



Differences in species-level growth responses to hydroclimate extremes in eastern US forests: Implications for a changing region

Journal:	<i>Global Change Biology</i>
Manuscript ID:	GCB-24-0776
Wiley - Manuscript type:	Research Article
Date Submitted by the Author:	15-Mar-2024
Complete List of Authors:	Maxwell, Justin; Indiana University Bloomington, Geography Au, Tsun Fung; University of Michigan-Ann Arbor, Institute for Global Change Biology; University of Michigan-Ann Arbor, Department of Ecology and Evolutionary Biology Kannenberg, Steven; West Virginia University, Department of Biology Harley, Grant; University of Idaho, Geography Dannenberg, Matthew; The University of Iowa, Geographical and Sustainability Sciences Ficklin, Darren; Indiana University Bloomington, Geography Robeson, Scott; Indiana University Bloomington, Geography Ferriz, Macarena; Indiana University Bloomington, Geography Lockwood, Ben; The Pennsylvania State University, Department of Ecosystem Science and Management Novick, Kimberly; Indiana University Bloomington, O'Neil School of Public and Environmental Affairs Phillips, Richard; Indiana University Bloomington, Biology Rochner, Maegen; University of Louisville, Department of Geographic and Environmental Sciences Pederson, Neil; Harvard University, Harvard Forest
Keywords:	Drought, Pluvial, Demographic Shift, Tree Rings, Climate Change
Abstract:	Forests around the world are experiencing changes due to climate variability and human land use. How these changes interact and influence the vulnerability of forests are not well understood. In the eastern United States (US), well-documented anthropogenic disturbances and land-use decisions have influenced forest species assemblages, leading to a demographic shift from forests dominated by xeric species to those dominated by mesic species. Contemporarily, the climate has changed and is expected to continue to warm and produce higher evaporative demand, imposing stronger drought stress on the forests. Here, we use an extensive network of tree-ring records from common hardwood species across ~100 sites and ~1300 trees in the eastern US to examine the magnitude of growth response to both wet and dry climate extremes. We find that growth responds asymmetrically to drought, with the positive growth response to pluvials failing to outweigh reductions of growth during drought. Mesic species such as <i>Liriodendron tulipifera</i> and <i>Acer saccharum</i> , which are becoming more

	<p>dominant, are more sensitive to drought than the xeric species such as oaks (<i>Quercus</i>) and hickory (<i>Carya</i>), especially the moderate and extreme drought intensities. Although more extreme droughts produce a larger annual growth reduction, mild droughts resulted in the largest cumulative growth decreases due to their more frequent occurrence. When using global climate model projections, all scenarios show drought frequency increasing substantially (3-9x more likely) by 2100. Thus, the ongoing demographic shift toward more mesic species in the eastern US combined with drier conditions results in larger drought-induced growth declines, suggesting that drought will have an even larger impact on aboveground carbon uptake in the future in the eastern US.</p>

SCHOLARONE™
Manuscripts

1 Title: Differences in species-level growth responses to hydroclimate extremes in eastern US

2 forests: Implications for a changing region

3 Running Title: Species growth responses to climate extremes

4

5 Justin T. Maxwell*¹, Tsun Fung Au^{2,3}, Steven A. Kannenberg⁴, Grant L. Harley⁵, Matthew P.

6 Dannenberg⁶, Darren L. Ficklin¹, Scott M. Robeson¹, Macarena Férriz¹, Benjamin R.

7 Lockwood⁷, Kimberly A. Novick⁸, Richard P. Phillips⁹, Maegen L. Rochner¹⁰, and Neil

8 Pederson¹¹

9 * Corresponding author information: maxweljt@indiana.edu, 812-855-5557

10 ¹ Department of Geography, Indiana University, Bloomington, IN, USA

11 ² Institute for Global Change Biology, University of Michigan, Ann Arbor, MI, USA

12 ³ Department of Ecology and Evolutionary Biology, University of Michigan, Ann Arbor, MI,
13 USA

14 ⁴ Department of Biology, West Virginia University, Morgantown, WV, USA

15 ⁵ Department of Earth and Spatial Sciences, University of Idaho, Moscow, ID, USA

16 ⁶ Department of Geographical and Sustainability Sciences, University of Iowa, Iowa City, IA,
17 USA

18 ⁷ Department of Ecosystem Science and Management, Pennsylvania State University, State
19 College, PA, USA

20 ⁸ O’Neil School of Public and Environmental Affairs, Indiana University, Bloomington, IN,
21 USA

22 ⁹ Department of Biology, Indiana University, Bloomington, IN, USA

23 ¹⁰ Department of Geographic and Environmental Sciences, University of Louisville, Louisville,
24 KY, USA

25 ¹¹ Harvard Forest, Harvard University, Petersham, MA, USA

26

27

28

29

30

31

32

33

34

35

36

37

38

39

40

41

42

43

44

45

For Review Only

46 Abstract

47 Forests around the world are experiencing changes due to climate variability and human land
48 use. How these changes interact and influence the vulnerability of forests are not well
49 understood. In the eastern United States (US), well-documented anthropogenic disturbances and
50 land-use decisions have influenced forest species assemblages, leading to a demographic shift
51 from forests dominated by xeric species to those dominated by mesic species. Contemporarily,
52 the climate has changed and is expected to continue to warm and produce higher evaporative
53 demand, imposing stronger drought stress on the forests. Here, we use an extensive network of
54 tree-ring records from common hardwood species across ~100 sites and ~1300 trees in the
55 eastern US to examine the magnitude of growth response to both wet and dry climate extremes.
56 We find that growth responds asymmetrically to drought, with the positive growth response to
57 pluvials failing to outweigh reductions of growth during drought. Mesic species such as
58 *Liriodendron tulipifera* and *Acer saccharum*, which are becoming more dominant, are more
59 sensitive to drought than the xeric species such as oaks (*Quercus*) and hickory (*Carya*),
60 especially the moderate and extreme drought intensities. Although more extreme droughts
61 produce a larger annual growth reduction, mild droughts resulted in the largest cumulative
62 growth decreases due to their more frequent occurrence. When using global climate model
63 projections, all scenarios show drought frequency increasing substantially (3-9x more likely) by
64 2100. Thus, the ongoing demographic shift toward more mesic species in the eastern US
65 combined with drier conditions results in larger drought-induced growth declines, suggesting that
66 drought will have an even larger impact on aboveground carbon uptake in the future in the
67 eastern US.

68 Keywords: Drought, Pluvial, demographic shift, tree rings, climate change

69 Introduction

70 Globally, forests are vulnerable to changes in climatic conditions (Breshears et al. 2005;
71 McDowell et al. 2008; Allen et al. 2010) and from human land-use decisions (Hamrick 2004;
72 McDowell et al. 2020). Forests in the eastern United States (US) are a classic example, with
73 well-documented, frequent, and extensive anthropogenic and natural disturbances over the past
74 *ca.* 200 years. Forest species composition across the eastern US have been strongly influenced by
75 human land management, where frequent fire and thinning before the 1850s by indigenous
76 groups prompted forests dominated by *Quercus* (oak) and *Carya* (hickory). European
77 colonization was followed by massive deforestation and later reforestation coupled with fire
78 suppression. Such practice resulted in a decline in the prevalence of *Quercus* and *Carya* species
79 and allowed the establishment of other less fire tolerant species such as *Acer spp.* (maple), *Fagus*
80 *grandifolia* (American beech), and *Liriodendron tulipifera* (tuliptree) (Fei et al. 2011), which are
81 now poised to replace the aging oaks and hickories (Novick et al. 2022).

82 The climate in the eastern US is also changing. Over the last several decades, an extended
83 wet period (or pluvial) has been prevalent across large parts of the eastern US (Karl et al., 1996;
84 Pederson et al., 2013; Ford 2014; Maxwell and Harley 2017). Further, the rate of temperature
85 increase in the eastern US compared to other regions of the world has been muted, in a large part,
86 due to reforestation (Barnes et al., 2024). Thus, the wetter and relatively cooler climate has
87 reinforced and accelerated ongoing demographic shifts in eastern US forests, resulting in less
88 drought and fire (McEwan et al., 2011). However, the rate of reforestation is likely to decrease,
89 diminishing the temperature buffering effect (Barnes et al., 2024). Combined with the continued
90 warming of global temperatures from anthropogenic emissions, hotter and drier conditions in the
91 eastern US are likely to accelerate (Ficklin and Novick, 2017; Wehner et al., 2017).

92 The likelihood of a shift from an exceptionally wet climate to a more arid climate in
93 eastern North America requires an understanding of the response of species and forest stands to
94 climatic extremes (Costanza et al., 2023). The impact of climatic extremes on tree growth
95 depends on the timing, context, duration and intensity of water stress (Schwarz et al., 2020), with
96 trees experiencing greater growth reductions when drought occurs during the growing season
97 (Delpierre et al., 2016; D'Orangeville et al., 2018; Hoffmann et al., 2018). Additionally, forest
98 composition may play a role in drought response since tree species diversity may buffer the
99 sensitivity of forests to climate extremes (Grossiord, 2020; Isbell et al., 2015; Anderegg et al.,
100 2018). Less is known about forest responses to pluvial conditions (but see Jiang et al., 2019;
101 Lockwood et al., 2023). Thus, characterizing the response of different species assemblages to
102 climatic extremes is crucial for understanding forest dynamics and productivity under projected
103 future scenarios. This understanding is especially important in the eastern US, where forests have
104 historically sequestered ca. 40% of regional carbon emissions (Pan et al. 2011). The future fate
105 of this sink is uncertain and hinges on tree- and stand-level responses to climate extremes.

106 While ecophysiological responses to droughts and pluvials, such as changes in gas
107 exchange, are useful for examining species-specific responses, the short length of such records
108 limits the number of extremes to examine species-specific responses. Using tree rings from
109 mature, canopy-dominant trees extends the number of extreme wet and dry events that we can
110 examine to see how growth responds to extreme climatic conditions. Further, growth is often
111 decoupled from photosynthesis (Dow et al. 2022; Cabon et al. 2022), and this is exacerbated
112 during drought (Kannenberg et al. 2022). Thus, examining growth directly can give insight into
113 how droughts will impact long-term carbon storage in woody biomass and how ongoing
114 demographic shifts could impact the response of eastern US forests to future climate change.

115 Here, we examine the magnitude of growth responses of several common eastern US
116 hardwood tree species to both wet and dry climate extremes. We hypothesize that ongoing
117 demographic shifts are producing forests that are more susceptible to deleterious drought impacts
118 on growth, a change with negative impacts to carbon sequestration. To test the hypothesis, we
119 leverage a broad and diverse network of ~100 tree-ring chronologies (~1300 trees), focusing on
120 five of the most widespread species throughout the eastern US. We further examine how this
121 demographic shift will impact future responses of eastern US forests by using climate model
122 projections for a number of greenhouse-gas scenarios.

123

124 Methods

125 Study region/samples

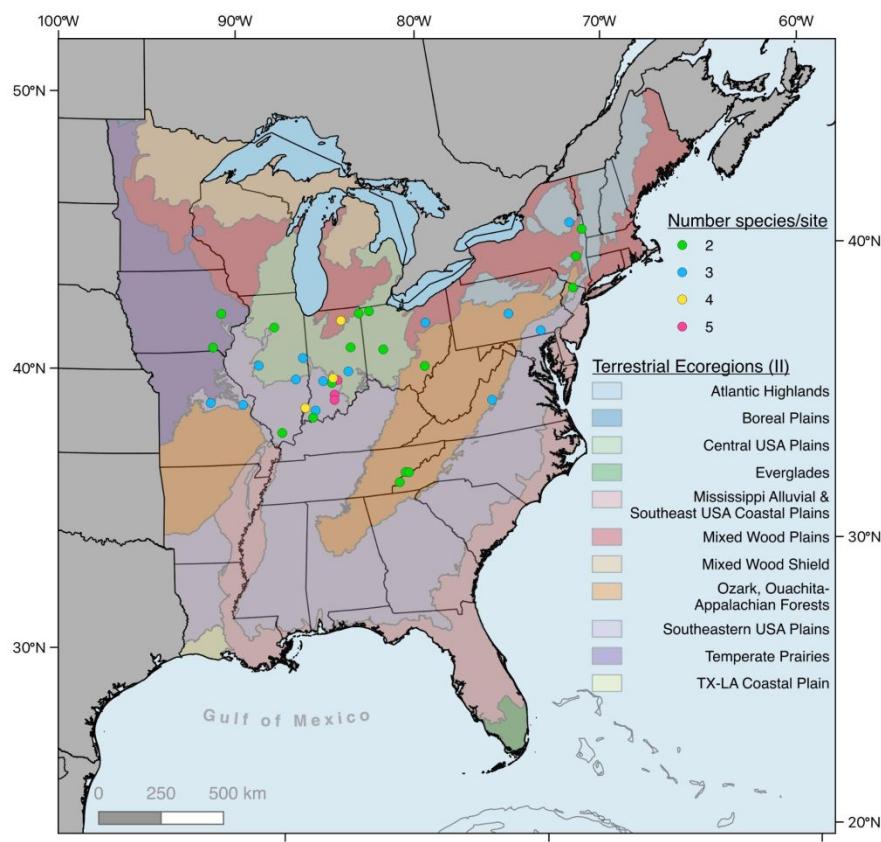
126 The study region encompasses a large portion of the Eastern Deciduous Forest biome in
127 North America (Figure 1). The climate over this area is humid continental with the southern
128 portion of the study region having a humid sub-tropical climate. We targeted canopy-dominant
129 trees from the most common species in this region, species that also employ a range of water-use
130 strategies from aggressive (*i.e.* anisohydric; *Q. rubra* and *Q. alba*), intermediate (*C. ovata* and *A.*
131 *saccharum*), and conservative (*i.e.* isohydric; *L. tulipifera*) (Roman et al. 2015; Yi et al., 2019;
132 Novick et al. 2022).

133

134 Tree-ring processing

135 We collected tree-ring chronologies (published in: Pederson, 2005; Maxwell et al., 2020)
136 across 36 forest stands for a total of 99 chronologies. All 36 sites had at least two co-occurring
137 species, while 19 sites had three, and six sites had four or more. While each of the five species

138 were not present at every site, sampling those that are growing in the same landscape position at
139 each site decreases the confounding influence of site conditions and allows for a clearer
140 comparison of species-specific growth response to extreme climate (Au et al. 2020, 2022a;
141 Lockwood et al. 2023). Overall, we gathered a total of 18 chronologies for *Q. rubra*, 25 for *Q.*
142 *alba*, 19 for *C. ovata*, 15 for *A. saccharum*, and 22 for *L. tulipifera* across the eastern deciduous
143 forest (Figure 1).



144
145 Figure 1: Tree-ring study sites across the Eastern Deciduous Forest biome. Map of tree-ring sites
146 showing the number of species-specific chronologies per study site along with the level two
147 terrestrial ecoregions as defined by Commission for Environmental Cooperation
148 (<http://www.cec.org/north-american-environmental-atlas/>).

149
150 For each site, we created species-specific chronologies that were generated from 5 to 50
151 trees, giving us a total of 1,299 trees used in this study. We followed standard collection methods
152 for dendrochronological studies. We targeted canopy-dominant trees and extracted two core

153 samples per tree. Once the samples were prepared using progressively finer sandpaper, we
154 visually crossdated the ring widths and used the program COFECHA (Holmes 1983) to
155 statistically ensure accurate dating. We reduced growth related to non-climatic influences by
156 standardizing each measurement series with a two-thirds spline (Cook and Peters 1981) and
157 adjusting for endpoints (Bussberg et al. 2020) using the program ARSTAN (Cook 1985).

158 Sampling from canopy-dominant trees whilst not including information from the
159 subcanopy can introduce a bias when examining species-specific responses to climate (Dye et al.
160 2016). Traditionally, canopy-dominant species were thought to be more sensitive to climate
161 (Alexander et al. 2018) compared to subcanopy trees, but this may not be the case in more mesic
162 forests, such as in the eastern US, where the sub-canopy trees could have a larger decrease in
163 growth during hot periods (Rollinson et al. 2021). However, canopy-dominant trees sequester
164 carbon for longer periods, account for more biomass and carbon reserve (Bennett et al. 2015),
165 and have lived long enough to experience multiple droughts and pluvials of differing intensities.
166 Thus, they are arguably a fundamental portion of the forest canopy for understanding drought
167 impacts on growth and their consequences for carbon sequestration.

168

169 Climate Data

170 We gathered monthly standardized precipitation-evapotranspiration index (SPEI;
171 Beguería et al. 2014) data for the nearest (0.5°) grid point to each forest stand. The SPEI is
172 standardized based on a probabilistic mapping of the precipitation (water supply) minus potential
173 evapotranspiration (water demand) distribution onto a standard normal distribution, producing an
174 index where 0 represents median conditions for a given location over the time period used for the
175 fitting. This approach allows us to compare the influence of extreme dryness and wetness on tree

176 growth across multiple sites that experience different climatic regimes. Because both water
177 supply and water demand can influence tree growth, incorporating an index that includes both
178 metrics is important (Williams et al. 2013, Novick et al. 2024). The SPEI is multiscalar, so it also
179 allows us to evaluate how the duration of abnormal to extreme wet and dry conditions impact
180 species-specific growth. We gathered SPEI for three temporal scales (1, 3, and 6 months) to
181 represent short-term to growing season length anomalies in soil moisture. To capture climatic
182 conditions that could influence the entire growing season, we examined conditions from March
183 through August, using SPEI6 for August. To represent the peak of the growing season, we
184 examined SPEI3 in July, which represents conditions from May through July. Lastly, for the
185 short timescale (SPEI1), we gathered data for each month from May–August, which are the most
186 important months for growth in eastern North America (D'Organville et al. 2018). We present
187 the results from SPEI6 in the main text and provide the shorter timescale results, which all were
188 similar to those from SPEI6, in the supplemental materials.

189

190 Drought and Pluvial effects

191 To determine how mild to extreme departures in water supply and demand impact
192 species-specific growth, we calculated drought and pluvial effects for each species. We defined
193 three, mutually exclusive drought thresholds, from mild ($\text{SPEI} \pm 1.0$, which probabilistically
194 equates to one standard deviation from the mean) to moderate ($\text{SPEI} \pm 1.5$) and extreme (SPEI
195 ± 2.0). For each threshold, we calculated the percentage growth change during a drought or
196 pluvial (i.e., drought and pluvial effects) by averaging the standardized ring width (SRW) for all
197 years that were classified as a drought (SRWd) or pluvial (SRWp). We then calculated the
198 averaged ring width for non-drought and non-pluvial years (SRWn). Drought effects were

199 calculated as the difference between the averaged SRW during drought from the normal years
200 then divided by averaged growth during normal years and multiplied by 100 to get a percentage
201 change (Kannenberg et al. 2019; Au et al., 2020). Pluvial effects were calculated in the same
202 manner except with the difference between the averaged growth during pluvial years and normal
203 years.

$$204 \quad Drought\ Effect = \frac{SRWd - SRWn}{SRWn} \times 100$$

205 where SRWd and SRWn are the average standardized growth for all drought years and normal
206 years respectively.

$$208 \quad Pluvial\ Effect = \frac{SRWp - SRWn}{SRWn} \times 100$$

209 where SRWp and SRWn are the average standardized growth for all pluvial years and normal
210 years respectively.

212

213 We calculated drought and pluvial effects for each month from May–August for SPEI1,
214 May–July average for SPEI3, and March–August average for SPEI6. To determine if the mean
215 effects were significantly different across the species, we used one-way ANOVA with a Tukey
216 HSD post hoc test.

217 To examine the lasting effects of extreme wet and dry conditions on SRW, we calculated
218 the growth differential percentage (Lloret et al., 2011) where we formed the ratio of the mean
219 SRW during the two years following the extreme year (Kannenberg et al., 2019) subtracted from
220 the growth during the extreme year in the numerator and the mean of SRW during the five years
221 preceding the extreme year (Au et al. 2022b) in the denominator:

222

$$223 \quad Growth\ Differential = \frac{SRW_{post1-2} - SRW_e}{SRW_{pre1-5}} \times 100$$

224

225 where e represents the year of an extreme event.

226 We excluded extreme years that occurred within five years following or two years prior
227 to another event. We chose to examine only the two years following an extreme event since
228 previous research has found that legacies longer than two years are relatively rare in Eastern
229 Deciduous Forests (Kannenberg et al. 2019). When examining the growth differential percentage
230 from pluvials, the resulting index typically will be negative, while recovering from droughts will
231 result in a positive growth differential percentage. In addition to drought and pluvial effect size,
232 we compared the overall response to climate for each species (i.e., between the SRW and SPEI
233 during the period of overlap, 1901 to 2015) using both a linear and a quadratic regression model.
234 We then used both the Akaike Information Criterion (AIC) and adjusted R^2 to evaluate model
235 performance and fit, finding AIC to be lower and adjusted R^2 to be higher for the quadratic
236 model for all species (Supplemental Table 1).

237

238 Future projections

239 To examine how future climate change could impact species-specific responses to
240 extreme wet and dry conditions, we extracted climate projections from 12 global climate models
241 (GCMs) from the Coupled Model Intercomparison Project – Phase 6 (CMIP6; Eyring et al.,
242 2016) listed in Supplemental Table 2 and for four Shared Socio-economic Pathways (SSP) that
243 include low (SSP1-2.6), moderate-low (2-4.5), moderate-high (3-7.0), and high (5-8.5) emission
244 pathways (O'Neill et al., 2016). For each GCM, we extracted average monthly air temperature,
245 precipitation, and the variables needed to estimate reference evapotranspiration (Allen et al.,
246 2006), which include maximum air temperature, minimum air temperature, wind speed, relative
247 humidity (used to estimate vapor pressure deficit), and incoming solar radiation (used to estimate

248 net solar radiation). After bi-linearly interpolating each GCM to a common 1.5° grid, climate
249 data for the nearest GCM grid node to each forest stand was extracted.

250 We used the ensemble mean of the monthly mean precipitation along with reference
251 evapotranspiration (Allen et al., 2006) averaged across the growing season months (March-
252 August), to calculate SPEI using the “SPEI” package in R (Beguería et al., 2014;
253 Vicente-Serrano et al., 2010). The parameters of the log-logistic distribution used to estimate
254 SPEI were calculated using the instrumental period (1901-2015). Because the ensemble mean of
255 the future climate conditions has muted interannual variance compared to the instrumentally
256 recorded data, we bias-corrected the projected SPEI after fitting the SPEI distribution using the
257 climate model outputs. The time series of SPEI had both positive and negative biases across its
258 cumulative probability distribution, thus we used quantile mapping to bias correct the projected
259 SPEI for each site (Ficklin et al. 2016; Robeson et al. 2020) with the “RQUANT” method in the
260 “QMAP” package in R (Gudmundsson et al. 2012).

261

262 Scaling for forest response

263 To examine how our results using 36 forest stands may scale to larger spatial scales, we
264 used a community-weighted mean approach based on composition estimates of the canopy from
265 the Forest Inventory and Analysis (FIA) for the eastern US. Because our study sites cover a large
266 area and species composition varies dramatically over space, we calculated multiple forest
267 response scenarios based on species compositions found throughout the eastern US including: 1)
268 a xeric species (*Quercus* and *Carya*) dominated forest with *Quercus* making up 40%, *Carya*
269 40%, *Acer* 10%, and *Liriodendron* 10%; 2) a mesic species dominated forest where *Acer* makes
270 up 40%, *Liriodendron* 40%, *Quercus* 10%, and *Carya* 10%; and 3) a mixed forest, where

271 *Quercus*, *Carya*, *Liriodendron*, and *Acer* each make up 25% of the forest. In the FIA data, we did
272 not see an *Acer* or *Liriodendron* dominated forest and thus did not create one for this analysis.
273 Similarly, there were other assemblages like “beech-maple” (*Fagus-Acer*) or “maple-hemlock”
274 (*Acer-Tsuga*) that were common in the FIA data, but we did not have the species well-replicated
275 in our co-occurring tree-ring network and thus, we did not examine these species combinations.

276 Lastly, to evaluate the cumulative effect of droughts and pluvials for each forest type, we
277 multiplied the magnitude of the drought and pluvial effects using 75th, 50th, and 25th percentiles
278 (shown in Figure 2) of the range of effects sizes across the sites for each species with the number
279 of occurrences over the observed period (1901-2015). We then scaled those cumulative effects
280 across the different forest types. For future conditions, we conducted the same calculation using
281 future climate projections for each climate scenario, where we used the same effect sizes but then
282 multiplied by the number of occurrences of extreme events for each threshold for the future
283 period (2016-2100). To make the comparison over the same number of years across the observed
284 and the future projections, we shorten the observed record to 1932-2016.

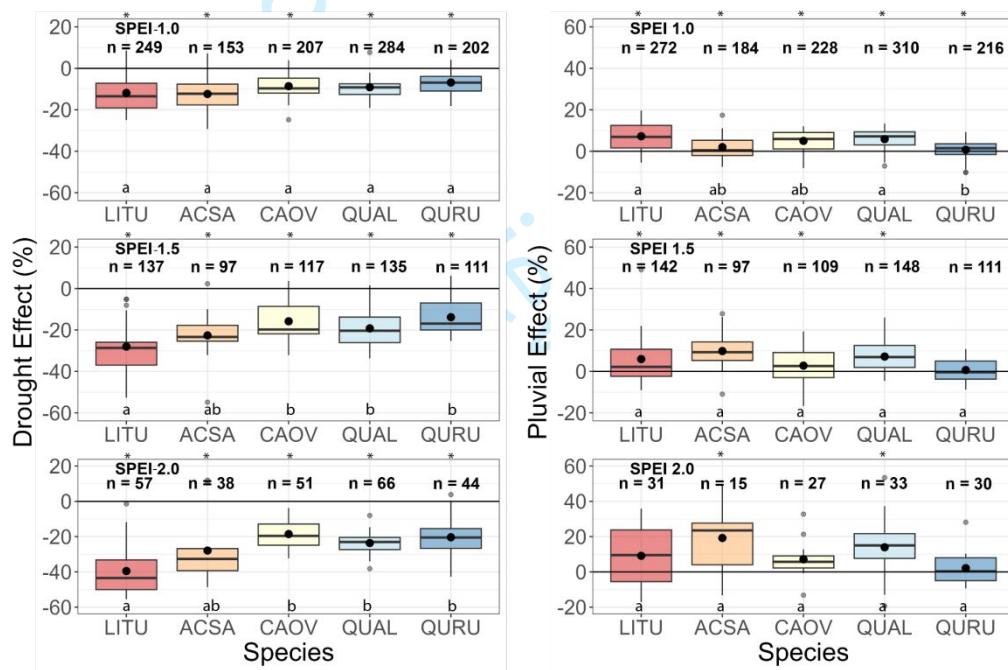
285

286 Results

287 Drought and Pluvial Effects

288 Of the five species examined, the growth of *L. tulipifera* was most sensitive to drought,
289 with a median drought effect ranging from a 19% decrease in growth during mild droughts to a
290 53% decrease during extreme droughts. During mild droughts (using the seasonal August SPEI6
291 = -1.0), all species had a similar growth decrease (Figure 2). As the drought intensity threshold
292 increased, drought response differences across species expanded in effect size and *L. tulipifera*
293 was significantly ($p < 0.05$) different from *C. ovata*, *Q. alba*, and *Q. rubra* (Figure 2). The

294 average drought effect of *L. tulipifera* and *A. saccharum* was two times greater than for the
 295 *Quercus* and *Carya* species during extreme droughts, while the difference is only 1.25 times
 296 greater during mild droughts. The same pattern existed when examining shorter drought
 297 durations (i.e., SPEI1 and SPEI3), where all species experienced similar growth decreases during
 298 mild droughts but as drought increased, *L. tulipifera* consistently experienced significantly
 299 greater decreases in growth during drought while *A. saccharum* had an intermediate response,
 300 and *C. ovata* and the *Quercus* species had smaller growth decreases (Supplemental Figure 1 and
 301 2).



302
 303 Figure 2: Species-level responses to hydroclimate extremes in eastern US forests. Effects of
 304 species growth to drought (left) and pluvial (right) conditions for seasonal (March-August
 305 average; August SPEI6) hydroclimate conditions for mild (SPEI6=±1.0; top), moderate
 306 (SPEI6=±1.5; middle), and extreme (SPEI6 = ±2.0; bottom) events. Lower-case lettering
 307 represents statistical significance differences in effect size between species via an ANOVA
 308 analysis Tukey HSD post hoc test. Asterisks represent the mean is significantly ($p \leq 0.05$)
 309 different from zero using a one-sample t-test. The sample size of the number of extremes
 310 experienced by each species is denoted. LITU=*L. tulipifera*; ACSA=*A. saccharum*; CAOV=*C.
 311 ovata*; QUAL=*Q. alba*; QURU=*Q. rubra*.
 312

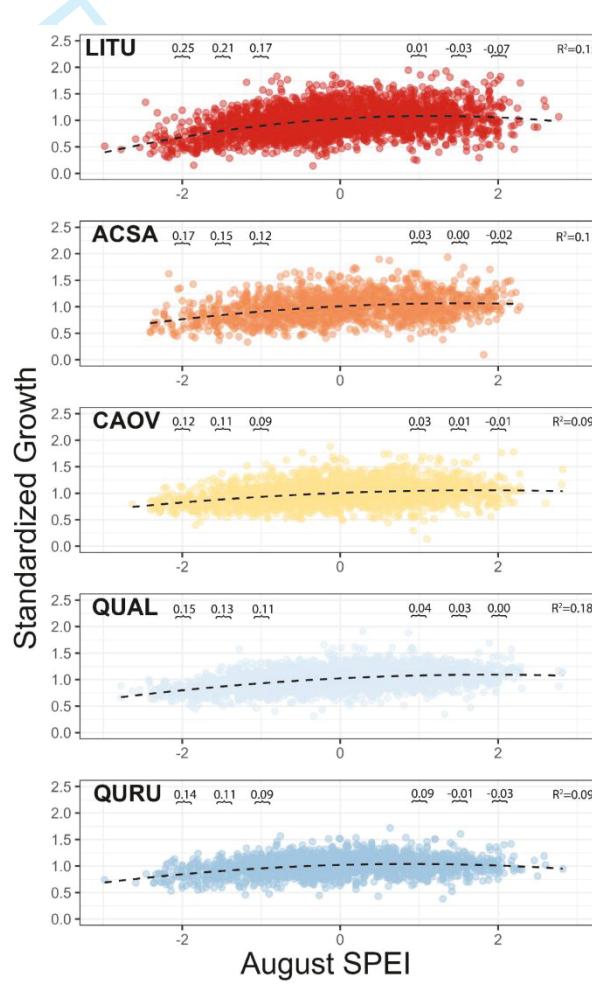
313 The pluvial effects were relatively smaller in magnitude (all less than 20% increases in
314 growth) than drought responses, and there were few differences in pluvial response among
315 species (Figure 2). As pluvial intensity increased, growth generally remained the same or
316 marginally increased (Figure 2). Further, the lack of species-specific response remained as
317 pluvial intensity increased, although the variance of the response increased to the point that only
318 *A. saccharum* and *Q. alba* had responses significantly different than zero (Figure 2). This same
319 pattern occurred during shorter pluvial durations (SPEI1 and SPEI3), with increased variance in
320 the pluvial response with increased intensity and a lack of a species-specific response
321 (Supplemental Figures 3 and 4). In all cases, the pluvial response was smaller than the drought
322 response.

323 Sensitivity to climatic extremes

324 When examining the slope of the non-linear relationship between growth and the full
325 spectrum of growing season SPEI (not just the extremes), we find similar species-specific
326 responses (Figure 3, Supplemental Table 1). *Liriodendron tulipifera* has the steepest slope
327 followed by *A. saccharum*, *Q. alba*, *C. ovata*, and *Q. rubra* during drought conditions (Figure 3).
328 Slopes during pluvial conditions were marginally positive or flat during mild wet conditions and
329 shifted to more negative slopes when pluvial conditions increased in intensity for all species
330 (Figure 3).

331 The growth differential percentage indicates that *L. tulipifera* growth post drought is the
332 closest to the pre-drought level (Figure 4) across the two lower drought intensity thresholds,
333 however *A. saccharum* has a higher growth after the most extreme droughts. Interestingly, we
334 see species-specific patterns in growth after drought throughout the various drought intensities
335 (Figure 4). However, for more intense droughts, species-specific responses become more

336 pronounced with the general pattern that growth from *L. tulipifera* and *A. saccharum* is closer to
 337 pre-drought conditions, while *C. ovata*, *Q. alba*, and *Q. rubra* have less growth compared to pre-
 338 drought conditions (Figure 4). When examining how growth returns to normal from wet events
 339 (Supplemental Figure 5), the growth differential percentage has generally lower negative
 340 percentages (i.e., above the pre-pluvial growth conditions) with few differences among species.
 341 As conditions get wetter, we see a weaker growth differential percentage with only *Q. alba*
 342 having a growth differential percentage different from zero during the most extreme pluvials.

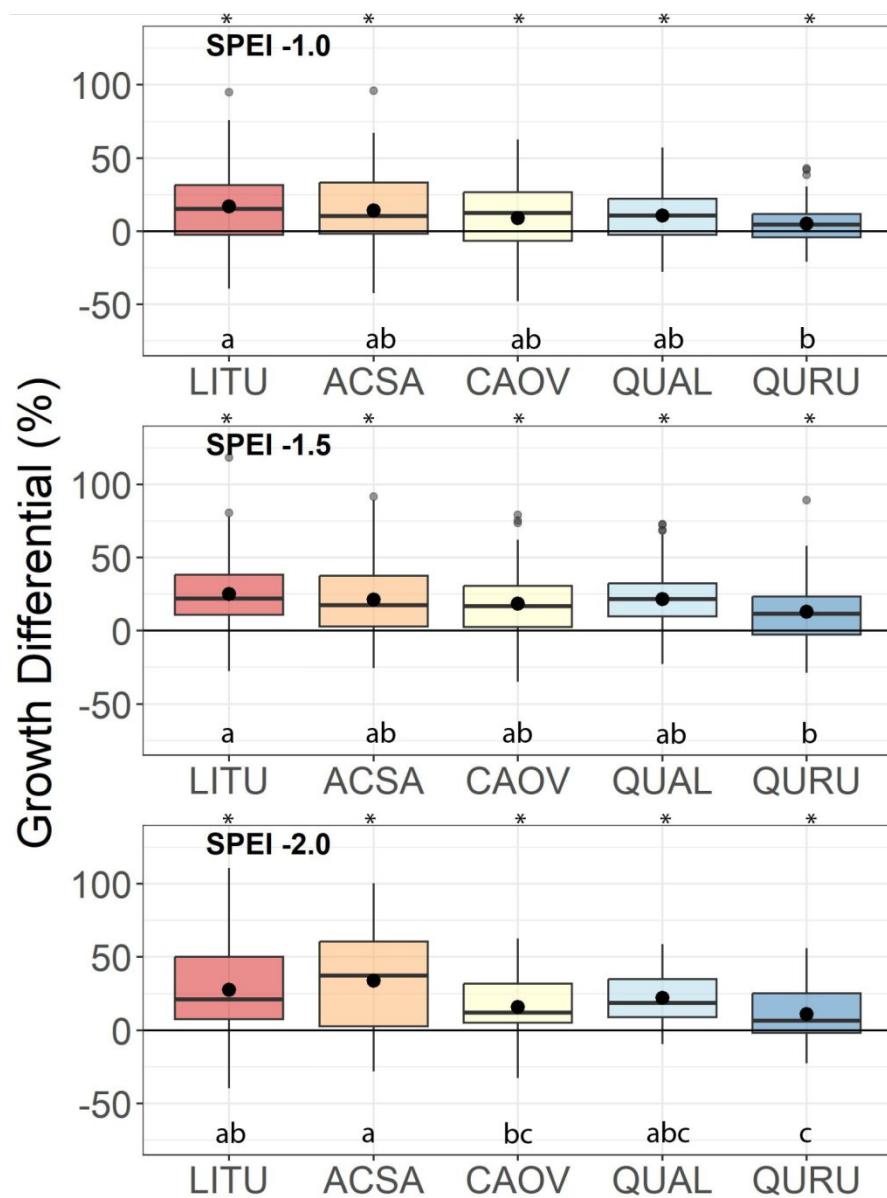


343

344 Figure 3: The relationship between standardized growth and March-August SPEI (August
 345 SPEI6) values for each species. Dashed line is the ordinary least-squares quadratic regression
 346 whose slope (SRW/ SPEI) at SPEI values of -2.0, -1.5, -1.0, 1.0, 1.5, and 2.0 along with R²
 347 are given in each panel.

348

349



350

351 Figure 4: Growth differential percentage from drought index for each species. The growth
 352 differential percentage is averaged from the two years after drought and accounts for the
 353 weighting of drought impacts on growth. A higher mean growth differential percentage indicates
 354 growth closer to the pre-drought conditions. Lettering represents statistical significance
 355 differences in effect size between species via an ANOVA analysis Tukey HSD post hoc test.
 356 Asterisks represent the mean is significantly higher than zero using a one-sample t-test.

357

358

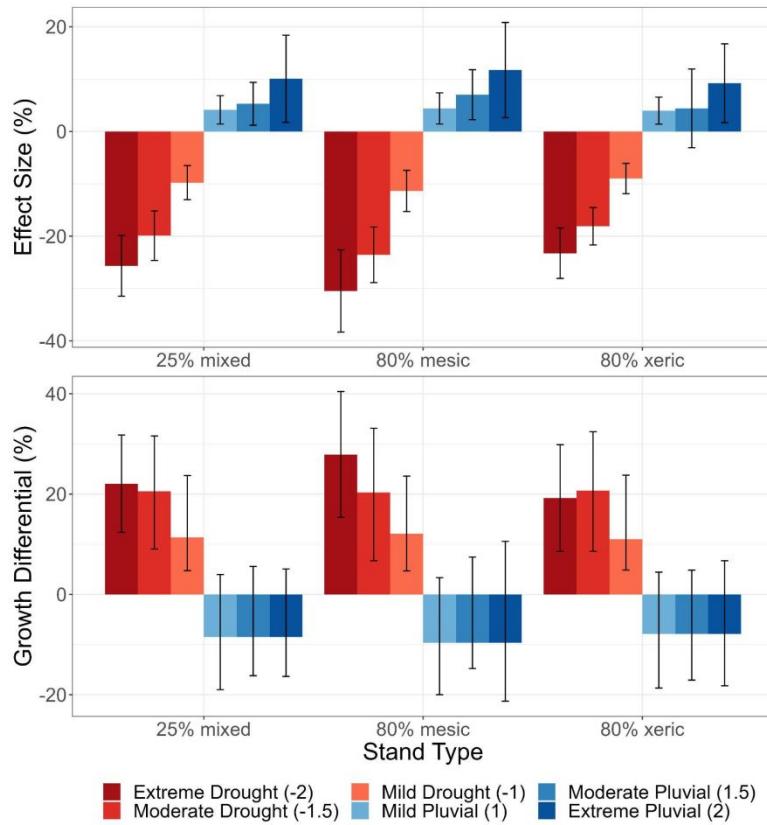
359

360

361

362 Forest Response

363 When scaled to represent various species compositions of canopy-dominant trees
364 throughout the eastern US, we found that the xeric species makeup (80% *Quercus-Carya*) had
365 the smallest mean growth reduction (14.5%) to drought compared to the other forest
366 compositions (15.9-18.7%) across drought intensities for the growing season (August SPEI6)
367 (Figure 5). The mesic species makeup (80% *Acer-Liriodendron*) had the highest mean growth
368 reduction at 18.7%. Thus, more intense droughts have a larger (up to 8% total) impact on growth
369 for mesic dominant forests compared to xeric (Figure 5). For pluvials, we see that xeric forests
370 have the smallest increase in growth and mesic forests have the largest increases. In all cases, the
371 increases in growth from pluvials are smaller than the decreases during drought (Figure 5). In
372 terms of growth differential percentage, forests composed of mesic species have higher growth
373 differential percentages than xeric forests, particularly for extreme droughts (Figure 5). The
374 return to normal growth from pluvial conditions is modest with large errors, independent of the
375 intensity of the pluvial event (Figure 5).



376

377 Figure 5: Effect size and growth differential percentage for Eastern Deciduous Forest species
 378 under hydroclimate extremes. Top) Pluvial and drought effect sizes for scaled canopy dominant
 379 species composition scenarios across intensity thresholds, with error bars. Bottom) Growth
 380 differential percentage percentages for scaled canopy dominant species composition scenarios
 381 across intensity thresholds, with error bars. Xeric = 80% of canopy dominant trees are *Quercus*
 382 and *Carya*; Mesic = 80% of trees are *Acer* and *Liriodendron*; Mixed = *Quercus*, *Carya*,
 383 *Liriodendron*, and *Acer* each make up 25% of the forest.

384

385 Future Climate

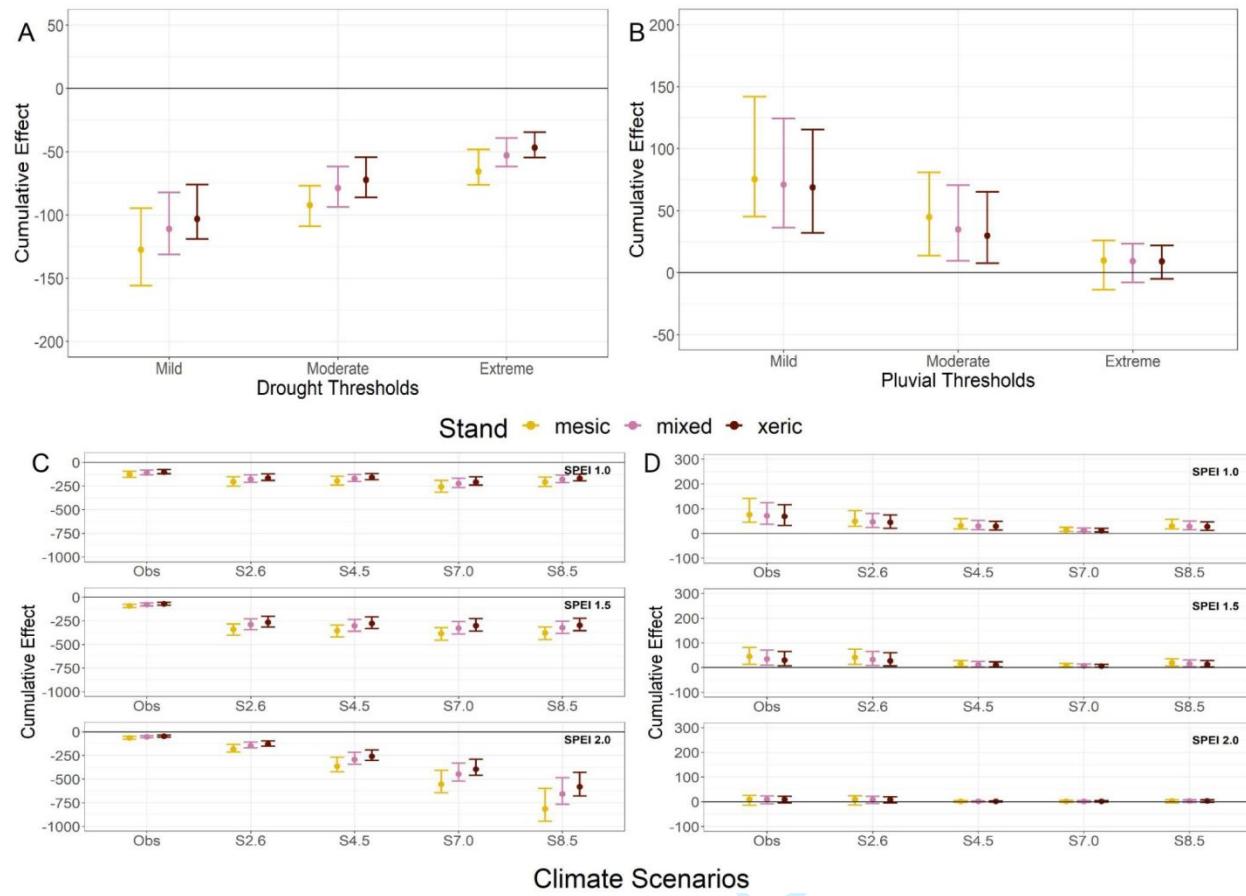
386 The ensemble of the climate models projects a drier climate in the future (2016-2100) for
 387 all scenarios across the 37 sites, ranging from a mean decrease of -0.38 (SSPI -2.6) to -1.00
 388 (SSP5-8.5) in August SPEI6 (Supplemental Figure 6). In addition to a shift in mean conditions,
 389 the distribution of SPEI values changes depending on concentration scenario and drought or
 390 pluvial category (Supplemental Figure 6). On the wet tail of the distribution, we see very little
 391 change across scenarios and only marginal changes from the observed period, although the

392 mildly wet SPEI values become much less frequent (2-5 times less frequent depending on the
393 scenario) compared to the observed period (Supplemental Table 3). On the drier side of the
394 distribution, we see dramatic shifts in occurrence for mild, moderate, and extreme droughts
395 (SPEI = -1.0, -1.5, and -2.0 respectively). Regardless of the scenario, all droughts are projected
396 to be much more frequent. The likelihood ratio for the mild droughts result in a 3-20 times
397 increase in occurrence (Supplemental Table 3), depending on the scenario of warming, while the
398 moderate and extreme droughts see dramatic increases but to a lesser extent (3-9 times and 3-5
399 times more likely, respectively) (Supplemental Table 3). Thus, the largest, most likely changes
400 are decreases in mild wet events and increases in all dry events, especially mild droughts.

401 While the effect size is an important feature, the frequency of occurrence of an extreme
402 event is also critical when thinking about the overall impact of extremes over time. Due to their
403 higher frequency during the observed period, mild droughts have a much larger cumulative effect
404 on growth than do more extreme droughts (Figure 6A). Similarly, mild pluvials lead to a larger
405 cumulative effect on growth due the increased frequency of occurrence, albeit a smaller overall
406 effect compared to drought decreases. Of the various forest types, mesic forests are the most
407 affected by changes in extremes with drought having a larger impact than pluvials (Figure 6A
408 and 6B).

409 Increased frequencies of hydroclimatic extremes in the future increase the cumulative
410 effect that drought has on growth (Figure 6C and 6D). All climate scenarios indicate an increase
411 in frequency of all drought categories (Supplemental Figure 6), thus we see increases in the
412 cumulative effect of drought across all scenarios with increasing cumulative effects as warming
413 increases. Conversely, more warming leads to less frequent pluvials (likely driven by water
414 demand), thus we see larger effects in the lower warming scenarios (SSP1-2.6 and SSP2-4.5),

415 but in all cases the effect is smaller than in the observed period. Further, the overall effect size is
 416 substantially smaller than the drought cumulative effect (Figure 6).



417

418 Figure 6: Cumulative effects (effect size multiplied by the number of events) of hydroclimate
 419 extremes on species growth for the observed (1901-2016) across each drought (A) and pluvial
 420 (B) thresholds and for each future (2016-2100) climate scenario of the 75th, 50th, and 25th
 421 quantities for droughts (C) and pluvials (D).

422

423 Discussion

424

425 We found that the impacts of drought on growth are larger than the pluvial effects
 426 (Figures 2 and 6), supporting assertions that nonlinear climate-growth responses are predominant
 427 in forests (Dannenberg et al. 2019) including in the eastern US (Rollinson et al. 2021; Anderson-
 428 Teixeira et al. 2022). If we assumed a linear relationship, the drought response would have been
 429 underestimated (by two-three times), and the pluvial response would have been overestimated

430 compared to the nonlinearly estimated responses (Supplemental Table 1). These findings
431 highlight the importance of accurately modeling the asymmetric climate-growth responses of
432 trees, with implications for estimating impacts to carbon sequestration. Such asymmetric
433 responses across the five studied species indicate that carbon lost from drought-induced radial
434 growth declines is not compensated by increases in growth during wet periods in the eastern US
435 (Figures 2 and 6).

436 The magnitude of the growth response to drought reveals species-specific responses for
437 all drought intensities. The effect size and the difference of the mean effect size between species
438 increased with drought intensity (Figure 2). Previous research on species-specific growth
439 responses to drought demonstrates mixed results. Some studies found little to no difference in
440 growth responses to drought among species in eastern US forests (e.g., LeBlanc and Terrell
441 2011; Martin-Benito and Pederson 2015). These studies often used correlation analyses and
442 compared species from different sites. The sensitivity of growth to drought from a given species
443 is dependent on the site climate, with individuals growing in the wettest portion of the range
444 being more sensitive (Heilmayr et al. 2023). Thus, comparing responses between species situated
445 across various locations of their respective range can confound analysis of their drought
446 sensitivity. Other studies that focused on the magnitude of the growth response and examined
447 species that were located in the same landscape positions found that species-specific growth
448 responses to drought depended on water-use strategy, with those being more conservative
449 (isohydric) having greater growth sensitivity to drought (Brzostek et al. 2014; Elliot et al. 2015;
450 Au et al. 2020; Novick et al. 2022; Lockwood et al. 2023). While others have hypothesized that
451 species-specific differences in growth response to drought intensity would increase in a warming
452 climate (e.g., Elliot et al. 2015), little to no evidence has been presented to support that

453 hypothesis. Here, we find support for this hypothesis, finding that species-specific responses
454 increase with drought intensity, likely due to changes in the water table (Brzostek et al. 2014).
455 During mild droughts, all trees have some access to water and thus have smaller growth
456 decreases. During extreme droughts, all species are impacted but those with shallow roots, such
457 as *L. tulipifera* and *A. saccharum* (Bzostek et al. 2014), have an even larger decrease in growth
458 (Figure 2).

459 While species-specific responses are prevalent in the moderate and extreme drought
460 intensities, the cumulative impact of mild droughts on growth are larger due to their greater
461 frequency (Figure 6). Eastern US forests are composed of several species that employ various
462 water-use strategies, thus drought conditions can impact certain forest types more than others.
463 Forest stands that have a larger component of mesic species, such as *A. saccharum* and *L.*
464 *tulipifera*, have larger growth reductions during moderate and extreme droughts compared to
465 those stands with larger proportions of xeric species, such as those in the *Quercus* and *Carya*
466 genera (Figure 6). However, we found the growth differential percentage from the mesic species
467 is higher and thus growth after drought is closer to the pre-drought growth compared to xeric
468 species (Figure 4). This is in part because all species returned close to pre-drought growth after
469 one year, resulting in larger growth differential from the mesic species because they had larger
470 drought effects. Nevertheless, drought will have a larger impact on growth in forests with mesic
471 species but with shorter drought legacies in the years following drought. These nonlinear and
472 species-specific growth responses are important to include in vegetation models to increase our
473 ability to predict how climate will impact forests in the future.

474 Climate models project a drier climate with more frequent drought in the future in eastern
475 US forests across all scenarios (Supplemental Figure 6), even those with aggressive mitigation

476 (e.g., SSP1-2.6). Given that even moderate droughts have a large impact on species-specific
477 growth responses, the future is very likely to see an increase in differential responses of forest
478 growth to drought, making understanding species responses to climate even more important in
479 the future. In the higher emissions scenarios (SSP3-7.0 and SSP5-8.5), increases in the frequency
480 of extreme drought could have a cumulative effect ~five-times greater than that of the observed
481 period. Depending on the emission scenario of the future, the relative growth after extremes
482 could also see more species-specific responses (Figure 5). The higher emissions scenarios show
483 mesic species returning to pre-drought growth better than the xeric species. Thus, future
484 warming will impact both the growth response during and after extremes, but the intensity of
485 future droughts will determine the degree of species-specific impacts from drought and thus the
486 impact to forest stands with various species composition. We note that this study does not
487 account for any acclimation/adaptation of a given species or new species compositions,
488 something important to consider as we try to better understand how forests will respond to
489 ongoing climate change.

490 Stem growth is the main above-ground carbon pool (Fahey et al. 2010) and thus, tree-ring
491 responses to climate have large carbon implications (Babst et al. 2014). Compared to variations
492 in photosynthetic processes, drought has a much larger impact on growth (Cabon et al. 2022;
493 Dow et al. 2022) and mortality (Martinez-Vilalta et al. 2019), so drought particularly affects long
494 residence carbon stored from growth and has the potential to reduce the residence time of carbon
495 and impact the amount of carbon that forests sequester (Kannenberg et al. 2022). Our results
496 indicate that as the demographic shift from xeric to mesic species continues throughout large
497 portions of the eastern US (Abrams 2003; McEwan et al. 2011; Novick et al. 2022), drought will
498 have a larger impact on growth and thus carbon storage. While drought induced mortality is

499 relatively less common in the mesic eastern US, mortality remains an important component of
500 the carbon budget. However, mortality tends to be uncorrelated with drought-driven growth
501 reductions (Novick et al. 2022). The need to better understand the role of mortality in the overall
502 above-ground carbon budget is critical to get a more complete picture of how climate impacts
503 carbon pools in mesic forests.

504

505 Conclusions

506 Across deciduous forests of the eastern US, we found that mesic species such as *L.*
507 *tulipifera* and *A. saccharum* were more sensitive to drought across all drought intensities.
508 Further, growth responded asymmetrically to drought, with the positive growth response to
509 pluvials failing to outweigh reductions of growth during drought. When accounting for forest
510 species composition, forests dominated by mesic species show greater reductions in growth
511 during drought but also higher growth differential percentages. Thus, the ongoing increase in
512 mesic species in eastern US forests in combination with the likely increase in drought conditions,
513 suggest that drought will likely have a larger impact on the carbon uptake in the future in the
514 eastern US. However, these same forests showed higher growth differential percentages
515 following droughts compared to those that are dominated by xeric species. Even with the higher
516 growth after drought, a warmer future with more frequent droughts along with more mesic
517 species will result in more drought-induced carbon losses from a forest that is a tremendous
518 carbon sink.

519

520

521

522 References

523 Abrams, M. D. (2003). Where Has All the White Oak Gone? *BioScience*, 53(10), 927–939.
524 [https://doi.org/10.1641/0006-3568\(2003\)053\[0927:WHATWO\]2.0.CO;2](https://doi.org/10.1641/0006-3568(2003)053[0927:WHATWO]2.0.CO;2)

525 Alexander, M. R., Rollinson, C. R., Babst, F., Trouet, V., & Moore, D. J. P. (2018). Relative
526 influences of multiple sources of uncertainty on cumulative and incremental tree-ring-
527 derived aboveground biomass estimates. *Trees*, 32(1), 265–276.
528 <https://doi.org/10.1007/s00468-017-1629-0>

529 Allen, C. D., Macalady, A. K., Chenchouni, H., Bachelet, D., McDowell, N., Vennetier, M.,
530 Kitzberger, T., Rigling, A., Breshears, D. D., Hogg, E. H. (Ted), Gonzalez, P., Fensham,
531 R., Zhang, Z., Castro, J., Demidova, N., Lim, J.-H., Allard, G., Running, S. W., Semerci,
532 A., & Cobb, N. (2010). A global overview of drought and heat-induced tree mortality
533 reveals emerging climate change risks for forests. *Forest Ecology and Management*,
534 259(4), 660–684. <https://doi.org/10.1016/j.foreco.2009.09.001>

535 Allen, R. G., Pruitt, W. O., Wright, J. L., Howell, T. A., Ventura, F., Snyder, R., Itenfisu, D.,
536 Steduto, P., Berengena, J., Yrisarry, J. B., Smith, M., Pereira, L. S., Raes, D., Perrier, A.,
537 Alves, I., Walter, I., & Elliott, R. (2006). A recommendation on standardized surface
538 resistance for hourly calculation of reference ETo by the FAO56 Penman-Monteith
539 method. *Agricultural Water Management*, 81(1), 1–22.
540 <https://doi.org/10.1016/j.agwat.2005.03.007>

541 Anderegg, W. R. L., Konings, A. G., Trugman, A. T., Yu, K., Bowling, D. R., Gabbitas, R.,
542 Karp, D. S., Pacala, S., Sperry, J. S., Sulman, B. N., & Zenes, N. (2018). Hydraulic
543 diversity of forests regulates ecosystem resilience during drought. *Nature*, 561(7724),
544 538–541. <https://doi.org/10.1038/s41586-018-0539-7>

545 Anderson-Teixeira, K. J., Herrmann, V., Rollinson, C. R., Gonzalez, B., Gonzalez-Akre, E. B.,
546 Pederson, N., Alexander, M. R., Allen, C. D., Alfaro-Sánchez, R., Awada, T., Baltzer, J.
547 L., Baker, P. J., Birch, J. D., Bunyavejchewin, S., Cherubini, P., Davies, S. J., Dow, C.,
548 Helcoski, R., Kašpar, J., ... Zuidema, P. A. (2022). Joint effects of climate, tree size, and
549 year on annual tree growth derived from tree-ring records of ten globally distributed
550 forests. *Global Change Biology*, 28(1), 245–266. <https://doi.org/10.1111/gcb.15934>

551 Au, T. F., & Maxwell, J. T. (2022). Drought Sensitivity and Resilience of Oak–Hickory Stands
552 in the Eastern United States. *Forests*, 13(3), Article 3. <https://doi.org/10.3390/f13030389>

553 Au, T. F., Maxwell, J. T., Novick, K. A., Robeson, S. M., Warner, S. M., Lockwood, B. R.,
554 Phillips, R. P., Harley, G. L., Telewski, F. W., Therrell, M. D., & Pederson, N. (2020).
555 Demographic shifts in eastern US forests increase the impact of late-season drought on
556 forest growth. *Ecography*, 43(10), 1475–1486. <https://doi.org/10.1111/ecog.05055>

557 Au, T. F., Maxwell, J. T., Robeson, S. M., Li, J., Siani, S. M. O., Novick, K. A., Dannenberg, M.,
558 Phillips, R. P., Li, T., Chen, Z., & Lenoir, J. (2022). Younger trees in the upper
559 canopy are more sensitive but also more resilient to drought. *Nature Climate Change*,
560 12(12), 1168–1174. <https://doi.org/10.1038/s41558-022-01528-w>

561 Babst, F., Alexander, M. R., Szejner, P., Bouriaud, O., Klesse, S., Roden, J., Ciais, P., Poulter,
562 B., Frank, D., Moore, D. J. P., & Trouet, V. (2014). A tree-ring perspective on the
563 terrestrial carbon cycle. *Oecologia*, 176(2), 307–322. <https://doi.org/10.1007/s00442-014-3031-6>

564 Barnes, M. L., Zhang, Q., Robeson, S. M., Young, L., Burakowski, E. A., Oishi, A. Christopher.,
565 Stoy, P. C., Katul, G., & Novick, K. A. (2024). A Century of Reforestation Reduced

567 Anthropogenic Warming in the Eastern United States. *Earth's Future*, 12(2),
568 e2023EF003663. <https://doi.org/10.1029/2023EF003663>

569 Beguería, S., Vicente-Serrano, S. M., Reig, F., & Latorre, B. (2014). Standardized precipitation
570 evapotranspiration index (SPEI) revisited: Parameter fitting, evapotranspiration models,
571 tools, datasets and drought monitoring. *International Journal of Climatology*, 34(10),
572 3001–3023. <https://doi.org/10.1002/joc.3887>

573 Bennett, L. T., Bruce, M. J., Machunter, J., Kohout, M., Krishnaraj, S. J., & Aponte, C. (2017).
574 Assessing fire impacts on the carbon stability of fire-tolerant forests. *Ecological
575 Applications*, 27(8), 2497–2513.

576 Breshears, D. D., Cobb, N. S., Rich, P. M., Price, K. P., Allen, C. D., Balice, R. G., Romme, W.
577 H., Kastens, J. H., Floyd, M. L., Belnap, J., Anderson, J. J., Myers, O. B., & Meyer, C.
578 W. (2005). Regional vegetation die-off in response to global-change-type drought.
579 *Proceedings of the National Academy of Sciences*, 102(42), 15144–15148.
580 <https://doi.org/10.1073/pnas.0505734102>

581 Brzostek, E. R., Dragoni, D., Schmid, H. P., Rahman, A. F., Sims, D., Wayson, C. A., Johnson,
582 D. J., & Phillips, R. P. (2014). Chronic water stress reduces tree growth and the carbon
583 sink of deciduous hardwood forests. *Global Change Biology*, 20(8), 2531–2539.
584 <https://doi.org/10.1111/gcb.12528>

585 Bussberg, N. W., Maxwell, J. T., Robeson, S. M., & Huang, C. (2020). The effect of end-point
586 adjustments on smoothing splines used for tree-ring standardization. *Dendrochronologia*,
587 60, 125665. <https://doi.org/10.1016/j.dendro.2020.125665>

588 Cabon, A., Kannenberg, S. A., Arain, A., Babst, F., Baldocchi, D., Belmecheri, S., Delpierre, N.,
589 Guerrieri, R., Maxwell, J. T., McKenzie, S., Meinzer, F. C., Moore, D. J. P., Pappas, C.,

590 Rocha, A. V., Szejner, P., Ueyama, M., Ulrich, D., Vincke, C., Voelker, S. L., ...

591 Anderegg, W. R. L. (2022). Cross-biome synthesis of source versus sink limits to tree

592 growth. *Science*, 376(6594), 758–761. <https://doi.org/10.1126/science.abm4875>

593 Cook, E.R. (1985). *A Time Series Analysis Approach to Tree Ring Standardization*. The

594 University of Arizona

595 Cook, E. R., & Peters, K. (1981). *The Smoothing Spline: A New Approach to Standardizing*

596 *Forest Interior Tree-Ring Width Series for Dendroclimatic Studies*.

597 <https://repository.arizona.edu/handle/10150/261038>

598 Costanza, J. K., Koch, F. H., & Reeves, M. C. (2023). Future exposure of forest ecosystems to

599 multi-year drought in the United States. *Ecosphere*, 14(5), e4525.

600 <https://doi.org/10.1002/ecs2.4525>

601 Dannenberg, M. P., Wise, E. K., & Smith, W. K. (2019). Reduced tree growth in the semiarid

602 United States due to asymmetric responses to intensifying precipitation extremes. *Science*

603 *Advances*, 5(10), eaaw0667. <https://doi.org/10.1126/sciadv.aaw0667>

604 Delpierre, N., Vitasse, Y., Chuine, I., Guillemot, J., Bazot, S., Rutishauser, T., & Rathgeber, C.

605 B. K. (2016). Temperate and boreal forest tree phenology: From organ-scale processes to

606 terrestrial ecosystem models. *Annals of Forest Science*, 73(1), 5–25.

607 <https://doi.org/10.1007/s13595-015-0477-6>

608 D'Orangeville, L., Maxwell, J., Kneeshaw, D., Pederson, N., Duchesne, L., Logan, T., Houle, D.,

609 Arseneault, D., Beier, C. M., Bishop, D. A., Druckenbrod, D., Fraver, S., Girard, F.,

610 Halman, J., Hansen, C., Hart, J. L., Hartmann, H., Kaye, M., Leblanc, D., ... Phillips, R.

611 P. (2018). Drought timing and local climate determine the sensitivity of eastern temperate

612 forests to drought. *Global Change Biology*, 24(6), 2339–2351.

613 <https://doi.org/10.1111/gcb.14096>

614 Dow, C., Kim, A. Y., D'Orangeville, L., Gonzalez-Akre, E. B., Helcoski, R., Herrmann, V.,

615 Harley, G. L., Maxwell, J. T., McGregor, I. R., McShea, W. J., McMahon, S. M.,

616 Pederson, N., Tepley, A. J., & Anderson-Teixeira, K. J. (2022). Warm springs alter

617 timing but not total growth of temperate deciduous trees. *Nature*, 608(7923), 552–557.

618 <https://doi.org/10.1038/s41586-022-05092-3>

619 Dye, A., Barker Plotkin, A., Bishop, D., Pederson, N., Poulter, B., & Hessl, A. (2016).

620 Comparing tree-ring and permanent plot estimates of aboveground net primary

621 production in three eastern U.S. forests. *Ecosphere*, 7(9), e01454.

622 <https://doi.org/10.1002/ecs2.1454>

623 Elliott, K. J., Miniat, C. F., Pederson, N., & Laseter, S. H. (2015). Forest tree growth response to

624 hydroclimate variability in the southern Appalachians. *Global Change Biology*, 21(12),

625 4627–4641. <https://doi.org/10.1111/gcb.13045>

626 Eyring, V., Bony, S., Meehl, G. A., Senior, C. A., Stevens, B., Stouffer, R. J., & Taylor, K. E.

627 (2016). Overview of the Coupled Model Intercomparison Project Phase 6 (CMIP6)

628 experimental design and organization. *Geoscientific Model Development*, 9(5), 1937–

629 1958. <https://doi.org/10.5194/gmd-9-1937-2016>

630 Fahey, T. J., Woodbury, P. B., Battles, J. J., Goodale, C. L., Hamburg, S. P., Ollinger, S. V., &

631 Woodall, C. W. (2010). Forest carbon storage: Ecology, management, and policy.

632 *Frontiers in Ecology and the Environment*, 8(5), 245–252.

633 <https://doi.org/10.1890/080169>

634 Fei, S., Kong, N., Steiner, K. C., Moser, W. K., & Steiner, E. B. (2011). Change in oak
635 abundance in the eastern United States from 1980 to 2008. *Forest Ecology and*
636 *Management*, 262(8), 1370–1377. <https://doi.org/10.1016/j.foreco.2011.06.030>

637 Ficklin, D. L., Abatzoglou, J. T., Robeson, S. M., & Dufficy, A. (2016). The Influence of
638 Climate Model Biases on Projections of Aridity and Drought. *Journal of Climate*, 29(4),
639 1269–1285. <https://doi.org/10.1175/JCLI-D-15-0439.1>

640 Ficklin, D. L., & Novick, K. A. (2017). Historic and projected changes in vapor pressure deficit
641 suggest a continental-scale drying of the United States atmosphere. *Journal of*
642 *Geophysical Research: Atmospheres*, 122(4), 2061–2079.
643 <https://doi.org/10.1002/2016JD025855>

644 Ford, T. W. (2014). Precipitation anomalies in Eastern-Central Iowa from 1640 – Present.
645 *Journal of Hydrology*, 519, 918–924. <https://doi.org/10.1016/j.jhydrol.2014.08.021>

646 Grossiord, C., Buckley, T. N., Cernusak, L. A., Novick, K. A., Poulter, B., Siegwolf, R. T. W.,
647 Sperry, J. S., & McDowell, N. G. (2020). Plant responses to rising vapor pressure deficit.
648 *New Phytologist*, 226(6), 1550–1566. <https://doi.org/10.1111/nph.16485>

649 Gudmundsson, L., Bremnes, J. B., Haugen, J. E., & Engen-Skaugen, T. (2012). Technical Note:
650 Downscaling RCM precipitation to the station scale using statistical transformations
651 – a comparison of methods. *Hydrology and Earth System Sciences*, 16(9), 3383–
652 3390. <https://doi.org/10.5194/hess-16-3383-2012>

653 Hamrick, J. L. (2004). Response of forest trees to global environmental changes. *Forest Ecology*
654 *and Management*, 197(1), 323–335. <https://doi.org/10.1016/j.foreco.2004.05.023>

655 Heilmayr, R., Dudney, J., & Moore, F. C. (2023). Drought sensitivity in mesic forests heightens
656 their vulnerability to climate change. *Science*, 382(6675), 1171–1177.
657 <https://doi.org/10.1126/science.ad1071>

658 Hoffmann, N., Schall, P., Ammer, C., Leder, B., & Vor, T. (2018). Drought sensitivity and stem
659 growth variation of nine alien and native tree species on a productive forest site in
660 Germany. *Agricultural and Forest Meteorology*, 256–257, 431–444.
661 <https://doi.org/10.1016/j.agrformet.2018.03.008>

662 Holmes, R. L. (1983n.d.). *COMPUTER -ASSISTED QUALITY CONTROL IN TREE -RING
663 DATING AND MEASUREMENT*.

664 Isbell, F., Craven, D., Connolly, J., Loreau, M., Schmid, B., Beierkuhnlein, C., Bezemer, T. M.,
665 Bonin, C., Bruelheide, H., de Luca, E., Ebeling, A., Griffin, J. N., Guo, Q., Hautier, Y.,
666 Hector, A., Jentsch, A., Kreyling, J., Lanta, V., Manning, P., ... Eisenhauer, N. (2015).
667 Biodiversity increases the resistance of ecosystem productivity to climate extremes.
668 *Nature*, 526(7574), 574–577. <https://doi.org/10.1038/nature15374>

669 Jiang, Y., Kim, J. B., Trugman, A. T., Kim, Y., & Still, C. J. (2019). Linking tree physiological
670 constraints with predictions of carbon and water fluxes at an old-growth coniferous
671 forest. *Ecosphere*, 10(4), e02692. <https://doi.org/10.1002/ecs2.2692>

672 Kannenberg, S. A., Cabon, A., Babst, F., Belmecheri, S., Delpierre, N., Guerrieri, R., Maxwell,
673 J. T., Meinzer, F. C., Moore, D. J. P., Pappas, C., Ueyama, M., Ulrich, D. E. M., Voelker,
674 S. L., Woodruff, D. R., & Anderegg, W. R. L. (2022). Drought-induced decoupling
675 between carbon uptake and tree growth impacts forest carbon turnover time. *Agricultural
676 and Forest Meteorology*, 322, 108996. <https://doi.org/10.1016/j.agrformet.2022.108996>

677 Kannenberg, S. A., Maxwell, J. T., Pederson, N., D'Orangeville, L., Ficklin, D. L., & Phillips, R.
678 P. (2019). Drought legacies are dependent on water table depth, wood anatomy and
679 drought timing across the eastern US. *Ecology Letters*, 22(1), 119–127.
680 <https://doi.org/10.1111/ele.13173>

681 Karl, T. R., Knight, R. W., Easterling, D. R., & Quayle, R. G. (1996). Indices of Climate Change
682 for the United States. *Bulletin of the American Meteorological Society*, 77(2), 279–292.
683 [https://doi.org/10.1175/1520-0477\(1996\)077<0279:IOCCFT>2.0.CO;2](https://doi.org/10.1175/1520-0477(1996)077<0279:IOCCFT>2.0.CO;2)

684 LeBlanc, D. C., & Terrell, M. A. (2011). Comparison of growth–climate relationships between
685 northern red oak and white oak across eastern North America. *Canadian Journal of
686 Forest Research*, 41(10), 1936–1947. <https://doi.org/10.1139/x11-118>

687 Lloret, F., Keeling, E. G., & Sala, A. (2011). Components of tree resilience: Effects of
688 successive low-growth episodes in old ponderosa pine forests. *Oikos*, 120(12), 1909–
689 1920. <https://doi.org/10.1111/j.1600-0706.2011.19372.x>

690 Lockwood, B. R., Maxwell, J. T., Denham, S. O., Robeson, S. M., LeBlanc, D. C., Pederson, N.,
691 Novick, K. A., & Au, T. F. (2023). Interspecific differences in drought and pluvial
692 responses for *Quercus alba* and *Quercus rubra* across the eastern United States.
693 *Agricultural and Forest Meteorology*, 340, 109597.
694 <https://doi.org/10.1016/j.agrformet.2023.109597>

695 Martin-Benito, D., & Pederson, N. (2015). Convergence in drought stress, but a divergence of
696 climatic drivers across a latitudinal gradient in a temperate broadleaf forest. *Journal of
697 Biogeography*, 42(5), 925–937. <https://doi.org/10.1111/jbi.12462>

698 Martinez-Vilalta, J., Anderegg, W. R. L., Sapes, G., & Sala, A. (2019). Greater focus on water
699 pools may improve our ability to understand and anticipate drought-induced mortality in
700 plants. *New Phytologist*, 223(1), 22–32. <https://doi.org/10.1111/nph.15644>

701 Maxwell, J. T., & Harley, G. L. (2017). Increased tree-ring network density reveals more precise
702 estimations of sub-regional hydroclimate variability and climate dynamics in the
703 Midwest, USA. *Climate Dynamics*, 49(4), 1479–1493. <https://doi.org/10.1007/s00382-016-3396-9>

704 Maxwell, J. T., Harley, G. L., Matheus, T. J., Strange, B. M., Van Aken, K., Au, T. F., & Bregy,
705 J. C. (2020). Sampling density and date along with species selection influence spatial
706 representation of tree-ring reconstructions. *Climate of the Past*, 16(5), 1901–1916.
707 <https://doi.org/10.5194/cp-16-1901-2020>

708 McDowell, N. G., Allen, C. D., Anderson-Teixeira, K., Aukema, B. H., Bond-Lamberty, B.,
709 Chini, L., Clark, J. S., Dietze, M., Grossiord, C., Hanbury-Brown, A., Hurtt, G. C.,
710 Jackson, R. B., Johnson, D. J., Kueppers, L., Lichstein, J. W., Ogle, K., Poulter, B., Pugh,
711 T. A. M., Seidl, R., ... Xu, C. (2020). Pervasive shifts in forest dynamics in a changing
712 world. *Science*, 368(6494), eaaz9463. <https://doi.org/10.1126/science.aaz9463>

713 McDowell, N., Pockman, W. T., Allen, C. D., Breshears, D. D., Cobb, N., Kolb, T., Plaut, J.,
714 Sperry, J., West, A., Williams, D. G., & Yepez, E. A. (2008). Mechanisms of plant
715 survival and mortality during drought: Why do some plants survive while others succumb
716 to drought? *New Phytologist*, 178(4), 719–739. <https://doi.org/10.1111/j.1469-8137.2008.02436.x>

719 McEwan, R. W., Dyer, J. M., & Pederson, N. (2011). Multiple interacting ecosystem drivers:
720 Toward an encompassing hypothesis of oak forest dynamics across eastern North
721 America. *Ecography*, 34(2), 244–256. <https://doi.org/10.1111/j.1600-0587.2010.06390.x>

722 Novick, K. A., Ficklin, D. L., Grossiord, C., Konings, A. G., Martínez-Vilalta, J., Sadok, W.,
723 Trugman, A. T., Williams, A. P., Wright, A. J., Abatzoglou, J. T., Dannenberg, M. P.,
724 Gentine, P., Guan, K., Johnston, M. R., Lowman, L. E. L., Moore, D. J. P., & McDowell,
725 N. G. (2024n.d.). The impacts of rising vapour pressure deficit in natural and managed
726 ecosystems. *Plant, Cell & Environment*, n/a(n/a). <https://doi.org/10.1111/pce.14846>

727 Novick, K., Jo, I., D'Orangeville, L., Benson, M., Au, T. F., Barnes, M., Denham, S., Fei, S.,
728 Heilman, K., Hwang, T., Keyser, T., Maxwell, J., Miniat, C., McLachlan, J., Pederson,
729 N., Wang, L., Wood, J. D., & Phillips, R. P. (2022). The Drought Response of Eastern
730 US Oaks in the Context of Their Declining Abundance. *BioScience*, 72(4), 333–346.
731 <https://doi.org/10.1093/biosci/biab135>

732 O'Neill, B. C., Tebaldi, C., van Vuuren, D. P., Eyring, V., Friedlingstein, P., Hurtt, G., Knutti,
733 R., Kriegler, E., Lamarque, J.-F., Lowe, J., Meehl, G. A., Moss, R., Riahi, K., &
734 Sanderson, B. M. (2016). The Scenario Model Intercomparison Project (ScenarioMIP)
735 for CMIP6. *Geoscientific Model Development*, 9(9), 3461–3482.
736 <https://doi.org/10.5194/gmd-9-3461-2016>

737 Pan, Y., Chen, J. M., Birdsey, R., McCullough, K., He, L., & Deng, F. (2011). Age structure and
738 disturbance legacy of North American forests. *Biogeosciences*, 8(3), 715–732.
739 <https://doi.org/10.5194/bg-8-715-2011>

740 Pederson, N. A. (2005n.d.). *Climatic sensitivity and growth of southern temperate trees in the*
741 *eastern United States: Implications for the carbon cycle* [Ph.D., Columbia University].

742 Retrieved March 13, 2024, from

743 <http://www.proquest.com/docview/305015289/abstract/4BB17745306E4C38PQ/1>

744 Pederson, N., Bell, A. R., Cook, E. R., Lall, U., Devineni, N., Seager, R., Eggleston, K., &

745 Vranes, K. P. (2013). Is an Epic Pluvial Masking the Water Insecurity of the Greater New

746 York City Region?., *Journal of Climate*, 26(4), 1339–1354. <https://doi.org/10.1175/JCLI-D-11-00723.1>

747

748 Robeson, S. M., Maxwell, J. T., & Ficklin, D. L. (2020). Bias Correction of Paleoclimatic

749 Reconstructions: A New Look at 1,200+ Years of Upper Colorado River Flow.

750 *Geophysical Research Letters*, 47(1), e2019GL086689.

751 <https://doi.org/10.1029/2019GL086689>

752 Rollinson, C. R., Alexander, M. R., Dye, A. W., Moore, D. J. P., Pederson, N., & Trouet, V.

753 (2021). Climate sensitivity of understory trees differs from overstory trees in temperate

754 mesic forests. *Ecology*, 102(3), e03264. <https://doi.org/10.1002/ecy.3264>

755 Roman, D. T., Novick, K. A., Brzostek, E. R., Dragoni, D., Rahman, F., & Phillips, R. P. (2015).

756 The role of isohydric and anisohydric species in determining ecosystem-scale response to

757 severe drought. *Oecologia*, 179(3), 641–654. <https://doi.org/10.1007/s00442-015-3380-9>

758 Schwarz, J., Skidmore, G., Kohler, M., Kunz, J., Schnabel, F., Vitali, V., & Bauhus, J. (2020).

759 Quantifying Growth Responses of Trees to Drought—A Critique of Commonly Used

760 Resilience Indices and Recommendations for Future Studies. *Current Forestry Reports*,

761 6(3), 185–200. <https://doi.org/10.1007/s40725-020-00119-2>

762 Vicente-Serrano, S. M., Beguería, S., & López-Moreno, J. I. (2010). A Multiscalar Drought

763 Index Sensitive to Global Warming: The Standardized Precipitation Evapotranspiration

764 Index. *Journal of Climate*, 23(7), 1696–1718. <https://doi.org/10.1175/2009JCLI2909.1>

765 Wehner, M. F., Arnold, J. R., Knutson, T., Kunkel, K. E., & LeGrande, A. N. (2017). *Chapter 8: Droughts, Floods, and Wildfires*. <https://doi.org/10.7930/J0CJ8BNN>

766

767 Williams, A.P., Allen, C. D., Macalady, A. K., Griffin, D., Woodhouse, C. A., Meko, D. M.,

768 Swetnam, T. W., Rauscher, S. A., Seager, R., Grissino-Mayer, H. D., Dean, J. S., Cook,

769 E. R., Gangodagamage, C., Cai, M., & McDowell, N. G. (2013). Temperature as a potent

770 driver of regional forest drought stress and tree mortality. *Nature Climate Change*, 3(3),

771 292–297. <https://doi.org/10.1038/nclimate1693>

772 Yi, K., Maxwell, J. T., Wenzel, M. K., Roman, D. T., Sauer, P. E., Phillips, R. P., & Novick, K.

773 A. (2019). Linking variation in intrinsic water-use efficiency to isohydricity: A

774 comparison at multiple spatiotemporal scales. *New Phytologist*, 221(1), 195–208.

775 <https://doi.org/10.1111/nph.15384>

776

777 Acknowledgements

778 We thank Josh Bregy, James Dickens, James McGee, Josh Oliver, Karly Schmidt-Simard, Brynn

779 Taylor, Brandon Strange, Michael Thornton, Senna Robeson, Matt Wenzel, and Luke Wylie for

780 assistance in the field and lab. We also would like to acknowledge the following funding:

781 USDA Agriculture and Food Research Initiative grant 2017-67013-26191, National Science

782 Foundation, P2C2 Program AGS-1805617 and AGS-1805276, the Harvard Forest Bullard

783 Fellowship, Harvard University, and Indiana University Vice Provost for Research Faculty

784 Research Program.

785

786 Author contribution: Justin T. Maxwell, Tsun Fung Au, Steven A. Kannenberg, and Neil

787 Pederson conceived the ideas and designed methodology; Justin T. Maxwell, Grant L. Harley

788 Tsun Fung Au, Benjamin Lockwood, Matthew P. Dannenberg, and Neil Pederson collected the
789 data; Justin T. Maxwell, Grant L. Harley, Darren L. Ficklin, and Neil Pederson organized and
790 analyzed the data; Justin T. Maxwell led the writing of the manuscript. All authors contributed
791 critically to the drafts and gave final approval for publication.

For Review Only

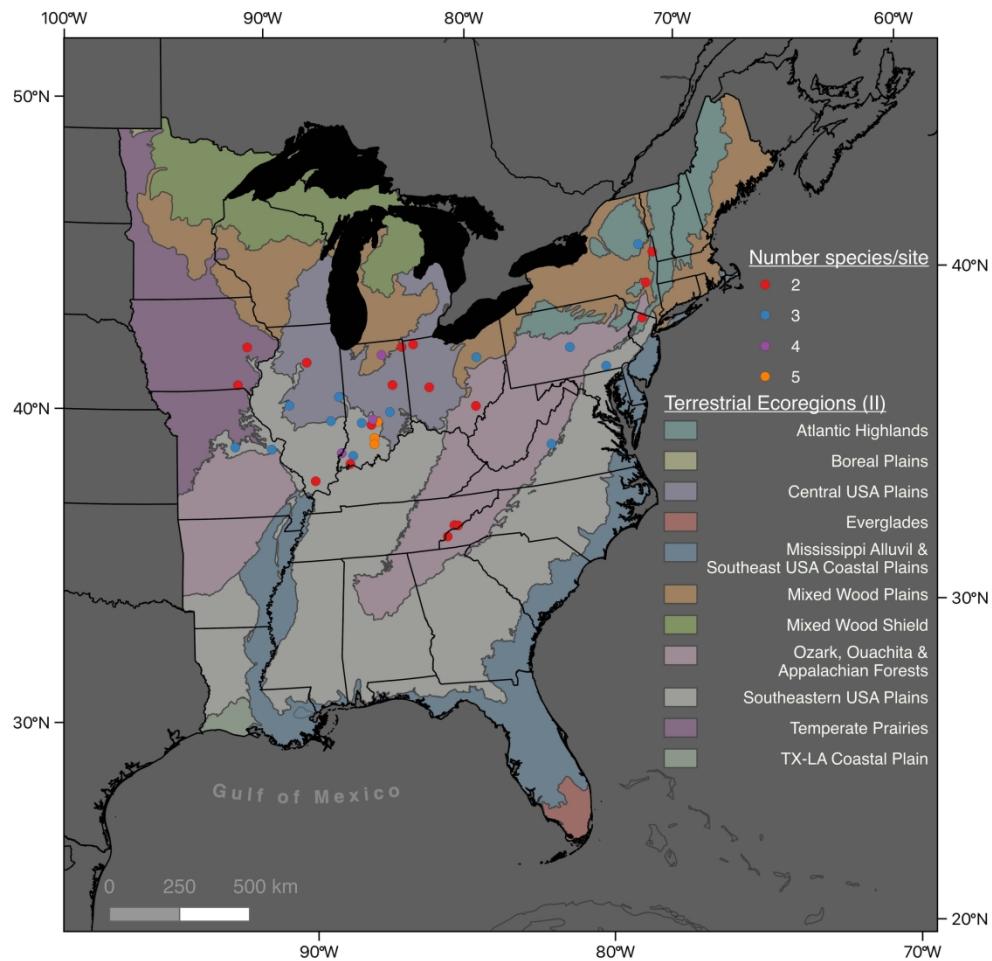


Figure 1: Tree-ring study sites across the Eastern Deciduous Forest biome. Map of tree-ring sites showing the number of species-specific chronologies per study site along with the level two terrestrial ecoregions as defined by Commission for Environmental Cooperation (<http://www.cec.org/north-american-environmental-atlas/>).

212x209mm (300 x 300 DPI)

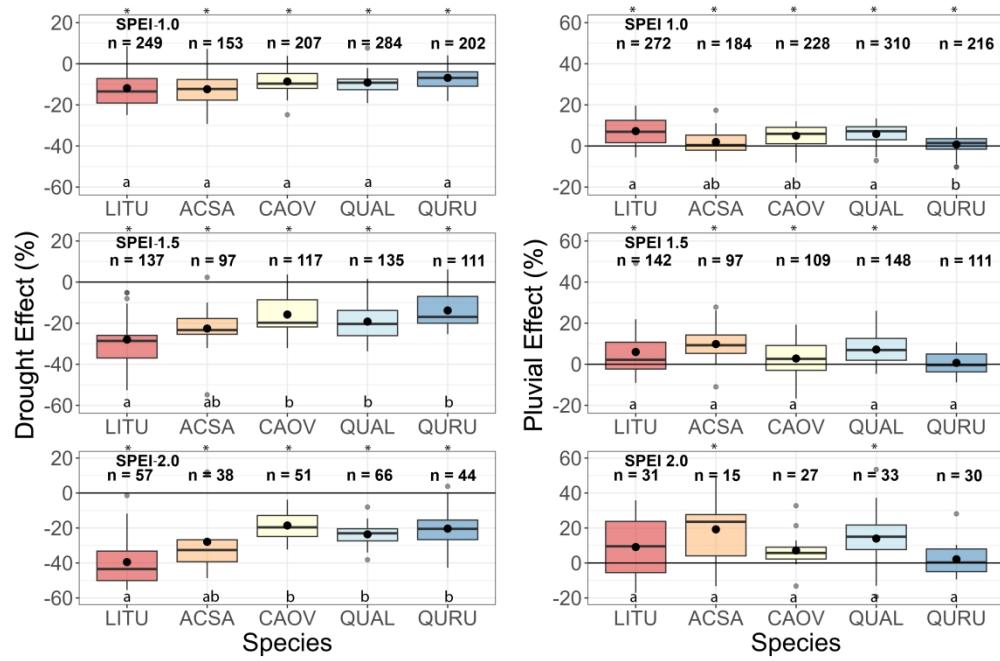


Figure 2: Species-level responses to hydroclimate extremes in eastern US forests. Effects of species growth to drought (left) and pluvial (right) conditions for seasonal (March-August average; August SPEI6) hydroclimate conditions for mild (SPEI6 = ± 1.0 ; top), moderate (SPEI6 = ± 1.5 ; middle), and extreme (SPEI6 = ± 2.0 ; bottom) events. Lower-case lettering represents statistical significance differences in effect size between species via an ANOVA analysis Tukey HSD post hoc test. Asterisks represent the mean is significantly ($p \leq 0.05$) different from zero using a one-sample t-test. The sample size of the number of extremes experienced by each species is denoted. LITU = *L. tulipifera*; ACSA = *A. saccharum*; CAOV = *C. ovata*; QUAL = *Q. alba*; QURU = *Q. rubra*.

776x511mm (118 x 118 DPI)

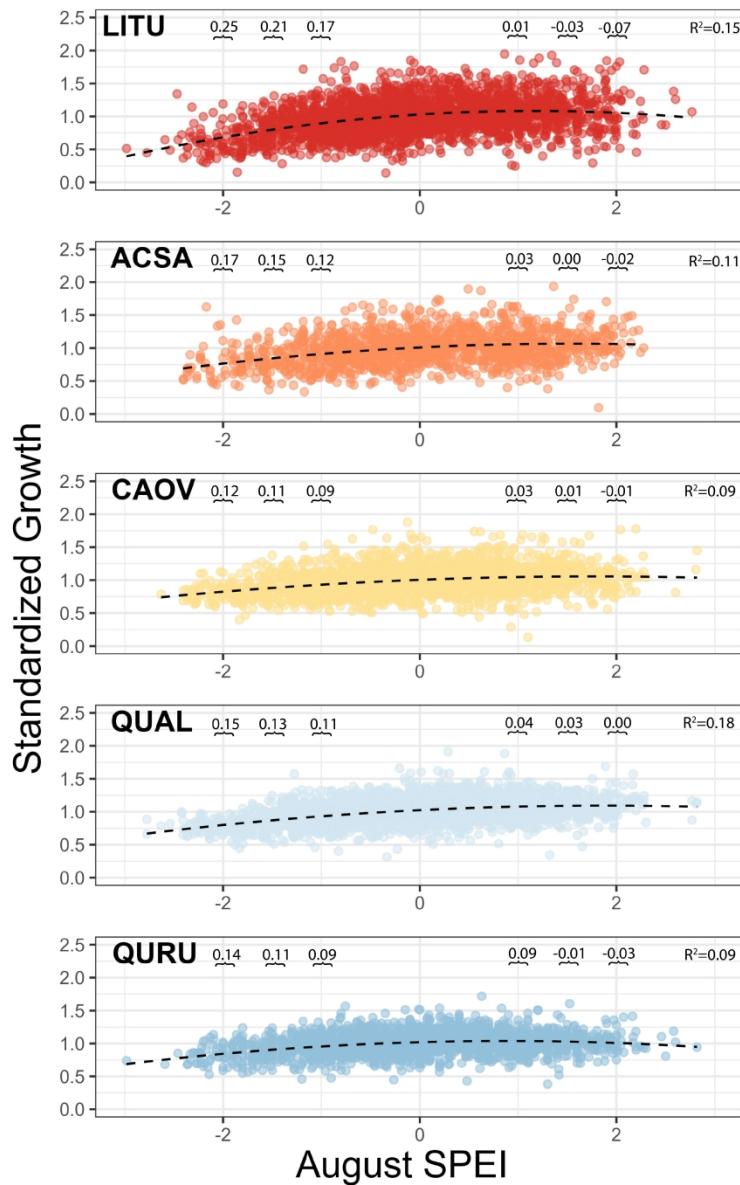


Figure 3: The relationship between standardized growth and March-August SPEI (August SPEI6) values for each species. Dashed line is the ordinary least-squares quadratic regression whose slope ($\frac{\Delta \text{SRW}}{\Delta \text{SPEI}}$) at SPEI values of -2.0, -1.5, -1.0, 1.0, 1.5, and 2.0 along with R^2 are given in each panel.

2722x4355mm (28 x 28 DPI)

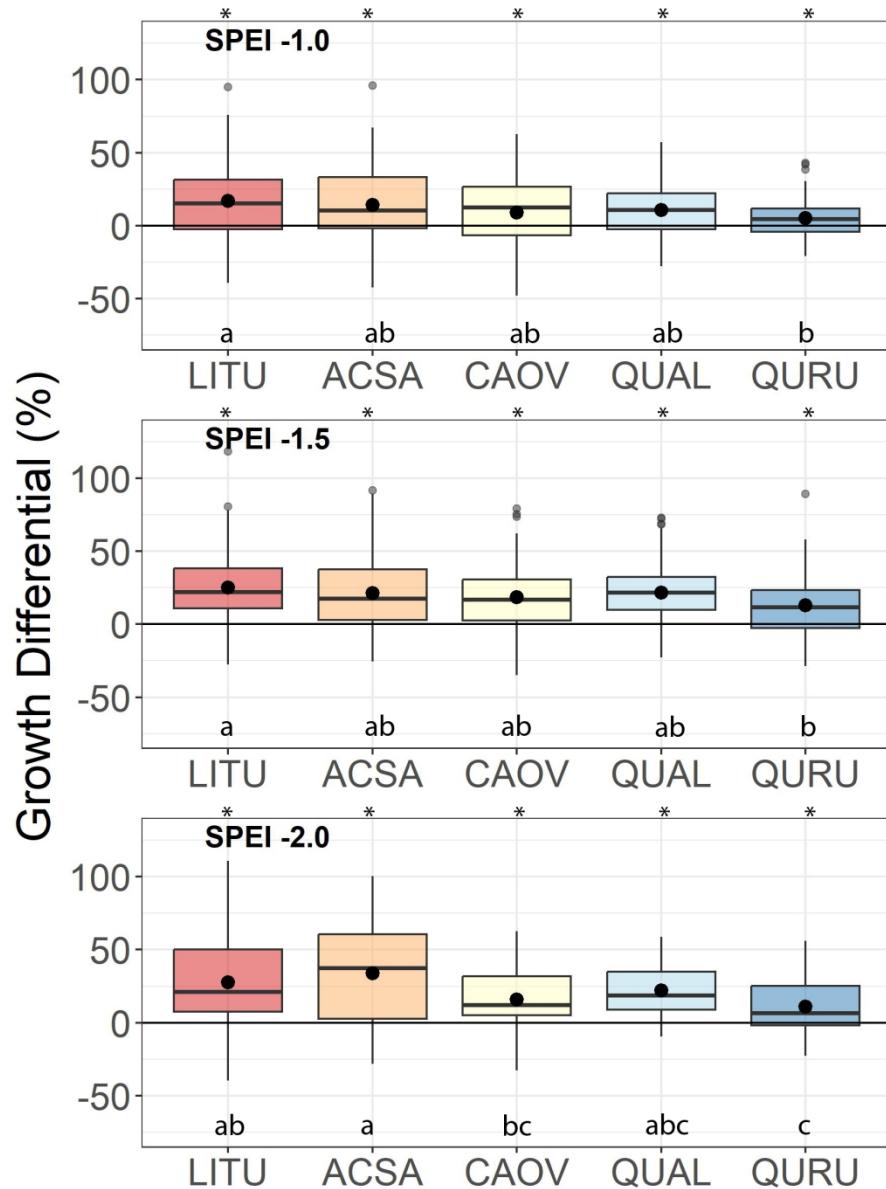


Figure 4: Growth differential percentage from drought index for each species. The growth differential percentage is averaged from the two years after drought and accounts for the weighting of drought impacts on growth. A higher mean growth differential percentage indicates growth closer to the pre-drought conditions. Lettering represents statistical significance differences in effect size between species via an ANOVA analysis Tukey HSD post hoc test. Asterisks represent the mean is significantly higher than zero using a one-sample t-test.

381x510mm (118 x 118 DPI)

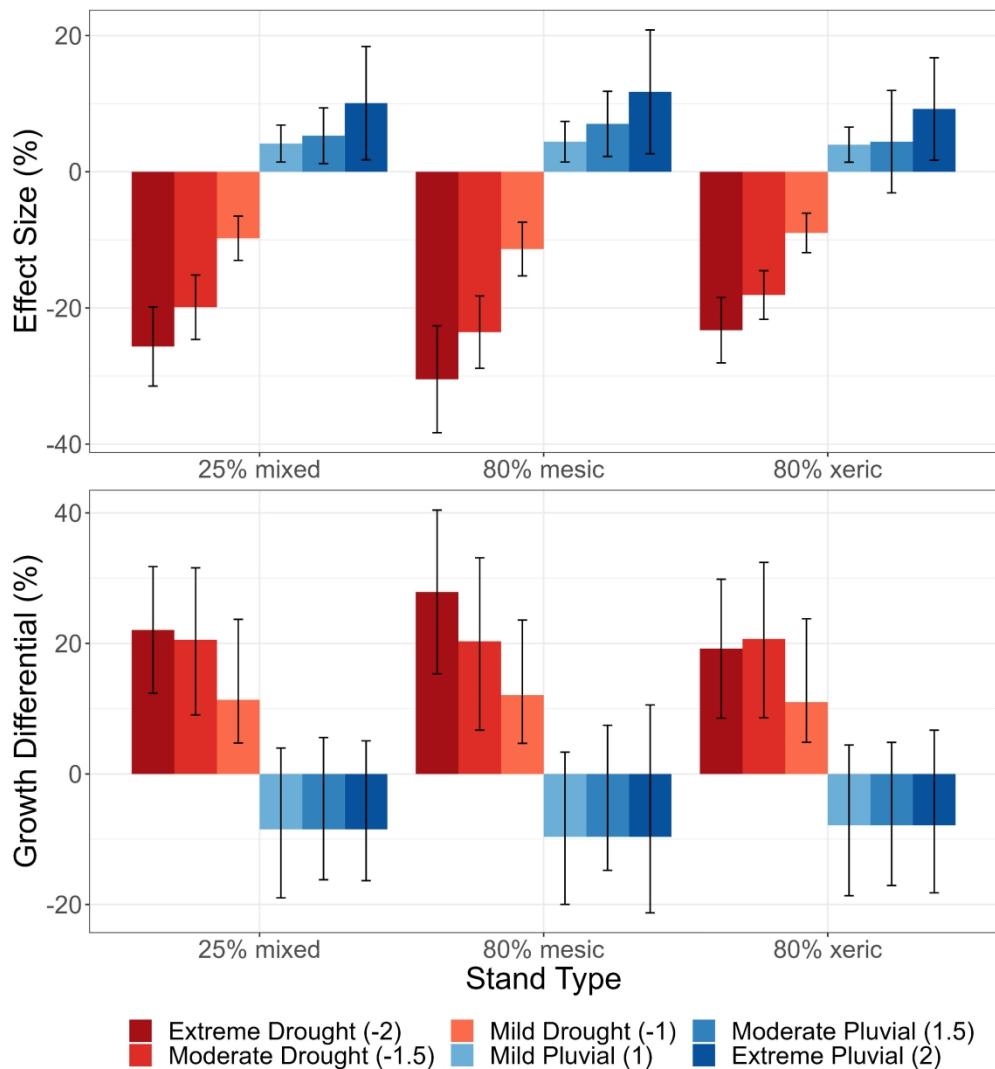


Figure 5: Effect size and growth differential percentage for Eastern Deciduous Forest species under hydroclimate extremes. Top) Pluvial and drought effect sizes for scaled canopy dominant species composition scenarios across intensity thresholds, with error bars. Bottom) Growth differential percentage percentages for scaled canopy dominant species composition scenarios across intensity thresholds, with error bars. Xeric = 80% of canopy dominant trees are *Quercus* and *Carya*; Mesic = 80% of trees are *Acer* and *Liriodendron*; Mixed = *Quercus*, *Carya*, *Liriodendron*, and *Acer* each make up 25% of the forest.

250x269mm (300 x 300 DPI)

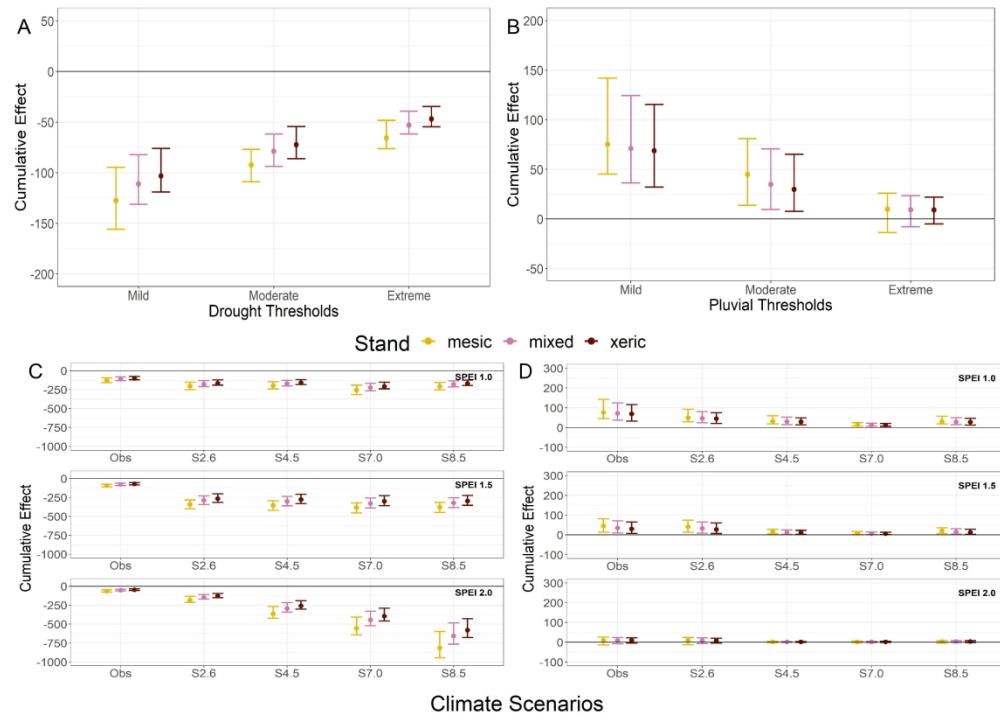


Figure 6: Cumulative effects (effect size multiplied by the number of events) of hydroclimate extremes on species growth for the observed (1901-2016) across each drought (A) and pluvial (B) thresholds and for each future (2016-2100) climate scenario of the 75th, 50th, and 25th quantiles for droughts (C) and pluvials (D).

363x258mm (300 x 300 DPI)