



# 1 Codominant tree species showcase opposing hydraulic 2 capacitance-reliance in Central Texas

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10 **Abstract.** Understanding the stress-responses of the tree species that comprise ecosystems is important for making  
11 predictions of how those ecosystems will respond to future stress events, e.g., drought. Past work has shown  
12 differential reliance on hydraulic capacitance, water released from internal storage, to maintain daily transpiration  
13 and gas-exchange with the atmosphere. In this study, we used wavelet coherence analysis to quantify the  
14 relationship between water supply (soil moisture), atmospheric water demand (VPD), and tree-water  
15 measurements (wood water content and sap flow) for two codominant tree species in a Central Texas woodland,  
16 escarpment live oak (*Quercus fusiformis*) and Ashe juniper (*Juniperus ashei*), over one and a half years including  
17 a period of extreme drought. Wavelet coherence provides a measure of correlation between the periodicity of  
18 signals at different frequencies or periods. The goal of our study was to analyze how the study trees' responses to  
19 environmental moisture conditions changed over time, especially during drought conditions. Our results revealed  
20 that, for both species, vapor pressure deficit and sap flow were strongly correlated at the 24-hour period, even  
21 during severe drought. Wood water content, especially for *Q. fusiformis*, was strongly correlated with soil  
22 moisture at daily to weeks-long periods. Our results also showed that vapor pressure deficit had a strong, and  
23 sustained, correlation with *Q. fusiformis* wood water content at the 24-hour period as drought severity increased,  
24 while *J. ashei* wood water content did not have this same relationship to vapor pressure deficit. Our results  
25 suggested that both species continue daily transpiration during severe drought, and that *Q. fusiformis* is more  
26 reliant on capacitance than *J. ashei* to maintain daily transpiration leading up to, and during, severe drought  
27 conditions. This difference in drought response strategy has important implications for each species' survival  
28 through shifting hydroclimatic conditions.

## 29 1. Introduction

30 Plant hydraulic responses to environmental conditions associated with water availability are critical to predicting  
31 plant-water status, drought-induced mortality, and the potential vulnerability and risk of wildfire (Ding et al., 2025;  
32 Ma et al., 2021; Powell et al., 2017) as well as understanding exchanges of carbon, water, and energy between the  
33 land and the atmosphere (Bonan, 2015). Over the last 45 years, atmospheric evaporative demand has caused a 40%  
34 global increase in the severity of droughts (Gebrechorkos et al., 2025; Vicente-Serrano et al., 2022), which have  
35 resulted in widespread tree mortality and landscape-level vegetation change due to plant-water stress or combined  
36 disturbances (Breshears et al., 2005; Crouchet et al., 2019; Fauset et al., 2012; Feeley et al., 2012).

37 Trees respond to their environment according to species-specific traits and behaviors. The traits and behaviors that  
38 determine how trees access, move, and store water comprise a plant's hydraulic strategy which includes rooting  
39 characteristics, stomatal regulation, and xylem architecture (Matheny et al., 2017b). Water moves through the soil-  
40 plant-atmosphere continuum along a water potential (MPa) gradient; as water evaporates from stomatal openings at  
41 the leaf surface, water held in tension within the tree is pulled upward from roots through the xylem and finally to the  
42 leaf. As tension increases (i.e., water potentials become increasingly negative), air bubbles can form in xylem vessels



43 (Tyree and Sperry, 1989). This process, known as embolism, causes a blockage which reduces overall hydraulic  
44 conductance. Some species of tree can buffer the effect of very negative water potential by utilizing capacitance,  
45 defined as the amount of water released from storage per change in water potential; capacitance can come from  
46 different sources within the stem including embolized vessels (Hölttä et al., 2009; Scholz et al., 2011). Coexisting  
47 species within an ecosystem often have different or opposing hydraulic strategies (Matheny et al., 2017a; Mcculloh et  
48 al., 2012; Pappas et al., 2018). Such diversity in species traits and behaviors can add to overall forest resilience (Fauset  
49 et al., 2012; Pappas et al., 2018). However, these differences also mean that some species may be more vulnerable to  
50 certain disturbances, depending on hydraulic strategy and species-specific responses to the environmental conditions  
51 that characterize the disturbance, and thus influence the ecosystem-level outcome (Brienen et al., 2015; Feeley et al.,  
52 2012; Matheny, 2021).

53 Ecosystems may exhibit stronger reliance on either vapor pressure deficit (VPD) or soil moisture, given vegetation  
54 type, species composition, or environmental conditions (Kibler et al., 2025; Novick et al., 2016; Sulman et al., 2016).  
55 Understanding vegetation sensitivity and responses to the environment is a complex problem given the natural  
56 interdependence among many controlling variables, such as solar radiation, temperature, VPD, and soil moisture.  
57 Furthermore, the dynamics of changing soil moisture and VPD fluctuations frequently operate at different timescales.  
58 Soil dries slowly over days and weeks, while VPD can change rapidly on the scale of minutes and hours. Yet both  
59 follow general trends over diurnal and seasonal periods. As hot droughts and VPD are expected to increase  
60 (AghaKouchak et al., 2021; Allen et al., 2015; Vicente-Serrano et al., 2022), it is critical to understand how forests,  
61 and the species that comprise them, respond to these stimuli in order to make predictions of vegetation moisture  
62 dynamics and ecosystem transition. Ecosystem-scale measurements such as those from eddy covariance flux towers  
63 and remote sensing products, can reveal how whole ecosystems rely on VPD or soil moisture (Kibler et al., 2025;  
64 Novick et al., 2016; Sulman et al., 2016). However, given the reliance of plant responses on species-specific traits and  
65 behaviors, species with opposing hydraulic strategies within an ecosystem can respond differently to changing weather  
66 patterns, leading to, for example, different drought-induced mortality rates. The seasonality and inherent differences  
67 across timescales of environmental conditions, along with plant responses across those timescales, present a challenge  
68 to disentangling relationships between environment and plant function that cannot be answered by linear regression  
69 alone. Rather, a time series analysis that detects time-lagged plant responses to environmental conditions, and how  
70 those responses change over time is needed.

71 Wavelet analysis, a signal processing technique, may provide new insights into how the relationships between plant-  
72 water and environmental moisture conditions change over time and across timescales. Wavelet transformation allows  
73 visualization of a one-dimensional timeseries in two-dimensional time-frequency, or time-period, space. In doing so,  
74 the frequencies, or timescales, at which a signal is strongest and how this strength changes over time are revealed.  
75 Wavelet transforms have been used in a wide range of Earth science studies, including examining how vegetative  
76 fluxes vary across timescales (Katul et al., 2001), how environmental factors control carbon assimilation in ecosystems  
77 (Stoy et al., 2005, 2009), and how environmental factors control water storage in an individual tree (Harmon et al.,  
78 2021). Wavelet coherence analysis offers a unique way to analyze the correlation between periodicities present in  
79 signals and to calculate time lags between those signals. In this study, we use wavelet coherence to compare how two  
80 codominant tree species in a semi-arid woodland respond to atmospheric demand and available soil moisture. While  
81 wavelet analysis alone is insufficient to identify the mechanisms behind the relationships we are studying (Stoy et al.,  
82 2005), it can nevertheless provide new insights into how these relationships change over time. The central goal of this  
83 analysis is to identify times when VPD or soil moisture dominates the control of plant hydraulic functions, and the  
84 timescales at which these controls occur.

85 This study aims to understand and disentangle the influences of VPD and soil moisture controls on vegetation  
86 hydrodynamics between two coexisting species with specific emphasis on uncovering how water stored in biomass  
87 (i.e., hydraulic capacitance) buffers the demands of these two forcings. We expect that trees with different hydraulic  
88 strategies will differ in their responses to these two drivers of the soil-plant-atmosphere continuum and that the



89 timescales of these responses will be determined by the timescales at which the environmental moisture dynamics  
90 change.

91 Our study site, on the Edwards Plateau in Central Texas, showcases opposing hydraulic strategies in the oft codominant  
92 species *Quercus fusiformis* (escarpment live oak), a diffuse porous angiosperm, and *Juniperus ashei* (Ashe juniper), a  
93 tracheid-bearing gymnosperm (Johnson et al., 2018a; Northup et al., 2022). Because VPD follows a predictable daily  
94 pattern and drives transpiration, we hypothesized that VPD will be correlated with sap flow in both species at the diel  
95 (24-hour) period. However, we expected that the species may stop transpiring as soil moisture decreases and drought  
96 severity increases, leading to a disruption in the diel correlation between VPD and sap flow signals, and that *J. ashei*  
97 sap flow will become less correlated with VPD sooner than *Q. fusiformis* during times of low soil moisture based on  
98 the findings of Northup et al (2022). If drought did not lead to a disruption of diel correlation, one explanation could  
99 be that the trees are utilizing capacitance to maintain transpiration. Thus, we also expected that, during drought, wood  
100 water content would be more correlated – as quantified by wavelet coherence – with VPD at the diel period, as we  
101 expected that the trees may be more reliant on their internal water stores to feed the daily transpiration stream. Given  
102 the longer timescales over which soil moisture changes, we expected wood water content and sap flow to be correlated  
103 to soil moisture at longer timescales.

104

## 105 2. Methods

### 106 2.1. Site and Data description

107 All tree data were collected at the White Family Outdoor Learning Center (WFOLC) located near Dripping Springs,  
108 TX on the Edwards Plateau (30°07'54"N 98°07'07"W). The Edwards Plateau in Central Texas is prone to frequent  
109 meteorological, hydrological, and hot droughts, including the exceptional drought of 2011 which resulted in  
110 widespread tree mortality (Crouchet et al., 2019; Klockow et al., 2018; Moore et al., 2016). The study site is  
111 characterized by very shallow soils (less than 10 cm) underlain by fractured limestone bedrock. It receives an average  
112 of 852 mm of precipitation annually and is considered a semi-arid woodland. Average temperatures range from 5° C  
113 in the winter to 38° C in the summer (Weather Underground, 2024).

114 The site is dominated by *Q. fusiformis* and *J. ashei*. *Q. fusiformis* is a diffuse porous species with xylem that are  
115 relatively vulnerable to cavitation and has been characterized as drought avoidant, which it may achieve by accessing  
116 alternative water sources rather than through stomatal closure (Northup et al., 2022). In contrast, *J. ashei* is a conifer  
117 with narrow cavitation-resistant tracheid and has been characterized as desiccation tolerant (Johnson et al., 2018b;  
118 McElrone et al., 2004; Northup et al., 2022). *Q. fusiformis* has also been found to root deeper into the fractured  
119 limestone bedrock than co-located *J. ashei* (Jackson et al., 1999), which potentially helped reduce drought mortality  
120 rates in *Q. fusiformis* during the notable 2011 drought (Northup et al., 2022). However, we do not have any  
121 measurements of rooting depths at our study site, and both species are known to root several meters into fractured  
122 bedrock.

123 Volumetric wood water content (WWC, m<sup>3</sup> m<sup>-3</sup>) was calculated from dielectric permittivity observations from  
124 capacitance-style soil moisture sensors (TEROS 12, METER Group Inc., Pullman, WA, USA ) using calibration  
125 equations for each species derived according to the methods of Matheny et al. (2017). Sap flow (SF) measurements  
126 (g s<sup>-1</sup>) were made using custom built Granier-style thermal dissipation sap flow sensors (Granier, 1986; Matheny et  
127 al., 2014). Raw sap flow data were filtered by removing measurements outside the expected thermocouple voltage  
128 range (0.2–0.8 mV), periods of sensor drift due to degradation, and anomalous spikes. Measurements were calibrated  
129 according to Oishi et al. (2016). Each of these measurements were made every fifteen minutes throughout the January  
130 2022 to December 2024 study period (CR6 Datalogger, Campbell Scientific, Logan, UT, USA). Sensors were installed  
131 at breast height (~1.37m) on the north face of seven mature *Q. fusiformis* and six *J. ashei* located within a ~40m radius  
132 of one another. Average diameters at breast height (DBH) were 27.1 cm and 13.5 cm for *Q. fusiformis* and *J. ashei*,



133 respectively. Sap flow and volumetric wood water content were then averaged across individuals of each species at  
134 each timestep.

135 Vapor pressure deficit (*VPD*, kPa) data were collected from the Dripping Springs RAWS meteorological station, about  
136 10.6 km from the study site (RAWS USA Climate Archive State Selection Map, 2025). For soil moisture, we used the  
137 0-10 cm estimates from the NLDAS remote sensing package (NLDAS Get Data | LDAS, 2026) since the actual soil  
138 depth at the site rarely exceeds 10 cm in depth. This soil moisture estimate is modeled based on precipitation and a  
139 parameterized soil moisture model (Xia et al., 2014).

140

## 141 2.2. Cross-wavelet transform coherence

142 To test how sap flow and WWC responses to *VPD* and soil moisture changed over time, we used cross-wavelet  
143 transform coherence to analyze the relationships between signals in time-period space. For example, a high coherence  
144 value at the diel period between *VPD* and WWC would indicate that the daily release and refill of WWC is highly  
145 correlated to the diel pattern of *VPD*, thus indicating that capacitance was utilized to supply the transpiration stream.  
146 Similarly, a high diel coherence between *VPD* and sap flow would indicate that transpiration is highly correlated to  
147 the diel pattern of *VPD*, thus stomata are open throughout the day. Cross-wavelet transform coherence is an appropriate  
148 timeseries analysis method given the different timescales at which *VPD* and soil moisture operate and the expected  
149 tree response time (Grinsted et al., 2004). A wavelet transform ( $W_x$ ) of a time series ( $x$ ) is a convolution of the time  
150 series with a wavelet function,  $\psi$ . This convolution is performed for each timestep, and at several scales, which relate  
151 to frequency and are determined according to the sampling frequency of the time series. In this way, a one-dimensional  
152 (time) time-series is transformed into two-dimensional (time-period) space, revealing information about the  
153 frequencies – or periodicities – of the signal and how they change with time. In this work, we used the complex,  
154 continuous Morlet wavelet for extracting information from time series because it has a good balance between time  
155 and frequency resolution (Grinsted et al., 2004).

156 Continuous wavelet transforms were calculated for all the environmental (i.e. *VPD*, soil moisture) and plant-water  
157 (i.e. species-specific *SF* and *WWC*) time series using the Wavelet Toolbox in MATLAB. It is recommended that  
158 wavelet transforms be used on data that is at least near-normally distributed (Grinsted et al., 2004; Torrence and  
159 Compo, 1998). This is not the case for our data. *VPD* and *SF* data had gamma-like distributions, while *WWC* and soil  
160 moisture had multimodal distributions. Probability distributions were fit using a distribution fitting function in  
161 MATLAB. All data were transformed into records of percentiles using their respective distributions' cumulative  
162 density functions as recommended by Grinsted et al (2004). This resulted in distributions with a more rectangular  
163 shape. All time-series were then standardized to zero-mean and unit-variance.

164 We calculated cross-wavelet coherences ( $L_{xy}$ ) as a measure of correlation between wavelet transforms of  
165 environmental and species-specific plant-water time series. Thus, the value of the cross-wavelet coherence at a given  
166 time and period represents how correlated the periodicities of the signals are at that period, at that particular time. In  
167 the following equation,  $x$  represents environmental time series, *VPD* and soil moisture, and  $y$  represents plant-water  
168 time series, *SF* and *WWC*:

$$169 \quad L_{xy}(a, b) = \frac{|\sum_m W_x(a, b)W_y^*(a, b)|^2}{|\sum_m W_x(a, b)||\sum_m W_y(a, b)|} \quad (1)$$

170 Where \* denotes the complex conjugate,  $a$  is representative of the period variable, and  $b$  represents the time variable.  
171 The parameter  $m$  is the number of time-steps used in smoothing the coherence. The value of  $m$  has implications for  
172 significance testing for the wavelet coherence (Ge, 2008). For our analysis, we set  $m$  as a function of frequency such  
173 that for any given period the coherence is given as an average coherence over 5 of those periods (e.g., at the 24-hour  
174 period, coherence is a 5-day average). Significance testing was performed using the wavelet coherence significance



175 test described by Ge (2008), with a significance level of 0.05. A total of eight cross-wavelet coherences were  
 176 performed: two species, two plant-water time series, and two environmental time series.

177 For areas in the time-period space of high and significant coherence, we calculated the phase angles, from which we  
 178 can calculate the time lags between the environmental drivers (VPD and SM) and plant-water responses. The phase  
 179 angle, in radians, is given by

$$180 \quad \theta = \arctan \left[ \frac{\Im (W_{xy}(a, b))}{\Re (W_{xy}(a, b))} \right] + \mu\pi \quad (2)$$

181 Where  $\Im$  and  $\Re$  are the imaginary and real part, respectively. And  $\mu = 0$  when the cross-wavelet transform,  
 182  $W_{xy}(a, b) = W_x(a, b)W_y^*(a, b)$ , is in the first or fourth quadrant of the complex plane,  $\mu = 1$  when  $W_{xy}(a, b)$  is in the  
 183 second quadrant, and  $\mu = -1$  when it's in the third quadrant. The phase angle,  $-\pi \leq \theta \leq \pi$ , represents a time lag  
 184 between the two signals. A phase angle of 0 means the two signals are in phase, and a phase angle of  $|\theta| = \pi$  means  
 185 the two signals are in antiphase. Lag times,  $\Delta t$ , were then be calculated as follows:

$$186 \quad \Delta t = \frac{\theta T}{2\pi} \quad (3)$$

187

188 Where  $T$  is the period (hours) for which the time lag is being calculated. A negative  $\Delta t$  implies that  $x$  (i.e., VPD or  
 189 soil moisture) leads  $y$  (i.e., sap flow or WWC), while a positive  $\Delta t$  implies the opposite.

190 Cross-wavelet coherence gives no indication of the magnitude of signals. For example, we cannot tell from comparing  
 191 cross-wavelet coherence values if sap flow is higher in either species at any given time.

### 192 2.3. Multivariate linear regression

193 We performed a multivariate linear regression analysis to assess the magnitude of the relative contributions of VPD  
 194 and soil moisture to the variation in SF and WWC in *J. ashei* and *Q. fusiformis* following a procedure similar to  
 195 Sulman et al (2016). Unlike Sulman et al (2016), who modeled the ratios of transpiration to a thirteen-year average,  
 196 we had a more limited dataset. Thus, we calculated the ratios of  $SF/SF_{max}$  to obtain relative sap flow. The statistical  
 197 models were fit for six conditions, low, middle, and high conditions for both soil moisture and VPD, as

$$199 \quad \frac{SF}{SF_{max}} = C_1 + C_2 \ln(VPD) + C_3 \Psi_{soil} + C_4 \ln(VPD) \times \Psi_{soil}, \quad (4)$$

198

200 Where  $\Psi_{soil}$  (MPa) is the soil water potential. Soil water potential was used since it is more representative, than soil  
 201 water content, of the water available for root-water uptake within the soil-plant-atmosphere continuum (Sulman et al.,  
 202 2016), and gave better results in the models. To calculate  $\Psi_{soil}$  from soil moisture, we used the Clapp-Hornberger  
 203 water retention curve using parameters from the NLDAS soil map ( $b = 8.17, \Psi_{sat} = 0.263, \theta_s = 0.465$ ). The natural  
 204 logarithm of VPD was used following the example of Sulman et al (2016) and based on previous studies showing a  
 205 linear relationship between stomatal conductance and  $\ln(VPD)$  (Oren et al., 1999). All linear regression models were  
 206 performed with the robust linear model, a least squares model that reduces the effects of outliers, in the Statsmodel  
 207 Python package (Seabold and Perktold, 2010).

208

### 209 2.4. Keetch-Byram Drought Index



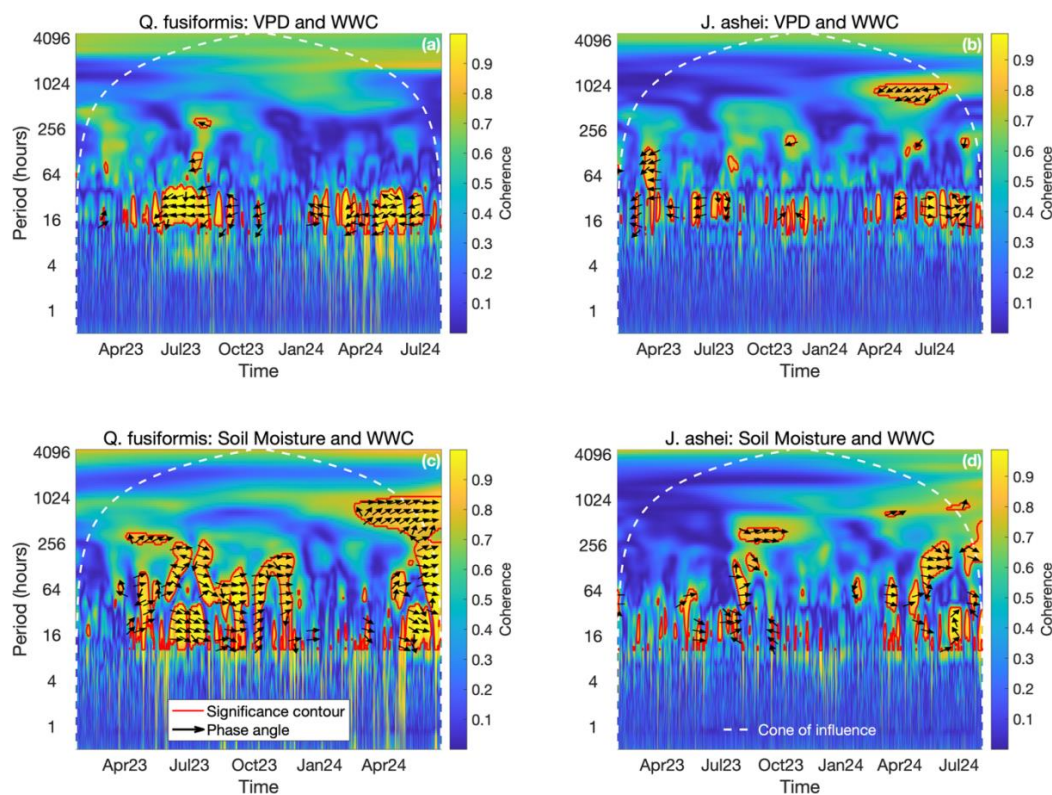
210 We calculated the Keetch-Byram Drought Index (KBDI) to characterize drought severity. We then analyzed how tree-  
211 water measurements, and their wavelet coherence with VPD and SM, evolve with drought severity. The KBDI ranges  
212 from 0 to 800, is used to monitor drought severity (higher values correspond to more severe drought), and is analogous  
213 to soil moisture deficit. It is calculated daily from precipitation, air temperature, and the previous day's KBDI.  
214 According to the Wildland Fire Assessment System (Keetch-Byram Drought Index, 2025) KBDI between 600 and  
215 800 is associated with severe drought and live fuels that contribute to intense fire. McCaw et al (McCaw et al., 2018a)  
216 found that KBDI above 500 corresponded to critically low foliar moisture in *J. ashei* at sites on the Edwards Plateau.

217

### 218 3. Results

#### 219 3.1. Wavelet coherence of WWC

220 *Q. fusiformis* water storage was more responsive to atmospheric moisture during dry soil moisture conditions than  
221 water storage of *J. ashei*, as seen by the longer durations of high coherence (Figure 1.a and b) for *Q. fusiformis* during  
222 the dry summer of 2023. For both species, high coherence between VPD and WWC were centered around the diel  
223 period (Figure 1.a and b). The summer of 2023 was an especially dry, hot summer with average daily high temperature  
224 of 38° C and average daily maximum VPD of 4.8 kPa. By July 15, 2023 KBDI reached 500, which is considered a  
225 threshold above which the foliar moisture content of *J. ashei* becomes critically low (McCaw et al., 2018a) and thus  
226 can be considered a time of water stress for the ecosystem. KBDI stayed at or above 500 through October 26, 2023.  
227 From May through October 2023, soil moisture ranged from a maximum of 0.42 m<sup>3</sup> m<sup>-3</sup> to a minimum of 0.11 m<sup>3</sup> m<sup>-3</sup>  
228 <sup>3</sup>. The minimum soil moisture occurred on August 22. Both *Q. fusiformis* and *J. ashei* demonstrated durations of high  
229 coherence between WWC and VPD at the diel period during this hot, dry summer (Figure 3.b and c) with *Q. fusiformis*  
230 maintaining high coherence for longer durations. Both species also showed a long-term decrease in WWC during this  
231 summer, with *Q. fusiformis* WWC ranging from 0.38 m<sup>3</sup> m<sup>-3</sup> in the spring to 0.23 m<sup>3</sup> m<sup>-3</sup> on September 8, and *J. ashei*  
232 WWC ranging from 0.24 m<sup>3</sup> m<sup>-3</sup> to 0.19 m<sup>3</sup> m<sup>-3</sup> on September 13.



233

234 **Figure 1: Wavelet coherence plots for (a) VPD and *Q. fusiformis* WWC, (b) VPD and *J. ashei* WWC, (c) soil moisture and**  
 235 ***Q. fusiformis* WWC, and (d) soil moisture and *J. ashei* WWC. The closer the coherence is to 1, the more correlated the**  
 236 **wavelet transforms of two time series are for a given period (y-axis) at a given moment in time (x-axis). The dashed white**  
 237 **line indicates the cone of influence, outside of which wavelet coherence values are considered unreliable due to edge effects.**  
 238 **The red contours show areas of significantly high coherence according to the significance test at a 0.05 significance level.**  
 239 **The black arrows indicate the direction of the phase angle. If the arrow is pointing straight to the right, the phase angle at**  
 240 **that time and period is 0; straight up means it's  $\pi/2$ ; straight down means it's  $-\pi/2$ ; straight left means  $\pm\pi$ . For both**  
 241 **species, most of the high coherence of VPD and WWC is centered around the diel period at times of low soil moisture. This**  
 242 **means that the diel cycles of WWC are highly correlated to the diel cycle of VPD at those times. Phase angle arrows mostly**  
 243 **point to the left, indicating a phase angle near  $-\pi$  meaning the signals are in antiphase. Overall, *Q. fusiformis* WWC signal**  
 244 **is more correlated to the soil moisture signal at any period. Phase angle arrows show that soil moisture and WWC signals**  
 245 **are often nearly in phase for both species.**

246 The phase angles (Figure 1.a and b, black arrows) associated with the regions of high coherence around the diel period  
 247 were mostly near  $-\pi$  radians, indicating that the signals were in antiphase. By Eq. (3), this indicates that VPD reaches  
 248 its daily maximum while WWC is around its daily minimum, then WWC has refilled as much as it is going to while  
 249 VPD is reaching its daily minimum (time-lag of approximately 12h).

250 For both species, high coherence between soil moisture and WWC occurred during times of low soil moisture and at  
 251 diel and longer periods. Between June 22 and August 22, 2023, soil moisture declined from  $0.31$  to  $0.11$   $\text{m}^3 \text{m}^{-3}$  (Figure  
 252 3.f). By October 5, rains had increased soil moisture back to  $0.36$   $\text{m}^3 \text{m}^{-3}$  (Figure 3.f). During this time of soil drying  
 253 and subsequent rewetting, wavelet coherence was high at daily to weeks-long periods, especially for *Q. fusiformis*  
 254 (Figure 1.c), suggesting that the *Q. fusiformis* water storage signal was more correlated to shallow soil moisture signal  
 255 during dry conditions than *J. ashei* (Figure 1.d). The phase angles at the sub-daily to daily timescales during the  
 256 summer of 2023 were near zero and negative, indicating that the signals were nearly in phase, with soil moisture



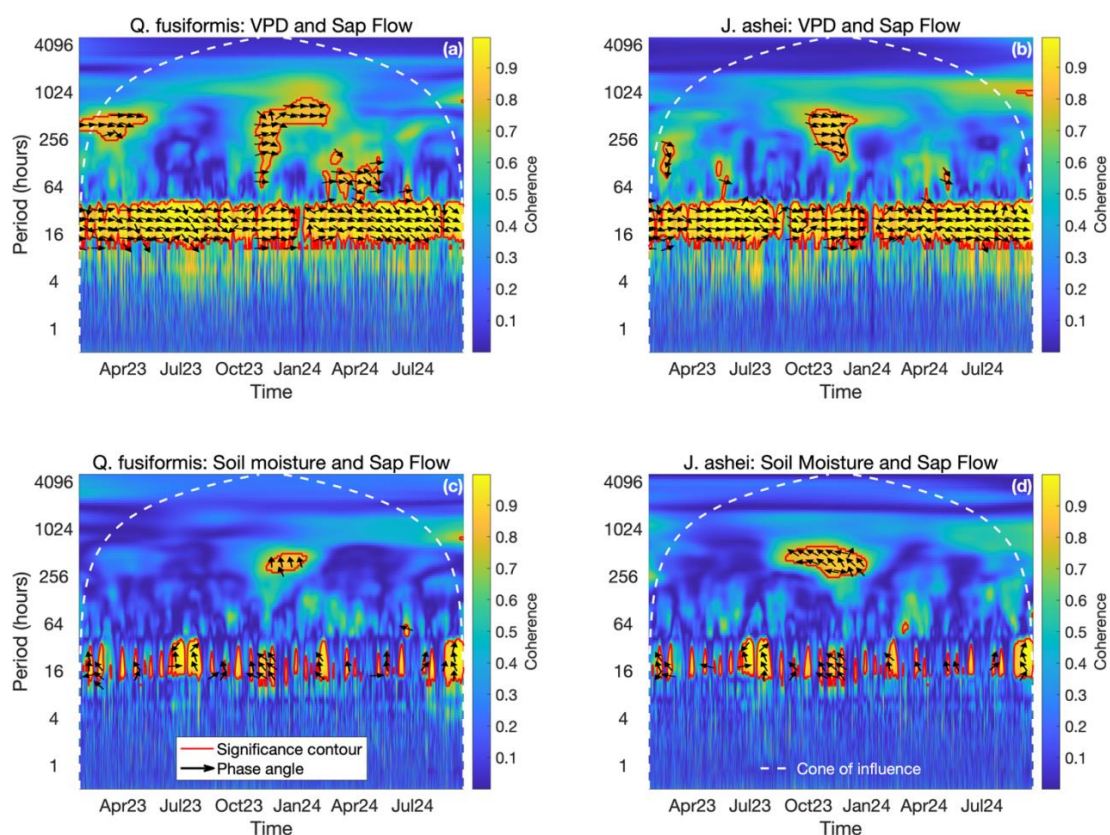
257 leading WWC (Figure 1.c and d; black arrows). Changes in WWC followed soil moisture closely during times of dry  
258 down (Figure 3.b and c). While *J. ashei* WWC had brief durations of coherence with soil moisture at the diel period  
259 leading up to the driest part of summer, as conditions reached their driest, the coherence shifted to longer periods  
260 (Figure 1.d), suggesting that the water content of *J. ashei* followed the long timescale trends of soil moisture as it  
261 dried. This agrees with the results in the time series plots – that *J. ashei* WWC dried with soil moisture but did not  
262 follow the soil moisture trend with the same fidelity as *Q. fusiformis* (Figure 3.b, c, and f).

### 263 3.2. Wavelet coherence of sap flow

264 In the wavelet coherence of VPD and sap flow of both species, significantly high coherence in the diel period existed  
265 for most of the study. The phase angles for both species were near zero at the diel period, indicating that VPD and sap  
266 flow were nearly in phase at the diel period. There were short breaks in the high coherence occurring at the end of  
267 summer 2023 and in January 2024. The break in coherence during summer 2023 was more pronounced for *J. ashei*  
268 (Figure 2.b) than for *Q. fusiformis* (Figure 2.a). For both species, this break in coherence occurred at the time when  
269 the KBDI reached its highest value of the season, indicating the drought was at its worst and soil moisture was very  
270 low (Figure 3.a). The breaks in coherence also coincided with precipitation events (Figure 3.d, e and f) and a decline  
271 in VPD (Figure 3.a), which likely indicated significant cloud cover during this period that would disrupt the diurnal  
272 pattern of VPD. The low coherence durations were longer and occurred more frequently for *J. ashei* than for *Q.*  
273 *fusiformis* (Figure 3.d and e).



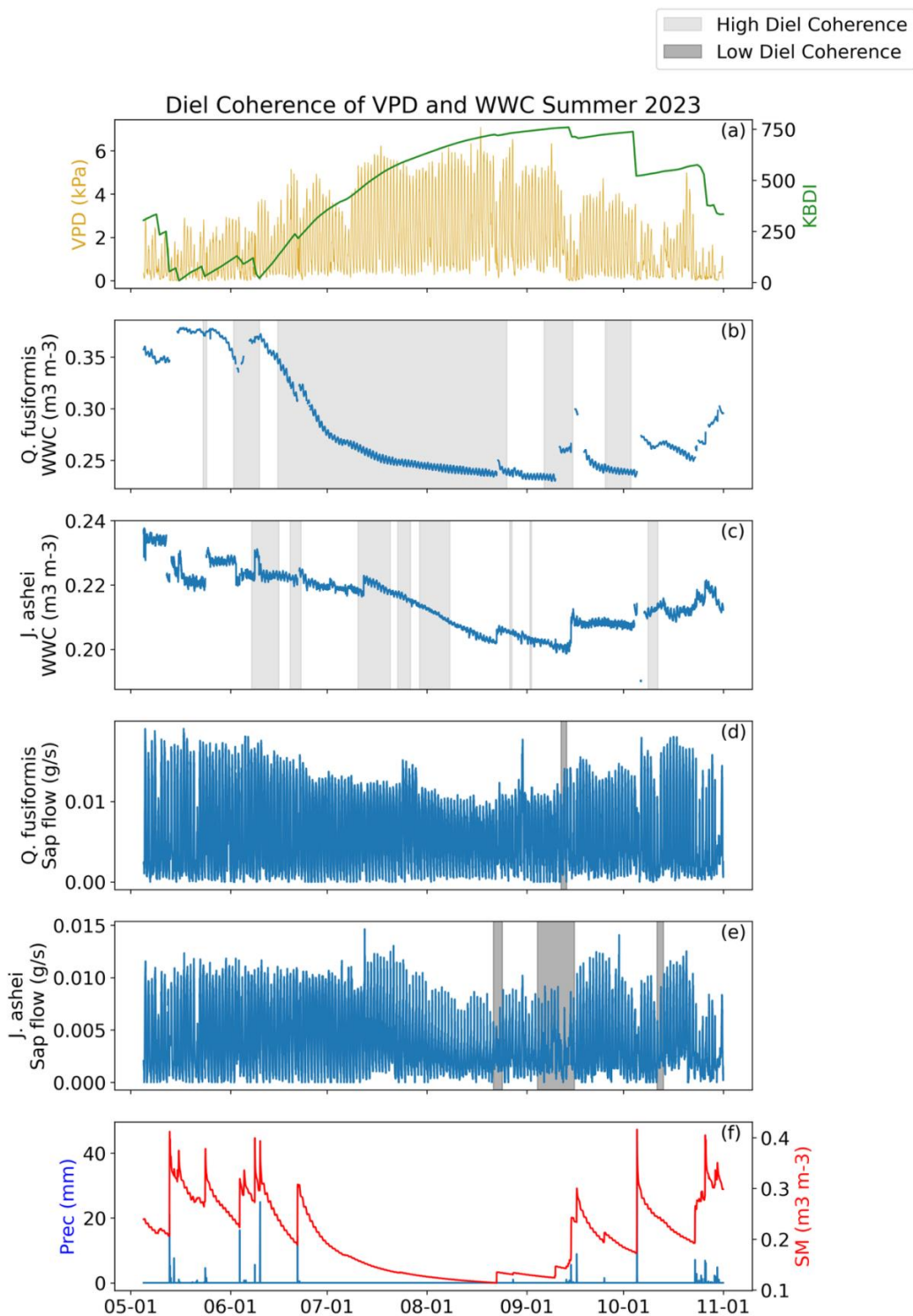
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275

276 **Figure 2:** Wavelet coherence plots of (a) VPD and *Q. fusiformis* sap flow, (b) VPD and *J. ashei* sap flow, (c) soil moisture  
 277 and *Q. fusiformis* sap flow, and (d) soil moisture and *J. ashei* sap flow. These plots demonstrate that both species have similar  
 278 responses to both environmental conditions with respect to sap flow. Durations of high coherence between VPD and sap  
 279 flow are centered around the diel period and are maintained for most of the study for both species, with some breaks in  
 280 high coherence. Phase angles are close to in-phase and mostly negative, indicating that VPD leads sap flow. For soil moisture  
 281 and sap flow, high coherence is centered around the diel period and is not consistent. When coherence is high, phase angles  
 282 are mostly positive, indicating that sap flow leads soil moisture.

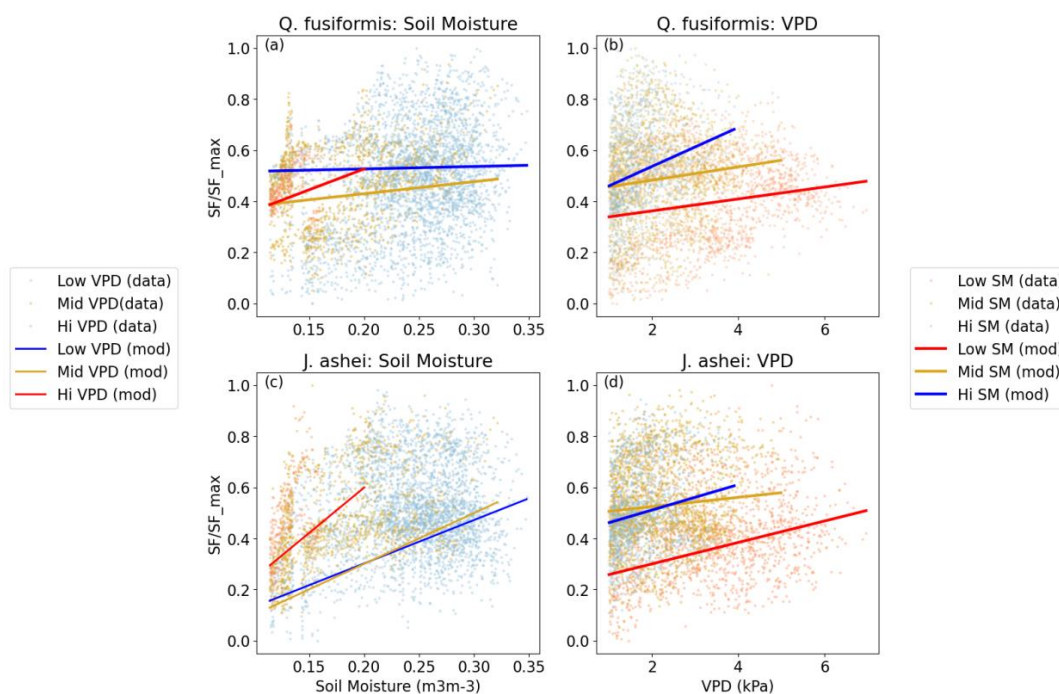
283 The durations of high coherence between soil moisture and sap flow centered at the diel period occurred at nearly the  
 284 same times for both species. The positive values of the phase angles, evidenced by the upward direction of the arrows  
 285 in Figure 2.(c and d), indicate that sap flow leads soil moisture. There was a strong diel coherence duration in the  
 286 summer of 2023, suggesting a correlation between daily transpiration and soil drying patterns when conditions were  
 287 hot and dry.





289 Figure 3: Meteorological conditions and sap flow time series from both species from May through October 2023. (a) VPD  
 290 (kPa) and Keetch-Byram Drought Index; (b and c) wood water content for *Q. fusiformis* and *J. ashei*, respectively – the  
 291 pink highlighted areas show times of high diel coherence between VPD and WWC for the respective species. *Q. fusiformis*  
 292 (b) maintains high diel coherence for longer than *J. ashei* (c). (d and e) Sap flow for the two species – the purple highlighted  
 293 regions represent times when diel coherence between VPD and sap flow becomes low. The drops in diel coherence are more  
 294 pronounced for *J. ashei* (e) than *Q. fusiformis* (d) and coincide with days with low VPD and/or precipitation;  
 295 (mm) and shallow soil moisture ( $m^3m^{-3}$ ).

296 **3.3. Observed linear regression of Sap flow**



297

298 Figure 4: Robust linear regression of relative sap flow with VPD and  $\Psi_{soil}$  as independent variables. (a) Relative sap flow  
 299 of *Q. fusiformis* as a function of soil moisture; (b) relative sap flow of *Q. fusiformis* as a function of VPD; (c) relative sap  
 300 flow of *J. ashei* as a function of soil moisture; (d) relative sap flow of *J. ashei* as a function of VPD. In all plots, dots represent  
 301 data points, and lines represent model predictions.

302 When soil moisture was low, both species maintained a positive relationship between sap flow and VPD. For both  
 303 species, relative sap flow was most sensitive to VPD when soil moisture was high, as seen by the steeper slopes for  
 304 high soil moisture compared to middle and low soil moisture (Figure 4.b and d), and by the high soil moisture  
 305 coefficients on  $\ln(VPD)$  of 0.2118 and 0.2350 ( $p < 0.001$ ) for *Q. fusiformis* and *J. ashei*, respectively, compared to  
 306 0.1032 and 0.1170 ( $p < 0.001$ ) for low soil moisture (Table 1 and 2). *J. ashei* was most sensitive to the interaction  
 307 variable,  $\ln(VPD) \times \Psi_{soil}$ , with coefficient 0.9938 ( $p = 0.004$ ). Maximum sap flow ( $SF_{max}$ ,  $g\ s^{-1}$ ) for the study period,  
 308 used to calculate relative sap flow, was 0.0196 and 0.0130 for *Q. fusiformis* and *J. ashei*, respectively. This is about  
 309 50% higher for *Q. fusiformis* compared to *J. ashei*. Multiplying by  $SF_{max}$  for both species, absolute sap flow for *Q.*  
 310 *fusiformis* was approximately 33% more sensitive to  $\ln(VPD)$  than *J. ashei* with new coefficients of 0.0020 and  
 311 0.0015, respectively, suggesting that *Q. fusiformis* may be capable of greater sap flow rates with a stronger absolute  
 312 response to VPD when soil moisture is low.

313 For both species, relative sap flow was most sensitive to  $\Psi_{soil}$  when VPD was high, with slopes on  $\Psi_{soil}$  of 0.0041  
 314 ( $p < 0.001$ ) and 0.0074 ( $p < 0.001$ ) at high VPD compared to 0.0013 ( $p < 0.001$ ) and 0.0024 ( $p < 0.001$ ) at low VPD for



315 *Q. fusiformis* and *J. ashei*, respectively (Table 1, Table 2). For all VPD levels, *J. ashei* relative sap flow was more  
 316 sensitive to soil water potential than *Q. fusiformis* (Figure 4.a and c). However, for all VPD levels, relative sap flow  
 317 was more sensitive to  $\ln(VPD)$  than soil water potential, suggesting that VPD has stronger control over sap flow.  
 318 Tables 1 and 2 provide a full description of linear model fit statistics.

319 **Table 1: The variables and their coefficients from the linear regressions for relative sap flow of *Q. fusiformis*. A total of six**  
 320 **models were made for six conditions: Low, Medium, and High VPD/SM**

<i>Q. fusiformis</i> relative sap flow						
Variables	Condition	Coefficient	p-value	Condition	Coefficient	p-value
$\ln(VPD)$	Low VPD	0.0750	<0.001	Low SM	0.1032	<0.001
	Mid VPD	0.1413	0.003	Mid SM	0.0596	0.001
	High VPD	-0.5307	<0.001	High SM	0.2118	<0.001
$\Psi_{soil}$	Low VPD	0.0013	<0.001	Low SM	-0.0007	<0.001
	Mid VPD	-0.0002	0.781	Mid SM	0.0278	0.013
	High VPD	0.0041	0.001	High SM	0.1558	0.474
$\ln(VPD) \times \Psi_{soil}$	Low VPD	-0.0011	0.003	Low SM	0.0005	0.001
	Mid VPD	0.0005	0.434	Mid SM	-0.0049	0.694
	High VPD	-0.0021	0.005	High SM	0.33745	0.307

321

322 **Table 2: The variables and their coefficients from the linear regressions for relative sap flow of *J. ashei*. A total of six models**  
 323 **were made for six conditions: Low, Medium, and High VPD/SM**

<i>J. ashei</i> relative sap flow						
Variables	Condition	Coefficient	p-value	Condition	Coefficient	p-value
$\ln(VPD)$	Low VPD	0.0696	<0.001	Low SM	0.1170	<0.001
	Mid VPD	0.2891	<0.001	Mid SM	0.0587	<0.001
	High VPD	-0.3868	0.069	High SM	0.2350	<0.001
$\Psi_{soil}$	Low VPD	0.0024	<0.001	Low SM	0.0014	<0.001
	Mid VPD	0.0004	0.604	Mid SM	-0.0337	0.001
	High VPD	0.0074	0.007	High SM	-0.6669	0.001
$\ln(VPD) \times \Psi_{soil}$	Low VPD	-0.0007	0.034	Low SM	-0.0002	0.162
	Mid VPD	0.0009	0.104	Mid SM	0.0134	0.251
	High VPD	-0.0036	0.024	High SM	0.9938	0.004

324

## 325 4. Discussion

326

327

### 4.1. The role of capacitance in buffering drought stress

328 Based on our results, *Q. fusiformis* was highly reliant on stem water stores when soil water became limiting, while *J.*  
 329 *ashei* did not demonstrate the same reliance on water stores, but depended on other mechanisms to maintain function  
 330 during drought. There is evidence of a continuum with trees on one end exhibiting vulnerable xylem and a reliance on  
 331 stored water, and trees on the other end having more resistant xylem (Meinzer et al., 2009). The two species in the  
 332 current study showcase the ends of this continuum. *Q. fusiformis* is classified as diffuse porous and has wider xylem  
 333 vessels compared to *J. ashei*, which has primarily tracheid-based conductive tissue, leading to the *Quercus* species  
 334 having greater water storage capacity (Figure 3.b and c; compare y-axes for WWC of both species). Another  
 335 consequence of wider xylem vessels is that *Q. fusiformis* is more vulnerable to embolism; stem P50, i.e., the xylem  
 336 tension at which 50% of conductance is lost, for *Q. fusiformis* is -1.8 MPa (Johnson et al., 2018a) and is -13.41 MPa  
 337 for *J. ashei* (McElrone et al., 2004). Based on wavelet coherence, *Q. fusiformis* water stores were more responsive to  
 338 the diel cycle of VPD than the water stores of *J. ashei* under low soil moisture (Figure 1.a and b). We hypothesized  
 339 that both species would rely on internal water stores during drought, and we observed a decline in wood water content  
 340 during drought for both species. However, *Q. fusiformis* was clearly utilizing stored water for its daily transpiration



341 needs far more than *J. ashei*, as evidenced by longer, sustained durations of diel coherence between WWC and VPD  
342 signals during times of low soil moisture (Figure 1.a). The high diel coherence between *Q. fusiformis* WWC and VPD,  
343 and the fact that the signals were in antiphase at the diel period, suggests that not only were the *Q. fusiformis* trees  
344 using their water stores throughout the day, but they were also able to refill them overnight enough to have a strong  
345 signal at the diel period.

346 Another source of water storage in *Q. fusiformis* could be in parenchyma cells that store non-structural carbon (NSC)  
347 and water. There is evidence that angiosperms with more vulnerable xylem vessels, that operate at smaller safety  
348 margins, contain more parenchyma as the starches and water stored in those cells would be necessary for embolism  
349 repair and refill, while there is no correlation between these measurements in gymnosperms (Kiorapostolou et al.,  
350 2019). *Q. fusiformis*, an angiosperm, has been recorded operating at a negative hydraulic safety margin, i.e., with  
351 water potentials below its P50 value (Johnson et al., 2018a). Theoretically, this would cause significant loss of  
352 conductance and embolism within the xylem network. A significant percentage of water released from storage to feed  
353 daily transpiration needs could come from embolized vessels (Tyree and Yang, 1990). A large proportion of water  
354 stored in the *Q. fusiformis* trees was released during the summer 2023 drought (Figure 3.b). If the capacitance comes  
355 from embolized vessels, this suggests that the trees are experiencing a fair amount of embolism as the drought  
356 continues. If reliance on capacitance during drought leads to embolism in *Q. fusiformis*, a greater amount of  
357 parenchyma could potentially aid in post-drought recovery.

358 Both species lost WWC during drought, following the same general drying trend as soil moisture, but *Q. fusiformis*  
359 WWC followed soil moisture with much more fidelity than *J. ashei* (Figure 3.b, c, and f). The shifts of high coherence  
360 between soil moisture and *J. ashei* WWC (Figure 1.d) to higher periods as the summer 2023 dry period persisted  
361 reflect the overall drying trend. *Q. fusiformis* WWC showed more dynamic behavior overall; this, together with the  
362 fidelity with which the *Quercus* water storage followed the soil moisture, was reflected in high coherence between  
363 soil moisture and *Q. fusiformis* WWC across sub-daily to weekly timescales (Figure 1.c).

364 It is possible that the *Q. fusiformis* WWC signal was more correlated to the shallow soil moisture than that of *J. ashei*  
365 (Figure 2.a and b) because its higher responsiveness to VPD at the diel period (Figure 1.a), which indicates greater  
366 use of capacitance, caused the *Q. fusiformis* WWC to dry out in a similar manner to the remote sensing soil moisture  
367 estimates (Figure 3.b and f). In contrast, *J. ashei* WWC was less responsive to VPD at the diel period (Figure 1.b),  
368 indicating that capacitance was not being used to the same extent as in *Q. fusiformis* and less connection between *J.*  
369 *ashei* WWC and the atmosphere. Thus, it could be that the differences of the wavelet coherence of soil moisture and  
370 WWC between the two species has more to do with capacitance use than rooting behavior. Currently, we do not have  
371 enough information about the subsurface at our study site, or the roots of the trees, to know the depths from which the  
372 study trees acquire water. We know that the soil is very shallow (~10 cm), and both species rely on water in the  
373 fractured bedrock, into which they are known to root several meters deep (Jackson et al., 1999). Thus, while the soil  
374 moisture estimates we used from the remote sensing product may be a good estimate of water in the shallow soil, it  
375 may not be representative of all the water available to the trees. A better understanding of the subsurface of our study  
376 site, the rooting behavior of the study trees, and how different roots contribute to transpiration and refill during drought  
377 (Grossiord et al., 2017; Nadezhdina and Cermak, 2000) would greatly improve our analysis.

#### 378 4.2. Vapor pressure deficit is the dominant driver of tree water relations in this ecosystem

379 The consistent diel relationship between VPD and sap flow signals (Figure 2.a and b) suggests that both study species  
380 continued transpiration regardless of changes in drought severity, unlike species which stop transpiration under  
381 drought conditions such as piñon pine in New Mexico (Grossiord et al., 2017; Sevanto et al., 2018). Even when KBDI  
382 surpassed 500, a threshold above which *J. ashei* is likely to have critically low foliar moisture (McCaw et al., 2018b),  
383 both species maintained high coherence between VPD and sap flow (Figure 2.a and b). This contradicted our  
384 hypothesis that drought would lead to a break in coherence between VPD and sap flow. In fact, times when sap flow  
385 and VPD signals were not coherent appear to coincide with precipitation and low VPD events and were likely not



386 related to drought severity. Results from the linear regression showed a positive relationship between VPD and sap  
387 flow at all soil moisture levels, although VPD's effect on sap flow may be reduced when soil moisture is low (Figure  
388 4; Table 1; Table 2). To maintain the relationship between diel cycles of VPD and sap flow, i.e., keep stomata open,  
389 when soil moisture is low the two species employed different strategies: *Q. fusiformis* showed a strong reliance on  
390 capacitance (Figure 1) as discussed in Section 4.1, while *J. ashei* likely reduces stomatal conductance, as found by  
391 Northup et al (2022), while not shutting its stomata completely. Wavelet coherence does not give magnitudes. Thus,  
392 it cannot show if one species is transpiring more than the other. The results from linear regressions suggest that *J.*  
393 *ashei* relative sap flow was more sensitive to VPD than *Q. fusiformis* relative sap flow under low soil moisture  
394 conditions (Table 1; Table 2), but further calculations (Section 3.3) showed that absolute sap flow of *Q. fusiformis* was  
395 more sensitive to VPD, and likely capable of higher rates, under low soil moisture. Both species are known to root  
396 several meters into fractured limestone bedrock (Jackson et al., 1999; McElrone et al., 2004). The observed  
397 relationship between VPD and sap flow is consistent with findings at other deep water-dependent ecosystems where  
398 evapotranspiration remains tightly coupled to atmospheric moisture as shallow soil moisture becomes depleted (Kibler  
399 et al., 2025).

#### 400 4.3. Future drought predictions and implications for the study species

401 Temperature and evaporative demand in Texas are forecasted to increase in the near future, which will exacerbate the  
402 severity of droughts even without any decreases in overall precipitation. By 2026, average temperatures in Texas are  
403 expected to be 1° C warmer than the average of the last 30 years. Days over 100° F (~38° C) are projected to be four  
404 times more common than in the 1980s (Escobedo et al., 2024). What does this mean for species, like *Q. fusiformis*  
405 and *J. ashei*, which maintain coupling with VPD even as shallow soils dry out? Both study species are highly drought-  
406 adapted with opposing strategies for dealing with drought. None of the trees in our study experienced mortality, but  
407 our results highlight key differences in their strategies, and we can speculate on how these strategies might play out  
408 during exceptional droughts. Overall, *J. ashei* experienced greater mortality rates than *Q. fusiformis* as a result of the  
409 2011 drought in Texas (Johnson et al., 2018a). Differences in mortality between the species were not consistent from  
410 site to site, and many factors were found to contribute that are not accounted for in this study (Crouchet et al., 2019).

411 *Q. fusiformis* may potentially be able to avoid drought stress and mortality by utilizing and refilling capacitance to  
412 maintain higher stomatal conductance for carbon uptake, especially at sites where it has access to deeper water.  
413 However, if the species were to experience prolonged drought characterized by increased temperatures and VPD, the  
414 aggressive water use strategy seen in the summer of 2023 could lead to significant, sustained loss of conductance. *Q.*  
415 *fusiformis* has been documented shedding leaves during drought (Northup et al., 2022), and we assume that stomata  
416 would eventually close, but the release of large proportions of water stores to the atmosphere early in a drought might  
417 cause irreparable damage if precipitation does not arrive on time. Crouchet et al. (2019) found that *Q. fusiformis*  
418 mortality from the 2011 drought was negatively correlated with stand density; if the relationship between VPD and  
419 the hydraulic systems of *Q. fusiformis* causes stress under drought conditions, the stress may be mitigated by the  
420 microclimate effects of denser stands, e.g., lower VPD and temperature.

421 *J. ashei* does not rely on capacitance in the same way as *Q. fusiformis* (Figure 1.a and b). *J. ashei* does release and  
422 refill water stores every day, and its water stores are lost over time when there is not adequate water available for  
423 refilling, but not to the same extent as its *Quercus* counterpart (Figure 1.a and b; Figure 3.b and c). Instead of  
424 capacitance, *J. ashei* relies on cavitation-resistant xylem (McElrone et al., 2004), stomatal regulation (Northup et al.,  
425 2022), and plasticity in leaf hydraulic traits (Johnson et al., 2018b) to maintain function during drought. Unlike *Q.*  
426 *fusiformis*, stand density was positively correlated with *J. ashei* mortality from the 2011 drought (Crouchet et al.,  
427 2019), potentially indicating that the effects of resource competition overcame any potential microclimate buffering.  
428 The different relationships between drought mortality and stand density between species could be due to how they  
429 access water, given that *Q. fusiformis* has been found to root deeper than *J. ashei* (Jackson et al., 1999).

#### 430 5. Conclusions



431 Our analysis revealed that vapor pressure deficit was the main driver of tree-water relations for both *Q. fusiformis* and  
432 *J. ashei* regardless of drought severity, *Q. fusiformis* was more reliant on capacitance than *J. ashei* when soil moisture  
433 was low, and that *Q. fusiformis* WWC dynamics were more closely related to shallow soil moisture dynamics than  
434 those of *J. ashei*. The results of this study also highlight the importance of representing hydraulic capacitance in plant  
435 hydrodynamic models.

436 Given that both species remain responsive to VPD when shallow soil moisture is low, future droughts characterized  
437 by high temperatures and high VPD threaten ecosystems dominated by these species. At sites where trees have access  
438 to deeper water sources through fractured bedrock effects may be mitigated. However, both species also rely on water  
439 input via precipitation to refill internal water stores. Therefore, deep water access may not be enough to survive  
440 droughts that coincide with increased temperatures and VPD. This applies not only to the ecosystem in this study, but  
441 to other ecosystems that rely on deep water and maintain a strong connection to the atmosphere during drought.

#### 442 **Data and code availability**

443 The Python and MatLab scripts used for analysis and visualization are available from the corresponding author upon  
444 request. Data used in the analysis is available here <https://doi.org/10.5281/zenodo.19377015>.

#### 445 **Author contributions**

446 CA and AM: conceptualization of the study. SC and AM: data curation. CA: formal analysis, visualization, and writing  
447 (original draft preparation). AM, SS, and AA: supervision. All authors: writing (reviewing and editing).

#### 448 **Competing interests**

449 The contact author has declared that none of the authors has any competing interests

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