}} = 0.75 \times 0.395/d_{\text{min}}\). Root tissue density, RTD = \((d_{\text{min}} \times l_{\text{s}})^{-1}\) with appropriate unit conversions. The extreme diameter (0.87 and 0.18 mm) classes were computed from all evergreen trees in the database (\(n = 180\)).

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published parameter values for a large range of species (Dataset S1). We chose parameters for the respective families (Pinaceae and Cupressaceae) and a Sclerophyllous group to represent climatic adaptation. Extreme large diameters \((d_{\text{min}} > E(d_{\text{min}}) + \text{SD}(d_{\text{min}}))\), where \(E(d_{\text{min}})\) is expected value of the minimum diameter, were obtained by selecting all ‘evergreen trees’ entries that had diameters larger than the mean + 1 SD. The extreme minimum values \((d_{\text{min}} < E(d_{\text{min}}) - \text{SD}(d_{\text{min}}))\) were the subset of evergreen trees with diameters that were smaller than the mean – 1 SD. We parameterized root life span using literature values from studies using minirhizotrons to examine the lifespan of functional roots. We did not find literature values on survival rates specifically for *Pinus edulis* and *Juniperus monosperma* fine roots, but we found studies that reported short (< 1 yr) and long (> 1 yr) fine root life spans (Joslin et al., 2006; Montagnoli et al., 2012), and studies with *Pinus* species with roots of 0.3–0.4 mm diameter having a life span of about 0.75 yr (Withington et al., 2006; McCormack et al., 2013). We used 0.75 yr life span for our 0.395-mm-diameter first-order roots, noting that shorter or longer times would make it costlier or cheaper, respectively, to grow or maintain fine roots. Root life spans were adjusted linearly with diameter relative to the baseline 0.395. Simulations were run for 5 yr at half-hourly time steps on the ambient treatment.

We also considered the sensitivity of \(A_R\) to choice of series used to compute \(n_k\). For example, an alternative series described by the recursive function, \(n_k = 0.5 r_{k-1}\), where \(r_1 = 0.5\), would allocate 75% of root C increment to the first two root orders, compared with 36% allocated using Eqn 5(a–c). The higher allocation of C to the first two root orders would yield a modest 76% increase in \(A_R\) increment compared with that produced with Eqn 5(a–c) when the species-specific \(d_{\text{min}}\) and \(l_{\text{s}}\) (Pregitzer et al., 2002) are used. By comparison, using the extreme minimum \(d_{\text{min}}\) and \(l_{\text{s}}\) instead of species-specific \(d_{\text{min}}\) and \(l_{\text{s}}\) would triple the \(A_R\) increment. As \(A_R\) was more sensitive to the choice of parameters than to the choice of \(n_k\) series, we report sensitivity analysis based solely on the \(n_k\) series in Eqn 5(a–c) and parameter sets selected from Table 2.

### Results

**H1: Bedrock groundwater source acquisition**

Modeled trees with high initial bedrock root area closely followed the predawn (Fig. 2) and midday water potentials of experimental trees (Fig. S3). Simulated canopy transpiration (\(E_C\)) closely followed the dynamics of \(E_C\) from sap flux data collected in 2016 (year 5) (Fig. S4). Predawn water potentials remained above −5 and −3 MPa, respectively, for juniper and pinyon, in all treatments. Modeled trees with low initial bedrock root area had larger declines in predawn water potentials, with ambient juniper and pinyon predawn water potentials reaching −9.5 and −4.0 MPa, respectively, in the first year of simulation. These lower predawn values were maintained for longer periods in the drought, heat and drought + heat treatments, and the values remained lower for the full 5 yr simulation. With low initial bedrock root area, both species also showed increasing time spent with percentage loss of conductance (PLC) in excess of 60%, a potential threshold for hydraulic failure (Adams et al., 2017), over the 5 yr of simulation when exposed to increasing drying (drought + heat) drought + heat > drought + heat > ambient (Table 3). Juniper had a notable increase in time spent with PLC > 60% in the low bedrock root area scheme, whereas the total time pi spent at PLC > 60% was insensitive to the initialization of bedrock root area.

For the high initial bedrock root area schemes, the proportion of \(E_C\) supplied by bedrock groundwater ranged, among years, from 5.5% to 19.4% for ambient juniper and from 5.3% to 16.6% for ambient pinyon. These numbers increased to 24.1–57.1% for drought + heat juniper and 20.2–52.5% for drought + heat pinyon. Both species showed similar patterns in shallow soil and bedrock \(E_C\) water source, but during the season when there were large interspecific differences in water uptake source it was because juniper took up more bedrock groundwater and pinyon took up more shallow soil water (Fig. 3). Alternatively, pinyon generally supplied a larger proportion of \(E_C\) from shallow soil, especially during the summer/autumn following monsoon. Bedrock groundwater uptake rapidly shifted between near zero immediately after rainfall events and high values during dry periods (Fig. S5). Bedrock groundwater uptake was high during drier periods (e.g. the first 500 d of simulation (Fig. S2)) and low during wetter periods (e.g. in the years 2014 and 2015). Trees initialized with low bedrock root area slowly increased groundwater uptake over time (Fig. 3; compare 2016 with 2012, red vs blue lines), although this was slower for drought + heat than for drought.

**H2: Rates of root growth**

The extreme differences in leaf water potentials between rooting schemes were mediated by year 5 in the ambient and heat treatments, but remained pronounced in the drought and drought + heat treatments (Fig. 2). The slow adjustment in water
relations follows a slow rate of change in root:leaf area ratios (Fig. 4). Shallow soil and bedrock layers maintained near steady-state root:leaf area ratios \( (R_{R/L}) \) when initialized with high bedrock root area in the ambient and drought treatments, and trended downward in heat and heat + drought treatments. Net fine root growth in the first two root orders in the bedrock layer declined with declining leaf predawn water potential, reaching zero net growth at \(-7.0 \) MPa for juniper and \(-3.5 \) MPa for piñon (Fig. 5). Mean growth rates were reduced by at least 50\% in the drought and heat treatments relative to ambient. For the heat + drought treatment, there was negligible net fine root growth, except for juniper at a water potential of \(-1.0 \) MPa.

Both species showed lower mean NSC by the year 5 of simulation when exposed to increased drying (drought + heat > drought > heat > ambient) (Table 3). No simulation resulted in exceptionally low NSC, however, with values never deviating outside a range of 3–8\% of structural C. Simulations initialized with low bedrock root area had higher NSC amounts than corresponding simulations that were initialized with high bedrock root area.

Sensitivity analysis

Root growth was sensitive to choice of parameter sets for \( d_{\text{min}}, l_{\text{nl}} \) and root life span (Fig. 6). For simulations using species-specific parameters, shallow soil and bedrock layers maintained steady-state root areas, while the deep soil layer, which received little infiltrated precipitation, had downward trends in root areas for both species. The use of species-specific parameters resulted in relatively small short-term variability in root areas, and by implication low root growth rates, compared with growth rates obtained using the other parameter sets. At the other extreme, the use of the extremely large diameters \( [d_{\text{min}} > E(d_{\text{min}}) + SD (d_{\text{min}})] \) parameter set resulted in greater than linear root area growth with time, which led to \( R_{R/L} \) values that were doubled relative to those obtained using species-specific parameters, with no appreciable change in transpiration (i.e. 23 mm increase for juniper, and 16 mm increase for piñon over 5 yr).

Discussion

Overarching hypotheses

During dry periods, the SUMO juniper and piñon trees maintained their hydraulic status by taking up water from bedrock sources, as shown by the simulations in this study, with stable isotopes in a previous study at SUMO (Grossiord et al., 2017a), and with piñon and another juniper species in Utah, USA (West et al., 2007a). Hypothesis H1 was supported because simulated trees took up bedrock groundwater when soil layers were dry, but
switched to soil water uptake following rain events. Hypothesis H2 was also supported because the rates of fine root growth were slow relative to the rate of demand of bedrock groundwater needed to prevent potentially lethal xylem water potentials. Our results imply that these trees must grow into bedrock water sources before drought in order to survive extreme conditions.

H1: Bedrock groundwater source acquisition

During the driest periods, a small fraction of total root area allocated to the bedrock supplied > 60% of transpiration for all treatments, whereas during wetter periods (e.g. ambient treatment, second half of 2013–2015) trees took up negligible bedrock groundwater (Figs 3, 5S). Under drought and drought + heat treatments, juniper increased water uptake from bedrock groundwater relative to piñon, particularly during winter and spring, whereas under heat and drought + heat treatments, piñon responded by increasing water uptake from shallow soils during and after monsoons more than was the case for juniper. This is consistent with experimental observations for both species at SUMO using isotopes of water (Grossiord et al., 2017a) and for piñon in other studies (West et al., 2007a,b). Juniper was more anisohydric (i.e. had more daily variation in water potential; Tardieu & Simonneau, 1998) than piñon, dried out the shallow soil layers, and thus depended on more deeply infiltrating winter rainfall as well as bedrock groundwater, as has been shown previously (Plaut et al., 2013; Grossiord et al., 2017a). These contrasting trait responses follow directly from the species’ contrasting xylem cavitation vulnerability curves (Mackay et al., 2015; Garcia-Forner et al., 2016).

Modeled trees spent an increasing amount of time at PLC > 60% with intensity of treatment (drought + heat > drought > heat > ambient) (Table 3). Piñon had PLC values that potentially predisposed them to drought-induced mortality (Adams et al., 2017) despite maintaining relatively high water potentials compared with simulations of warm droughts that did not consider bedrock groundwater (McDowell et al., 2016). Alternatively, when bedrock groundwater access was reduced, modeled trees in all treatments experienced potentially lethal predawn water potentials for extensive periods of time (Fig. 2), which were well below thresholds used previously to predict mortality (i.e. April–August mean water potentials of −5.3 MPa for juniper and −2.4 MPa for piñon; McDowell et al., 2016). Juniper responded to an impaired bedrock groundwater with a notable increase in time spent at high PLC, whereas piñon showed no such response. This follows from piñon’s greater ability to respond to small rainfall events that penetrate only shallow soil water layers at SUMO (Grossiord et al., 2017a) and other locations (West et al., 2007a,b; Plaut et al., 2013).

H2: Rates of root growth

Root area : leaf area ratio, R/L, adjusted slowly (Fig. 4) because of slow root growth (Fig. 5), requiring seasons to years to change water-uptake rates. The slow root growth was not attributed to a lack of C resources, as the biggest decline in NSC was 26% relative to the least-stressed treatment (i.e. ambient initialized with high root area in the bedrock) (Table 3). This was well below a lethal decline (Anderegg & Anderegg, 2013; Adams et al., 2017). Simulations initialized with low bedrock root area maintained relatively higher NSC because the allocation of C to growth declined more rapidly than photosynthesis when PLC was > 50% (Mackay et al., 2015).

We considered the possibility that the slow root growth was an artifact of either model parameterization or an overly conservative algorithm for determining the allocation of NSC to growth. Two lines of evidence suggest there were no such artifacts. First, parameters that favored the highest growth rates would not have allowed root area to increase enough to mediate the effects of drought during the relative dry first 500 d of simulation. Second, the parameter set that was most favorable for high root growth rates would have more than doubled R/L by the year 5 of simulation (Fig. 6), which most certainly would have overbuilt the root areas for piñon and juniper (West et al., 2008), resulting in suboptimal water transport (Sperry et al., 2002). No matter what root trait parameter values were used, the fine root growth would not do the job of maintaining tree hydraulic status after the onset of drought. Moreover, a more aggressive use of NSC for growth allocation that did not overbuild roots would have increased leaf area, contrary to observations in the experimental treatments.
The results here also lend mechanistic support to explain tree recovery post-drought. Prolonged drought inhibits the ability of trees to recover when drought is relieved, and this has been attributed to a lack of C sequestered during drought or to declining C reserves (Trugman et al., 2018). The simulations here add an additional explanation for the observed survival or mortality of piñon-juniper woodlands, in particular, and woody systems in general, which is that fine root growth is too slow to allow for rapid tree hydraulic recovery after drought even if drought does not cause a substantial decline in C reserves. Use of existing roots with access to reliable water during drought obviates the need for rapid root growth, and offers an explanation for the rapid shifts seen in water-uptake sources during drought and rapid return to shallower layers following precipitation events (Dawson, 1993; Burgess et al., 1998; Joslin et al., 2000; Barton & Montagu, 2006; Metcalfe et al., 2008; Bleby et al., 2010). The simulations here presume that such rapid changes are passive responses to pressure gradients. The underlying mechanisms are likely to be more complex, such as an active control over root water uptake via refilling of shallow-rooted piñon roots (West et al., 2007a) or expression of aquaporins (Gambetta et al., 2012; Venturas et al., 2017). Rapid fine root growth (Barnes, 2002; Laclau et al., 2013) may be restricted to more productive systems than piñon-juniper woodlands, in actively developing plants that may be more able to alter their $R_R/L$ than mature individuals.

**Broader implications**

As we gain an improved understanding of abiotic controls over the heterogeneity of reliable water sources to support tree survival under climate change-type droughts (McLaughlin et al., 2017), there is a need for a complementary understanding of the physiological responses to these water sources. The model presented here addresses a significant knowledge gap in understanding such
biotic–abiotic processes because it explicitly links C allocation and plant hydraulics (Fisher et al., 2018; Hartmann et al., 2018). This model was used to explain why trees frequently maintain deep coarse root systems with only small amounts of absorbing fine roots (Laclau et al., 2013; Pinheiro et al., 2016) to obtain a high proportion of water from deeper groundwater sources during drought (David et al., 2007; Miller et al., 2010; Grossiord et al., 2017a). The simulations also provide a mechanistic

Fig. 4 Simulated root : leaf area ratios ($R_{R/L}$) for juniper (a) and piñon (b), initializing TRees with high (14.4% for juniper, 11.1% for piñon) and low (1% for both species) percentage of the root area with access to bedrock groundwater. The seasonal cycle is dominated by the change in leaf area associated with leaf phenology. Total $R_{R/L}$ by plot can be computed by summing shallow, deep, and bedrock values at a given point in time.

Fig. 5 Simulated relative rate of growth of juniper (a) and piñon (b) first- and second-order roots (summed) in the bedrock layer as a function of mean predawn water potential. Results are for simulations that were initialized with 1% of fine roots in the bedrock and then roots were allowed to grow. Open circles and dashed lines had reduced precipitation, and red circles represented heat treatment. Growth was computed as the difference in sum of first- and second-order root carbon content between consecutive days. The mean of these results was then computed for each bin of predawn water potential, with steps of 0.4 and 1.0 MPa for piñon and juniper, respectively.
explanation for the lack of fine root growth seen during drought (Joslin et al., 2000; Metcalfe et al., 2008), and supports the observation that trees make sufficient C investments in fine roots before drought, enabling them to gain access to reliable water for survival during drought (Jackson et al., 1999; Rose et al., 2003; Eberbach & Burrows, 2006; Johnson et al., 2018). The new modeling approach could be adapted for use with other hydraulically sensitive ecosystem models, and tested in a wide range of systems, including where roots do not have access to reliable water sources. The modified TREES model demonstrates how fine root growth could be integrated into ecosystem models (McCormack et al., 2015) and answers a broader call for constraining ecosystem models with observations (Law, 2014).

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Author contributions

DSM led the model theory development, developed the model, and wrote the manuscript; PRS assisted with the model theory development and edited the manuscript; JRP, XT and DRW assisted with the model development and edited the manuscript; CG, HDA and NGM designed the field experiment, conducted fieldwork and edited the manuscript; JSS helped with conceptual design of the model and edited the manuscript.

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References


Supporting Information

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

Dataset S1 Fine root data compiled for species spanning a wide range of families and life forms.

Fig. S1 Specific root length and root diameter data based on first-and second-order fine roots.

Fig. S2 Daily total precipitation and midday vapor pressure deficit for ambient, drought and heat treatments.

Fig. S3 Simulated vs observed water potentials by species, plot treatment, for predawn and midday simulations.

Fig. S4 Transpiration by species treatment for year 2016, days 61–262, using sap flux data collected at SUMO.

Fig. S5 Rhizosphere flux in the bedrock layer by species and plot treatment.

Notes S1 Data collection methods at SUMO.

Table S1 Parameters used in TReES for each species.

Table S2 Dates used to force xylem refilling by species, year and plot.

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