

1 Radial growth decline in a tropical Andean treeline in Bolivia

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21 **Abstract.** The impact of rising temperatures on tropical treeline ecosystems remains understudied. Here we report on the
22 growth history of *Polylepis pepeii* BB.Simpson trees growing in Madidi National Park, a tropical forest setting at elevational
23 treeline in the Andes-Amazon ecotone of Bolivia. Using dendrochronological methods, we developed an annually-resolved
24 tree-ring width chronology spanning from 1850 to 2018 C.E. To our knowledge this is the longest tree-ring record (168 yrs)
25 for this species. The standardized ring-width chronology revealed a significant radial growth decline after the 1962-1963
26 growth ring (i.e. 1963-2018). From 1960-2015, smaller ring-widths were associated with drier and warmer conditions during
27 the previous wet season (~November-March), while wetter and cooler conditions led to enhanced growth in the following
28 year. *P. pepeii* radial growth decline occurred at the same time a significant increase in minimum temperature and decrease in
29 precipitation and diurnal temperatures was observed both locally and regionally across tropical South America between 1960-
30 2015. These climate trends at treeline may indicate a reduction in moisture convergence and transport to such higher elevation
31 settings in the Andes. If temperature continue to rise at current rates, one of the highest-elevation tree species on the globe, *P.*
32 *pepei*, may continue to decline leading to forest mortality and jeopardizing the survival of treeline ecosystems.

33 1 Introduction

34 South American treelines or ‘high montane’ forests refer to the upper range limit of tree growth within the Andes Mountains,
35 and are the longest, most biodiverse ecotone in the tropical latitudes (Körner, 2012; Young and León, 2006; Zelazowski et al.,
36 2023). Whether or not such tropical treelines will remain stable under global warming is a topic of great concern due to
37 increasing anthropogenic and climate-related pressures that threaten South American forest dynamics. In recent decades, land-
38 use change and warming temperatures have led to observed shifts in species composition, distribution and increased tree
39 mortality in the tropical Andes treelines (Cuesta et al., 2020; Feeley et al., 2012, 2011; Macek et al., 2009; Young and León,
40 2006). In southern Peru, ice core records indicated that recent surface warming in the central Andes has been unprecedented
41 in the past 5000 years, and land-atmosphere models suggest the net primary production of upper montane forests in South
42 America will decrease as temperatures increase (Fajardo et al., 2019; Nagy et al., 2023; Thompson et al., 2006). Despite the
43 value of these modeling and plot-based studies of tropical South American treelines, our knowledge of forest age, annual
44 growth patterns, and historical response to climate trends remains limited.

45
46 Dendrochronological studies in tropical South America are generally scarce (~8°N-24°S) but have been key in reconstructing
47 long-term tree-growth and climate variability in the Andes (Andreu-Hayles et al., 2023; Groenendijk et al., 2025; Quesada-
48 Román et al., 2022). Tree-ring studies have reported that several Andean treeline species form annual growth rings and that
49 the width of the rings (i.e. RW) are sensitive to precipitation and/or temperature variability e.g. *Escallonia myrtilloides*,
50 *Polylepis tarapacana*, *Polylepis reticulata* (Alvites et al., 2019; Argollo et al., 2004; Morales et al., 2004; Requena-Rojas et
51 al., 2021, 2020). The relationship between annual RW in treeline species and local climate variability in the tropics can vary
52 depending on the latitudinal and elevational position of the forest along the Andes (e.g. *Polylepis pepeii* in northern vs. Southern
53 Bolivia) (Jomelli et al., 2012; Roig et al., 2001). Hydroclimate conditions in South America are largely influenced by the El
54 Niño-Southern Oscillation (ENSO). In addition to inter-annual and decadal fluctuations in temperature and precipitation,
55 circulation patterns related to the timing and strength of the South American Summer monsoon, are modulated by ENSO as
56 well (Garreaud, 2009; Vera et al., 2006; Vuille et al., 2000). ENSO-related SST anomalies have contributed to extreme weather
57 events in South America such as seasonal drought, flooding, and other geohazards (Vera et al., 2006; Vuille et al., 2000). At
58 the central tropical Andes treeline, *Polylepis tarapacana* tree-ring width (RW) and oxygen isotopes have recorded these
59 extreme events and provided year-to-year to centennial records of ENSO variability.

60
61 *Polylepis* is the dominant genus for tropical Andes treelines and represent the highest elevation tree species in South America
62 (Pyre et al. 2025; Simpson 1979). The name *Polylepis* is derived from the Greek word ‘many layers’, which refers to the
63 multiple sheets of thin, compressed of bark useful for thermal insulation at high elevations (Rada et al. 2001; Rodriguez-et al.
64 2021). However, as the climate warms, frost tolerance may no longer be an advantageous trait, and some studies predict that the
65 germination and spatial distribution of *Polylepis* at the treeline will decrease as vapor pressure deficits, temperatures, and

66 overall aridity increase in the Andes (Cuyckens et al., 2016; López et al., 2022). To date, it is not known whether such projected
67 decline has already occurred in treeline *Polylepis* populations.

68

69 Here we describe a new tree-ring record of *Polylepis pepeii* from a treeline site at the Madidi National Park (MNP) in Bolivia,
70 a hotspot for biodiversity in the southwestern Andes-Amazon region in Bolivia (BB.Simpson)(Simpson, 1979). The
71 geographic range of *P. pepeii* (family Rosaceae; common name “Kenua” or “Queñoa”) spans from central Bolivia to northern
72 Peru (Simpson, 1979) between 3550-4800 m.a.s.l (Espinoza and Kessler, 2022). The wide dispersion of leaves and long fruit
73 distinguishes *P. pepeii* from other *Polylepis* species. Tree-ring studies have shown *P. pepeii* can reach significant age (>135
74 years) and the ring-width can be sensitive to both prior and current-year climate variability (Jomelli et al., 2012; Roig et al.,
75 2001). This species is also adapted to cold soil temperatures in the eastern cordillera of the Andes. Ecological studies of *P.*
76 *pepeii* sites between 3800-4300 m.a.s.l. in Peru and Bolivia recorded shallow soil temperatures ranging from 3-5°C during the
77 wet season (Kessler et al. 2014; Hoch and Körner, 2005). Like much of the treeline species across the Andes-Amazon, *P. pepeii*
78 is under threat of shifting temperature regimes and human impacts on the ecosystem.

79

80 In the small community of Keara Bolivia where this study is located, temperatures were so warm that a catastrophic glacial
81 lake outburst flood eliminated roads, livestock, and structures in November 2009 (~3800 m.a.s.l.; Hoffmann and Weggenmann,
82 2013). A recent forest-monitoring study in the MNP observed an upslope migration of mid-elevation tree species due to
83 increased temperatures and tree mortality in the Upper Andes vegetation (Farfan-Rios et al. 2025). Prior to the designation of
84 Madidi as a National Park in 1995, large swaths of economically valuable trees were harvested for timber (e.g. *Amburana*
85 *cearensis*; Macía, 2008). In recent years, illegal mining and logging activities have deteriorated forest structure and health,
86 with increasing loss of forest cover at lower-elevations (below 2000 m.a.s.l; Finer and Mamani, 2023). At treeline, *P. pepeii* is
87 at risk of endangerment due to habitat loss related to fires and land conversion for cattle ranching or religious practices
88 (Espinoza and Kessler, 2022; Kessler et al., 2014). Since the MNP is facing a rapidly changing environment due to climate
89 and human-related disturbances, high resolution tree-ring records may offer valuable insight on the past and current response
90 of this treeline to rising temperatures. At present, only two tree-ring studies have been published for the MNP: one for *Juglans*
91 *boliviana* in Oelkers et al. (2023) (14°40' S, 68°41' W; 1300 m.a.s.l.) and another by Andreu-Hayles et al. (2015), the latter
92 confirming the formation of annual rings in a *Pseudomedia rigida* cross-section (14°33'S, 68°49'W; 1000 m.a.s.l). The
93 objectives for this study are i.) to generate the first RW chronology for *P. pepeii* in the MNP and assess radial growth patterns;
94 ii.) to identify the primary climate variables (e.g. temperature or precipitation) modulating tree growth variability, and iii.) to
95 assess the impacts of extreme climate events at this treeline.

96 2 Materials and Methods

97 2.1 Climate data

98 The network of *Polylepis pepeii* for this study is located at treeline in the MNP near the small community of Keara, Bolivia
99 (14°42'S, 69°05'W; 3795-4400 m.a.s.l.). Local precipitation data and gridded temperature products were used to generate
100 monthly climate indices for the site between 1960 and 2015. This period was selected for the site climatology and climate-
101 growth analyses due to the limited availability of continuous precipitation data for this region of Bolivia. Daily precipitation
102 from the Italaque station in Bolivia (15.48°S, 69.03°W; 3500 m.a.s.l.) was gap filled with nearby station data to generate a
103 continuous monthly timeseries using the 'redprecc' package in R (Huerta et al., 2026). Nearest-neighbor interpolation was
104 constrained to precipitation stations above 3000 m.a.s.l. Raw precipitation data for Italaque (1978-2005) and nearby stations
105 (~1945-2015, non-continuous) were obtained from the DECADE dataset, which were originally sourced from the National
106 Meteorology and Hydrology Service of Bolivia (Hunziker et al., 2018)(SENAMHI <https://senamhi.gob.bo/index.php>).
107 Satellite-derived rainfall data from the Climate Hazards Infrared Precipitation with Station group V2.0 (CHIRPS, accessed
108 2025;(Funk et al., 2015)) was used for spatial precipitation analyses (see Section 2.6). Precipitation and temperature from the
109 Climatic research Unit TS 4.08 was used to determine the average climate during years of known ENSO-DJF events between
110 1901-2018 (see section 2.7; nearest grid point 14.75°S, 69.25°W;(Harris et al., 2020)). Precipitation values from the nearby
111 station were also evaluated for climate during extreme ENSO years, but the timeseries is limited to the period of 1960-2015.
112 Although CHIRPS v2.0 is limited to observations after 1981, it has a higher spatial resolution (0.05°) than Climatic research
113 Unit TS 4.08 precipitation (0.5°), which can be more effective for spatial climate-growth analyses in regions with complex
114 topography such as our site in the Andes-Amazon.

115

116 Monthly mean, minimum, maximum temperature data (Tavg, Tmin, Tmax) from the nearest CRU grid point were used for
117 local temperature-growth analyses (1960-2015; 14.75°S, 69.25°W). CRU minimum and maximum temperatures for tropical
118 south America region (15°N~24°S, 82°W~39°W) were used for spatial correlation analyses with the *Polylepis pepeii* ring-
119 width chronology (see section 2.6 for more details). In addition to the 1960-2015 climate-growth analysis period, we also used
120 CRU monthly diurnal temperature range (i.e. DTR; the difference between Tmax and Tmin) to evaluate long-term annual
121 temperature variability at the site between 1901-2015, as well as seasonally for the wet (October to April) and dry (June to
122 August) seasons.

123

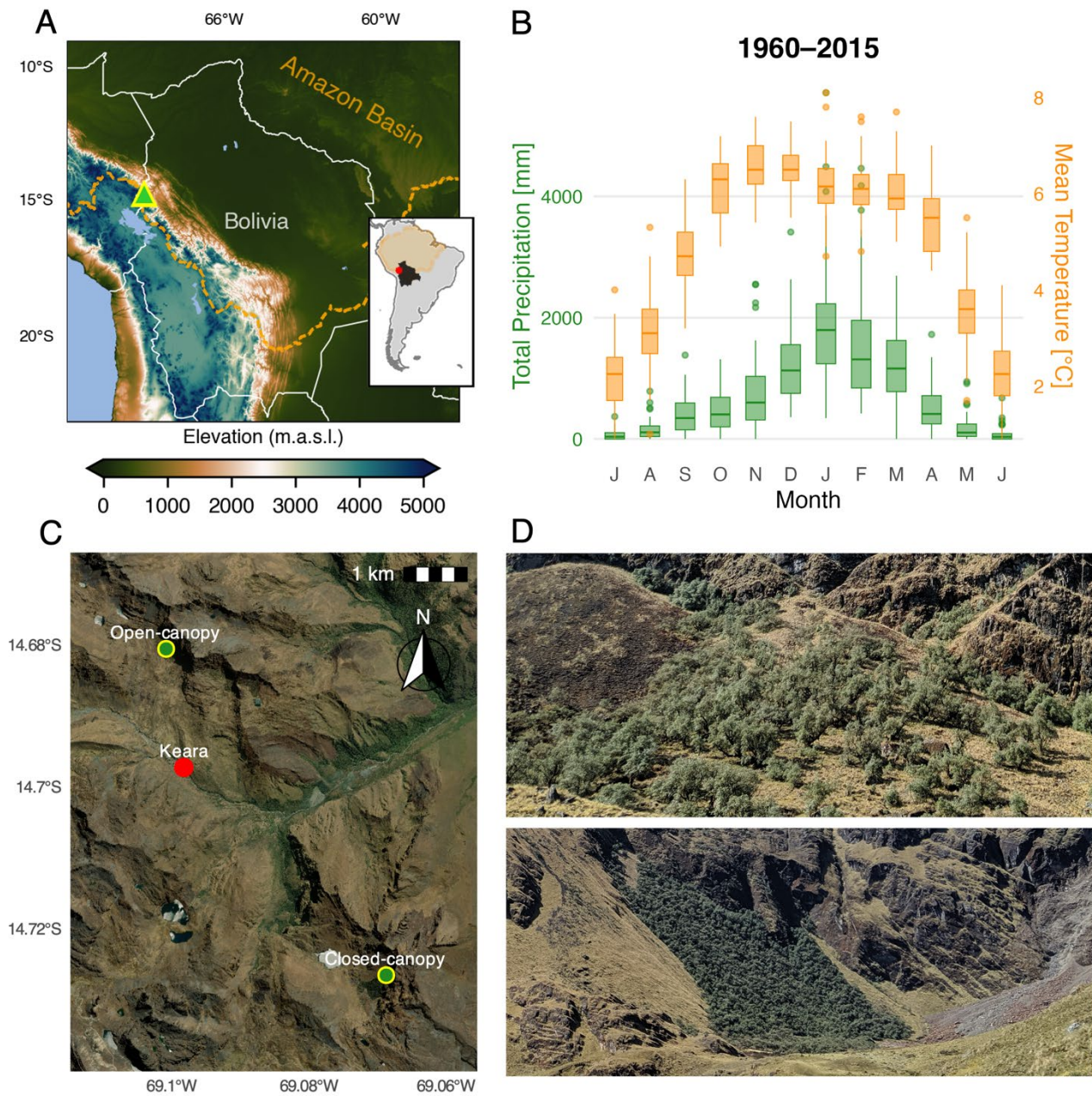
124 In situ high-resolution climate data was also obtained for this study. In August 2011, HOBO® temperature and relative
125 humidity data loggers (<https://www.onsetcomp.com/>) were installed near the *P. pepeii* trees (14°40'S, 69°06'W; 4158 m.a.s.l.)
126 and data was recorded hourly from 1 September 2011 to 2 September 2014. New HOBO sensors were installed in 2021 and
127 collected during fieldwork in 2023. Unfortunately, the 2021 system batteries failed within 9 months of the launch, and data
128 was limited to only 20 September 2021 to 23 May 2022. To compare the in-situ records of wet-season climate with the long-

129 term DTR data, daily minimum, maximum and mean temperature and relative humidity was calculated from 1 October to 30
130 April (~212 days) for the 2011-2012, 2012- 2013, 2013-2014 and 2021-2022 seasons. Daily climate distributions were
131 visualized using ridgeline density curves generated with the ‘ggridges’ package in *R* (Wilke and Wilke, 2022). Differences
132 among the four seasonal distributions for temperature and humidity were evaluated using the nonparametric Kolmogorov-
133 Smirnov (KS) statistic and implemented via the ‘stats’ package in *R* (Kolmogorov, 1933; Smirnov, 1948). To account for serial
134 dependence in daily observations, p-values were estimated using a moving-block bootstrap (e.g. contiguous 7-day blocks
135 resampled 5000 times; (Kunsch, 1989)).

136 **2.2 Site description and climatology**

137 Two populations of *Polylepis pepeii* were sampled at the MNP treeline in October 2012 and July 2019 (3795-4400 m.a.s.l.
138 14°40'-14°43'S; 69°04'-69°06'W; Fig. 1A, C, D). The 2012 campaign took place within an open-canopy south-facing forest
139 (3795-4100 m.a.s.l.), while the 2019 collection was primarily located within a closed-canopy west-facing forest called Waca-
140 cocha (named after a nearby lagoon; 4000-4400 m.a.s.l. Fig. 1D). The forests were largely monospecific, dominated by
141 fragmented patches of *P. pepeii*, and a small number of *Gynoxys compressissima* trees. This treeline is characterized as Alto-
142 Andino Yungueño vegetation (Upper Andean Yungas) with a seasonally humid climate illustrated by the monthly climatology
143 (Fig. 1B)

144
145 There is a distinct wet season from October-April and dry season from June-August at this site (Fig. 1B). 90% of annual rainfall
146 occurs during the wet season with an average of 1045 mm per month and a mean temperature range between 5-7 °C. In contrast,
147 the dry season is characterized by mean precipitation totals of 95 mm and cooler average temperatures between 2.2-3.1 °C
148 (June-August). October and November are the warmest months of year with maximum temperatures around 14.5 °C (austral
149 spring), but precipitation totals peak in January with an average of ~1900 mm (austral summer). The coldest and driest month
150 of the year is July when minimum temperatures and precipitation are as low as -8.3 °C and 65 mm, respectively (Fig. 1, A1,
151 austral winter). There is a marked seasonality in diurnal temperature range with smaller differences in minimum and maximum
152 temperatures during the peak wet -season (~14.5 °C, December-March; Fig. A1). Specifically, minimum temperatures range
153 from -2.2 °C to 0 °C and maximum temperatures from 12 °C to 14.5 °C during the wetter months of October-April. Diurnal
154 temperature differences are largest between June-August (~20°C) with minimum temperatures ranging from -7 °C to -8.3 °C
155 and maximum temperatures between 12.5 °C to 13.3 °C.



156

157 **Figure 1: (A) Location of *Polylepis* site in the Eastern Cordillera of the Andes-Amazon ecotone in Bolivia. The orange**
 158 **dotted line (and orange shading in the inset map) represent the spatial limits of the Amazon Basin. The elevation map**
 159 **was generated using the “ETOPO-1” model (<https://www.ncei.noaa.gov/products/etopo-global-relief-model>); (B)**
 160 **the monthly climatology for the region of Keara between 1960–2015. Monthly distribution of mean temperature was**
 161 **generated using the nearest temperature CRU gridpoint (14.75°S, 69.5°W) and total precipitation is reconstructed for**
 162 **the Italaque station (15.48°S, 69.03°W). (C) Aerial view of the sampling locations near the community of Keara, and**
 163 **the MNP treeline (D) photos of open-canopy (top) and closed canopy (bottom) forest patches sampled at altitudinal**
 164 **treeline in Bolivia’s MNP (~3800–4400 m.a.s.l.). The basemap in (C) was obtained through opensource ESRI images.**

165 **2.3 Wood processing and anatomical analyses**

166 Tree-ring samples from at least 30 living *P. pepei* were collected using a 2-threaded 16-inch increment borer (5 mm in
167 diameter) in October 2012 and July 2019. Two to four cores were extracted from varying directions within the stem (north,
168 south, east, and/or west radii) in an effort sample an accurate representation of radial growth. Trees were cored at breast height
169 (1.2 meters), or near the base of the largest stem for multi-stemmed trees (~30 cm). During the 2019 field campaign, tree-
170 diameter was also measured at the height core samples were sampled (1.2 meters on average). Tree-ring cross sections from 3
171 recently dead trees were sliced using a gas-powered chainsaw or a standard saw-tooth blade.

172

173 Wood samples were shipped from the National Herbarium of Bolivia in La Paz to the Lamont-Doherty Earth Observatory
174 (LDEO) in NY, USA for dendrochronological analysis. Cores and cross sections were finely sanded up to 1000 grit using an
175 orbital sander and manually polished with microfiber paper. Most samples had surficial color differences within the stem,
176 mainly reflecting transitions between the heartwood (functional xylem near the pith, darker color) and sapwood (the active
177 xylem beyond the cambium layer, lighter color).

178

179 *P. pepei* is an angiosperm with diffuse porous wood anatomy, which is particularly harder to cross-date due to less distinct
180 boundaries between the latewood of the prior year growth ring and the earlywood of the next year (Fig. A2). To aid in
181 identifying ring-boundaries in the wood, histological (micro) cuts were performed on a single cross-section following the
182 techniques described in von Arx et al. (2016) using a WSL Core microtome ([https://www.wsl.ch/en/services-
183 produkte/microtomes/](https://www.wsl.ch/en/services-produkte/microtomes/)). High-resolution images of the thin sections were captured using an Echo Revolve R-4 microscope
184 camera, with a magnification of 40X. Further information regarding the wood anatomy of this species is included in the
185 Appendix (Fig. A2).

186 **2.4 Tree-ring width chronology development**

187 Tree rings from 31 trees (51 radii) were dated visually using standard dendrochronological techniques (Stokes and Smiley,
188 1968). Each wood sample was scanned using an Epson Expression 11000XL scanner at 3200 dpi resolution. RW was measured
189 digitally using the *CooRecorder* image analysis program (Cybis Elektronik, 2010) and statistically crossdated using the
190 program COFECHA (Holmes, 1983) and dplR package in R (Bunn, 2008). To independently confirm annual periodicity of
191 the growth rings, radiocarbon dating was conducted on a cross-section sample (SP20X; Fig. A2) collected in 2019 in the
192 closed-canopy forest. Individual growth rings associated with the years 1957, 1958, 1962, 1963, 1964, 1965, 1971, and 1972
193 were sliced, extracted for cellulose, and processed for modern radiocarbon analyses. All radiocarbon measurements were
194 compared to the monthly SH $\Delta^{14}\text{C}$ radiocarbon curve (1950-2019 C.E.) from the designated atmospheric Zones 1-2 and 3 (SH
195 Zone 1-2; Hua et al. 2022). Further details on the radiocarbon analyses and earlier iterations of the Keara RW chronology can be
196 found in the Appendix (Fig. A2C).

197

198 Once final calendar dates were assigned to the tree-ring samples, the Schulman convention (Schulman, 1956), which assigns
199 each ring date to the year growth began, was applied. Individual RW time series were detrended conservatively with age-
200 dependent cubic splines using the ‘dplR’ package in *R* (initial spline stiffness of 60 yrs) (Bunn et al. 2008; Cook and Peters,
201 1981; Melvin, 2004). Standardized indices were generated by taking the ratio of the fitted and observed RW values of detrended
202 time series and combined using a robust Tukey bi-weight mean to produce a dimensionless ‘standard’ RW chronology (Cook
203 et al., 1990). For the residual chronology, autocorrelation was removed from the series using autoregressive modelling
204 determined by the Akaike Information Criterion (Akaike, 1974). A third chronology of average (raw) RW was also generated
205 to review the absolute growth of these trees at treeline. The final chronologies (raw, standard, residual) represent the RW for
206 entire *P. pepei* network (2012 and 2019 tree samples) and thus variance stabilization was applied to account for temporal
207 changes in sample depth (Frank et al., 2006). The standard RW chronology was primarily used for changepoint and RW-
208 climate correlation analyses (section 2.5), and the residual chronology was used for i.) identification of small or large outliers
209 in the chronology (top 5th and 95th percentiles) and ii.) analyses of the growth response of *P. pepei* to detrended monthly and
210 residual climate and extreme ENSO events (see section 2.6).

211

212 The subsample signal strength (SSS) calculation was used to determine how well the available tree-ring samples represent the
213 common growth signal of the *P. pepei* population (i.e. site)(Cook and Pederson, 2011). SSS quantifies the strength of the
214 shared variance through time by incorporating the number of cores per tree, the number of individual trees, and the mean
215 interseries correlation among RW series. An SSS threshold of 0.85 (or better) is commonly used in dendrochronology and
216 signifies the years when sample replication is adequate and the chronology is considered robust (see discussions *Buras et al.*,
217 (2017) and *Wigley et al.* (1984) for more details). An annual SSS Index was generated using the ‘dplR’ package in *R* (Bunn,
218 2008).

219 **2.5 Changepoint detection**

220 The period in which SSS was greater than 0.85 was used for changepoint analyses of the raw and standard RW chronologies
221 to detect abrupt shifts in radial growth before and after low-frequency is removed (i.e. detrending). The Pettit’s (1979)
222 changepoint test was applied using the ‘trend’ package in *R* (Pohlert, 2016) which identifies a single year when the median
223 tendency of the RW chronology is significantly higher or lower after the changepoint. Significance of the RW trend for the
224 post-changepoint period was evaluated using the non-parametric Mann-Kendall test on an estimated Sen’s slope (Sen, 1968).

225 **2.6 Climate-growth analyses**

226 To explore the climate sensitivity of *Polylepis pepei*, we correlated annual RW with local monthly and seasonal precipitation,
227 and minimum, maximum, and mean temperature (1960-2015; see Section 2.2). Correlations were evaluated using a 24-month
228 window spanning from prior July to current June to account for lagged climate effects on *P. pepei* RW. Long-term and inter-

229 annual growth response was assessed by correlating i.) standard RW with mean (raw) climate, and ii) residual RW with linearly
230 detrended climate respectively. 3-and 4-month seasonal climate correlations (standard and residual) were calculated to
231 determine whether growth at this site is influenced by cumulative rather than monthly temperature and precipitation conditions.

232

233 Pearson correlations (r) were estimated using stationary block-bootstrapping methods and implemented with the ‘boot’
234 package in *R* (Canty and Ripley, 2017). This technique resamples contiguous, randomly-sized blocks of data 1000 times to
235 preserve autocorrelation and quantify the uncertainty of the RW-climate relationship (see Politis & Romano 1994).
236 Significance was inferred from the two-tailed 95% confidence intervals of the median bootstrapped correlation (i.e., 95% CI
237 excludes zero). Due to the covariance between temperature and precipitation in this region, we used bootstrapped partial
238 correlations (r_p) to evaluate the independent effect of one variable on RW (e.g. temperature), while controlling for the other
239 (e.g. precipitation). Following methods of Meko et al. (2011), partial correlation coefficients for RW-temperature were
240 obtained by: i.) first performing a linear regression between RW and precipitation and ii.) calculating bootstrapped correlations
241 between temperature and the residuals from this regression. After removing the influence of precipitation, partial correlations
242 represent the distinct portion of RW variability that is explained by monthly temperature.

243

244 Spatial RW-Climate correlations across tropical south America were used to test the magnitude of the climate sensitivity of
245 the standard RW chronology to regional temperature (Tmax and Tmin) and precipitation variability. The seasonal window for
246 each climate variable was inferred from the significant monthly climate-growth relationships and the results from 3-to-4-month
247 standard and residual correlations. Precipitation correlations were conducted using the gridded CHIRPS dataset for the 1981-
248 2015 period (see section 2.2). Otherwise, local climate-growth analyses which used reconstructed station data and spatial
249 temperature (gridded CRU data) correlation analyses was between 1960-2015. Field significance was assessed using a
250 binomial test (e.g. the probability that n number of grid-cell correlations were significant by chance (raw $p < 0.05$) due to the
251 high number of comparisons; see: Livezey & Chen, 1983).

252

253 Linear trends in the seasonal climate (1960-2015) were assessed using the Sen’s slope estimator (section 2.4), and significance
254 was evaluated using Mann Kendall tests. Slopes were reported as the average change in climate in units per decade. Annual
255 and seasonal (October-April; June-August) trends of diurnal temperature anomalies (DTR) were also evaluated for the same
256 period. Additionally, monthly anomalies of minimum and maximum temperature were calculated relative to the 1901-2015
257 CRU baseline (full temporal extent of CRU data) to illustrate long-term temperature variability during the wet and dry season.

258 **2.7 Superposed Epoch Analysis**

259 ENSO varies between warmer (El Niño) and cooler (La Niña) SST phases (Ropelewski and Halpert, 1987) in the Pacific
260 Ocean, and both extremes substantially impact precipitation and temperature conditions over tropical South America. To
261 investigate the effects of extreme ENSO events on tree-growth, Superposed Epoch Analysis (SEA) was performed on the

262 residual RW timeseries using the method originally described by Haurwitz and Brier (1981) and modified by Rao et al. (2019).
 263 SEA is widely used to statistically determine whether the effects of episodic events (*e.g.* extreme climate events) on a response
 264 variable (in this case RW) are statistically significant or due to random noise. The Rao method uses 1000 random-sample
 265 double bootstrapping to quantify the RW response at the time of the event (lag= 0) and several years after (in this case four
 266 years).

267
 268 We analyzed twenty-six years of RW based on the top-ranked December-February (DJF) El Niño and La Niña events ($n=13$
 269 each) listed by the National Oceanic and Atmospheric Administration’s Physical Science Laboratory (NOAA-PSL:
 270 <https://psl.noaa.gov/enso/>). These ranked DJF years are determined by NOAA-PSL with the multivariate ENSO indices (MEI;
 271 1871~2024). MEI reflects the principal components, or dominant modes, of the entire tropical Pacific ENSO domain (30°N-
 272 30°S, 100E°-70°W) rather than any one region (*e.g.* Niño 3.4) and integrates observations of sea level pressure (SLP), SSTs,
 273 meridional (north-south) wind, and outgoing longwave radiation (see Wolter and Timlin, 2011). Extreme years are defined by
 274 Pacific SST anomalies during DJF, coincidentally when ENSO is phase-locked with the peak monsoon season (Rasmusson and
 275 Carpenter, 1982). The list of DJF- ENSO-events obtained from NOAA-PSL are included in Appendix Table A1.

276 3 Results

277 3.1 Growth decline in a *P. pepei* tree-ring chronology

278 The *P. pepei* RW chronology covers the period 1850-2018 and consists of 51 tree-ring samples (31 individual trees) from open
 279 and closed-canopy forests near the MNP treeline (Table 1; Fig. 2). Radiocarbon and standard dendrochronological methods
 280 confirmed these trees form annual rings and share a common growth signal in this region (Figs. 2A, C, D). Site metadata and
 281 RW chronology statistics for the *P. pepei* network are summarized in Table 1.

282

283 **Table 1. Summary of *P. pepei* tree-ring sample location, age, sample size, and mean correlation among RW timeseries.**
 284 **The RW chronologies represent the entire collection of cross-dated *P. pepei* samples in Keara obtained in both 2012**
 285 **and 2019.**

Site	Location (elevation)	<i>n</i> trees (<i>n</i> samples)	Mean age [yrs]	Timespan	Mean Correlation [\bar{r}]
<i>Open-canopy forest</i> (south-facing)	14°40'S 69°06'W (3795-4100 m.a.s.l.)	16 living 2 dead (33)	89	1850-2018	0.53
<i>Closed-canopy forest</i> (west-facing)	14°43'S 69°04'W (4000-4400 m.a.s.l.)	12 living 1 dead tree (18)	101	1871-2018	0.44
Full network (mean Raw, standard, residual chronologies)	-	31 (51)	93	1850-2018	0.50

286

287 Due to extreme suppression in radial growth (Fig. 2B), only one or two cores from the living trees could be measured and
288 included in the final RW chronologies. Despite the complex anatomy of *P. pepei*, the cross-dated samples shared a coherency
289 in the RW patterns with a mean inter-series correlation of $\bar{r} = 0.50$ for the 1850-2018 period. The oldest living tree sampled
290 was 168 years of age from the open-canopy forest. The average age of the trees was 93 years and the average growth rate
291 between 1850-2018 was $\sim 1.0 \text{ mm yr}^{-1}$. DBH measurements in 2019 confirmed these trees were slow-growing with stem
292 diameters ranging from 10 cm to 54 cm (mean DBH 30 cm). The sub-sample signal strength indicated the RW chronology is
293 particularly robust between 1900-2018 when sample size exceeds 17 (SSS >0.85 Fig. 2C).

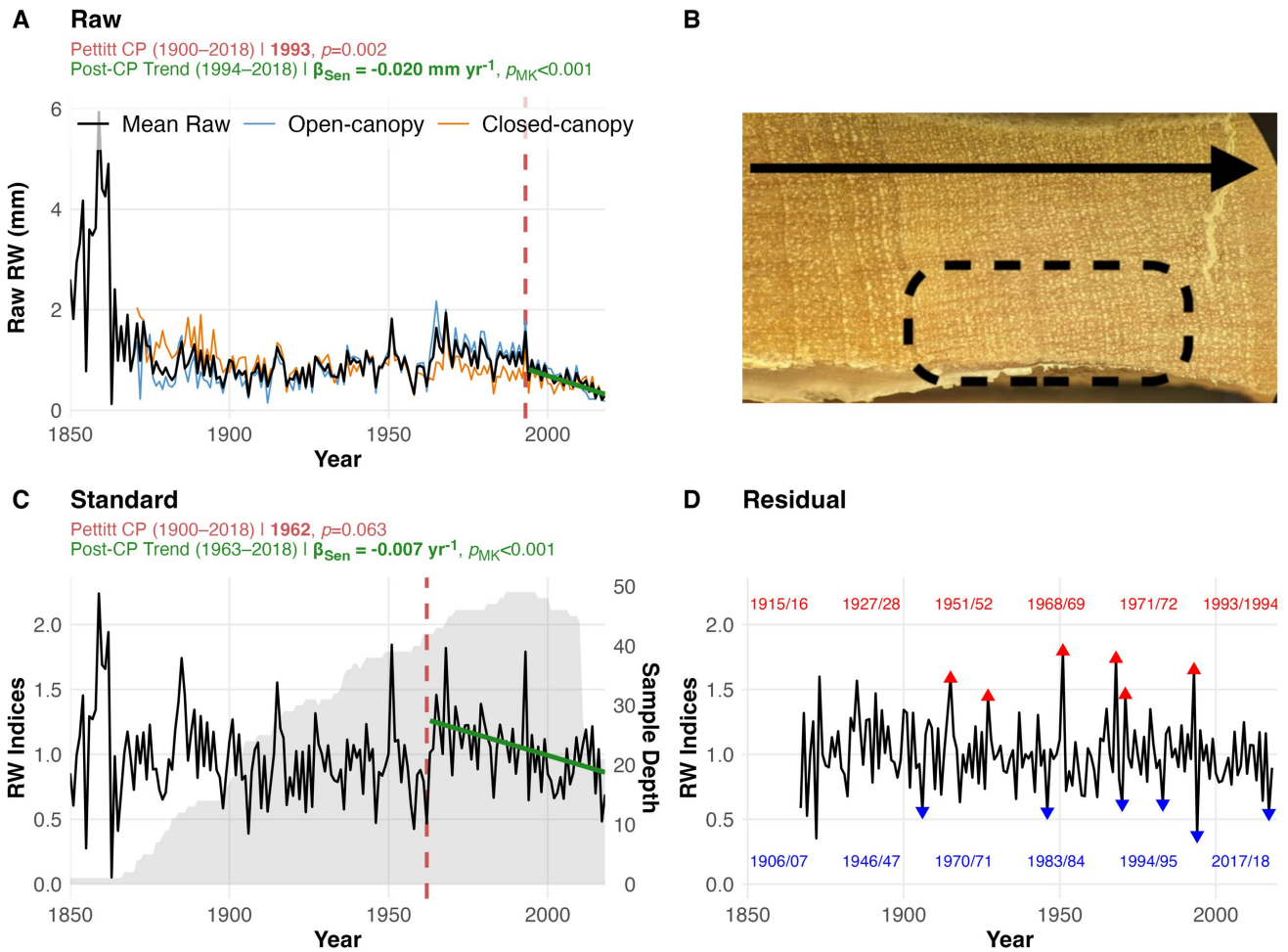
294

295 *P. pepei* RW at this treeline has been declining steadily since the 1960s (Fig 2.A, C). A changepoint in the raw RW chronology
296 was detected in 1993 ($p=0.0019$), with a significant decline of 0.02 mm yr^{-1} after 1994. This trend in absolute growth is evident
297 in both the open-canopy and closed-canopy forests at the site (Fig. 2A). Although a non-significant changepoint in the standard
298 RW chronology was detected in 1962 ($p=0.062$), two-tailed Mann-Kendall tests determined the negative slope between 1963-
299 2018 was significant ($-0.007 \text{ units yr}^{-1}$, $p<0.001$; Fig. 2C).

300

301

302



303

304 **Figure 2.** The (A) Raw (C) Standard and (D) Residual RW chronologies of *P. pepei* in Keara. The chronologies (1850-
305 2018) are plotted using the Schulman convention (*i.e.*, anchored on the year of initial ring formation; see section 2.3).
306 (B) An image depicts a core sample where several rings are suppressed within a 4 mm distance (dashed circle). The
307 black arrow indicates the direction of radial growth for this core (from left to right). (C) The standard RW is plotted
308 with the mean sample-depth of the full network through time (grey-shading). Blue and red triangles on the residual
309 timeseries (D) signify the years within the top 5th and 95th percentiles of RW since 1900 (SSS > 0.85). Since the tree rings
310 are estimated to form during the wet season (two calendar years ~October–April), both years are labeled in the colored
311 text and ordered chronologically within the plot (D). Vertical red lines signify the significant change points detected in
312 the raw and standardized RW chronologies (1900–2018; SSS > 0.85), while the green solid lines represent the trends
313 after the change point estimate via Sen’s slope. Change point and trends statistics for the raw and standard RW
314 chronologies are included in the subtitles for A and C. There was a significant decline in raw (radial) RW after the
315 1993/94 growth-year (*i.e.* 1994–2018) and between 1963–2018 in the standard RW chronology.

316

317 3.2 Monthly and seasonal climate-growth relationships at Keara's treeline

318 The climate sensitivity of *P. pepei* RW is illustrated by bootstrapped correlations with monthly mean precipitation and
319 temperature timeseries (1960-2015; Fig. 3, A3). *P. pepei* RW benefits from higher rainfall in the months of December-April
320 (Pre lag=1, $r=0.39$ for February; Fig.3A). This forest had a more significant RW response to cumulative rather than monthly
321 precipitation variability, with the highest correlations observed for the 4-month December-March season (DJFM lag=1, $r=$
322 0.42 , $p=0.008$; Figs. A3A, A4A-B). DJFM precipitation significantly decreased at this site at a rate of ~ 6 mm decade⁻¹ after
323 1960 (Fig. 4A), but residual RW-precipitation correlations remained significant after climate trends were removed (Fig. A4B).

324

325 In general, wet season precipitation is significantly and negatively correlated with temperature at this treeline (Oct-April $r=-$
326 0.31 , $p=0.02$). However, after controlling for precipitation effects on RW variability, partial correlations revealed a robust and
327 independent relationship between RW and current-year monthly temperature (*i.e.* the monthly correlations persist after the
328 covariance with precipitation is removed; Lag=0 Fig. 3B-D). *P. pepei* RW correlates positively with current-year mean and
329 maximum temperatures particularly for the months of April (Tavg $r=0.37$, Tmax= 0.38 ; Fig. 3B, D). The strongest positive
330 correlations with temperature were for the 4-month February-May season, emphasizing the importance of late summer
331 temperature variability for tree-growth at this site (FMAM Tmax, lag=0, $r=0.47$, Fig. A5H). There was weak evidence of a
332 decreasing trend FMAM maximum temperatures for the analyses period $-0.06^{\circ}\text{C decade}^{-1}$, $p=0.17$; Fig. 4D).

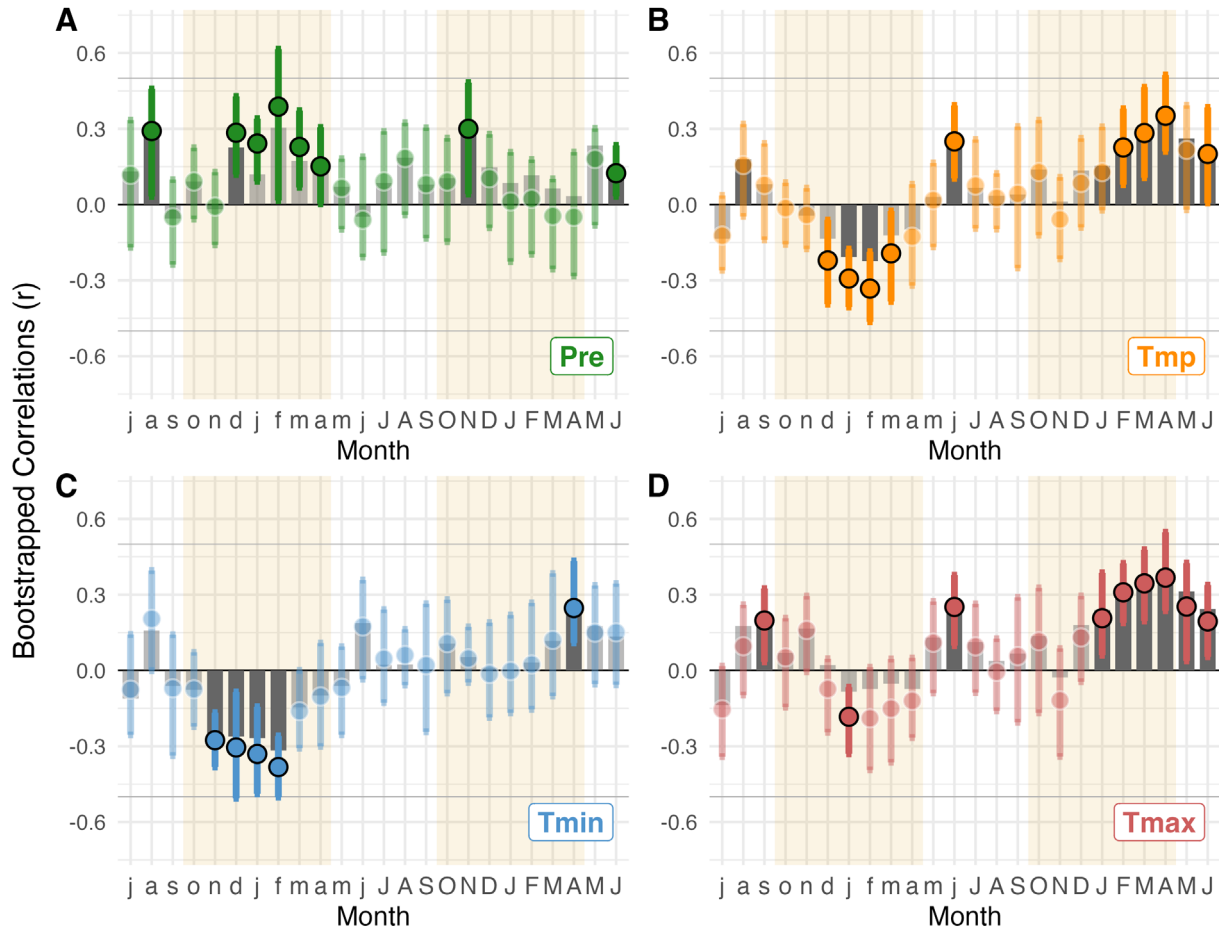
333

334 Monthly residual correlations showed *Polylepis pepei* RW was primarily sensitive to interannual temperature variability at this
335 site (Figs. 3B-C, A3B-C, Tavg, Tmax, Tmin). However, prior-year correlations with monthly Tmax and Tavg decreased after
336 accounting for lagged precipitation effects on RW (Fig. 3B and D lagged partial correlations). Lagged minimum temperature
337 variability, however, had a distinct and significant impact on RW (Tmin lag=1, Figs. 3C, A3C). Negative correlations were
338 found between RW and prior-year wet-season minimum temperatures (NDJF lag=1; $r=-0.40$; Fig. A4F). Although there were
339 no significant trends in FMAM mean and max temperature for the 1960-2015 period ($p > 0.16$; Fig. 4B, C), NDJF minimum
340 temperatures significantly increased at a rate of $0.15^{\circ}\text{C decade}^{-1}$ ($p=0.02$, Fig. 4C). In summary, *Polylepis pepei* RW was
341 limited by minimum and maximum temperatures for distinct seasons (e.g. positive relationship with FMAM Tmax lag=0 vs.
342 negative relationship with NDJF Tmin lag=1).

343

344 Overall *Polylepis pepei* RW is larger under wetter and cooler conditions. However, this site experienced significant warming
345 and drying trends between 1960-2015 at the same time a growth decline in the treeline *P. pepei* was observed (Figs. 2C, 4A,
346 4C). Residual climate-growth correlations were significant for the same seasons as the standard correlations, despite the linear
347 trends identified in both mean climate and RW (Fig. 2, 4). These results emphasize that both long-term and inter-annual climate
348 variability had significant impacts on *P. pepei* RW at this treeline in the MNP. Further, RW was primarily limited by prior-

349 year wet-season conditions (e.g. ~Nov-March Pre and Tmin lag=1) but was also significantly and positively related to current-
 350 year temperature variability (FMAM lag=0).

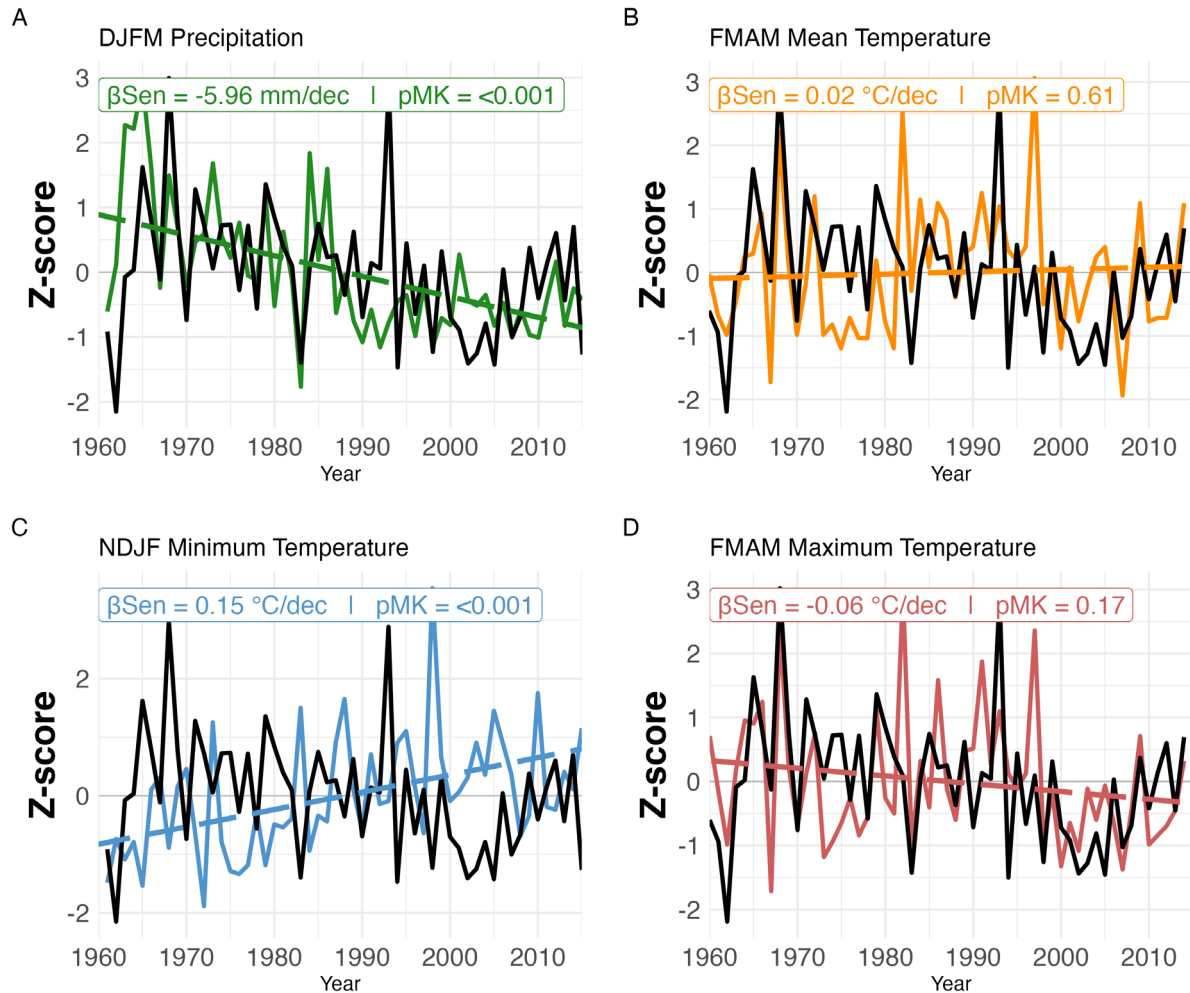


351

352 **Figure 3: Standard bootstrapped correlations between *P. pepei* RW and average monthly climate from 1960-2015.**
 353 **Monthly temperature data was from the nearest CRU gridpoint for our site: 14.75°S, 69.25°W, while monthly**
 354 **precipitation was obtained from local station data. The x-axis covers the 24-month period between July of the prior-**
 355 **year (lag=1, lowercase letters) and June of the current year (lag=0, uppercase letters). Tan shading indicates the wet-**
 356 **season for this region of the MNP (~October-April). (A) Monthly RW-precipitation correlations (green, Pre). (B-D)**
 357 **RW correlations with monthly mean (orange, Tavg), minimum (blue, Tmin), and maximum temperature (red, Tmax).**
 358 **Significant correlations are represented by solid-colored circles (median Pearson r) or solid grey bars (partial**
 359 **correlations, r_p) and inferred from 95% confidence intervals.**

360

Seasonal climate trends near Keara



361

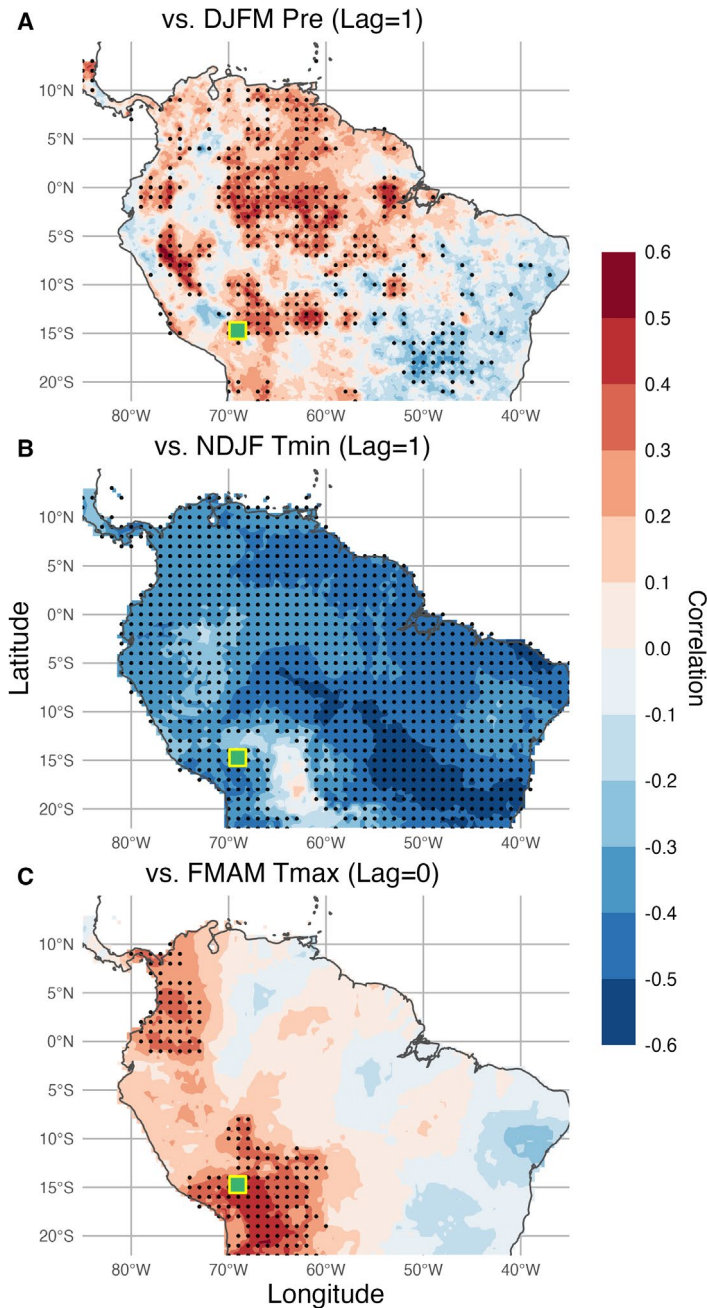
362 **Figure 4. Z-scored *P. pepei* RW (black lines) and seasonal climate variables (colored solid lines) between 1960-2015.**
 363 **The selected seasons were based on the highest and most significant correlations between local climate and RW for this**
 364 **period (standard and residual, $p < 0.05$). Lagged DJFM Precipitation is in green (A, lag=1), FMAM mean temperature**
 365 **is orange (B, lag=0), lagged NDJF minimum temperature is blue (C, lag=1) and FMAM maximum temperature is in**
 366 **red (D, lag=0). Averaged seasonal trends are reported in units of climate per decade (β , colored dashed lines). Mann-**
 367 **Kendall tests were used to estimate the two-tailed significance of the linear trend (p_{MK}). Precipitation data is**
 368 **reconstructed from nearby station records and temperature data is sourced from the nearest CRU TS 4.08 grid point.**

369

370

371 **3.3. *Polylepis pepeii* RW and climate variability across tropical South America**

372 Figure 5 highlights the regional extent of the seasonal climate signal recorded in *P. pepeii* RW at the MNP treeline. Gridded
373 DJFM precipitation (lag=1), NDJF minimum temperature (lag=1), and FMAM maximum temperature fields for tropical South
374 America were used for spatial correlation analyses and were limited by the temporal extent of available data (Fig. 5A 1981-
375 2015, Fig. 5B-C 1960-2015). The RW-precipitation correlations reflect the heterogeneity of precipitation variability along the
376 Andes, whereas the temperature fields are more uniform (Fig.5A vs. Fig. 5B, C). *P. pepeii* RW is positively correlated to prior-
377 year DJFM rainfall in most of tropical South America for the 1981-2015 period (Fig. 5A). The strongest precipitation signal
378 is observed along the eastern flanks of the Peruvian Andes and the northern Amazon Basin in Brazil ($r = 0.45$ to 0.50). There
379 were significant correlations with wet-season minimum temperature variability for most of tropical South America, especially
380 near southeastern portions of Brazil (NDJF T_{min} lag=1, 1960-2015; Fig. 5B). RW is also positively correlated to late austral
381 summer maximum temperature variability in the northern and southern portions of the tropical Andes (FMAM T_{max} lag=0,
382 1960-2015, Fig. 5C). The highest correlations between RW and maximum temperature were centralized both locally to the
383 site and near the Bolivian Altiplano ($r > 0.50$). In summary, the relationship between *P. pepeii* RW and large-scale seasonal
384 climate variability was significant for the same seasons identified in the local climate-growth analyses (1960-2015, Figs. 3,
385 A4, A5).
386



387

388 **Figure 5. Spatial correlations between *P. pepei* RW and (A) DJFM precipitation (lag=1), (B) NDJF minimum**
 389 **temperatures (lag=1), and (C) FMAM maximum temperatures (lag=0) in tropical South America. Black dots represent**
 390 **the significant grid points ($p < 0.05$). Gridded precipitation analysis was limited to 1981-2015 due to the availability of**
 391 **CHIRPS data (A) while RW-temperature correlations covered the 1960-2015 period (B-C, CRU data). There were**
 392 **more significant cells than expected by chance for all spatiotemporal relationships (binomial field test $\alpha = 0.05$, $p < 0.001$).**

393 **3.4. Long-term changes in diurnal climate conditions at the Keara treeline**

394 In agreement with the decreased DTR trend observed since 1960, in situ daily temperature loggers independently confirmed
395 that minimum temperatures for the October-April season have increased at this site (Fig. A6 A-C). Minimum temperatures for
396 the October-April season in 2011-2014 ranged from 1.9-2.3 °C and increased to 3°C in 2021-22, which is higher than the 0°C
397 average between 1960-2015 (Section 2.2, T_{min} Fig. A1). The distribution of daily wet season minimum temperatures in 2021-
398 2022 was significantly higher than the 2011-2014 seasons overall (bootstrap KS test $p < 0.001$; Fig A6).

399
400 Interestingly, average October-April maximum temperatures recorded by the data loggers ranged 7.2°C-8.2°, which is
401 substantially lower than the 1960-2015 average (Section 2.2, T_{max}~13°C; Fig. A1). Daily maximum temperature for the 2021-
402 2022 season was significantly lower than the 2012-13 and 2013-14 seasons ($p < 0.01$), while no significant difference was
403 detected relative to the 2011-2012 season ($p = 0.37$).

404
405 The daily data loggers also recorded a significant reduction in relative humidity within the 2011-2022 period (October-April;
406 Fig A6D-F). Relative humidity declined from an average of 98% in 2011-2014 to 90% in 2021-2022, while minimum daily
407 humidity decreased from ~94% to 80% (Fig. A6D, F). Although 2011-12 corresponded to a DJF-La Niña year (Table A1), the
408 distribution of daily relative humidity values during that season was comparable to the 2012-13 and 2013-14 years ($p > 0.05$),
409 and all 3 seasons were more humid than the 2021-22 year (Fig. A6. $p < 0.001$). In fact, almost one third of daily relative humidity
410 values for 2021-2022 (63 days) were below 90% for October-April, while less than 25 days were recorded for the same
411 threshold in the 2011-2014 seasons.

412
413 In summary, the long-term warming and drying trends observed in precipitation and temperature between 1960-2015 (Figs. 4,
414 A5) are consistent with in situ measurements of higher temperature and lower relative humidity for 4 distinct October-April
415 seasons between 2011-2022 (Fig. A6).

416

417 **3.5 Growth response of treeline *P. pepei* to extreme climate events**

418 Table A1 lists the years of known hydroclimate anomalies in tropical South America connected to El Niño (warmer SST) or
419 La Niña (cooler SST) conditions in the Pacific Ocean. Regardless of the spatial temporal resolution, climate datasets agree that
420 the El Niño-DJF events were linked to drier and warmer conditions at this site, while extreme La Niña-DJF years were wetter
421 and cooler (Fig. 6A, B). SEA of the residual RW response depicts temporal growth anomalies for top 26 ENSO DJF events:
422 13 El Niño and 13 La Niña (Figs. 6CD). There was a one-year delayed and negative RW response to El Niño events, i.e., one
423 year after an event, *P. pepei* showed a significant decrease in radial growth ($\alpha = 0.1$). There was a significant negative response
424 two and three years after an extreme DJF-La Niña.

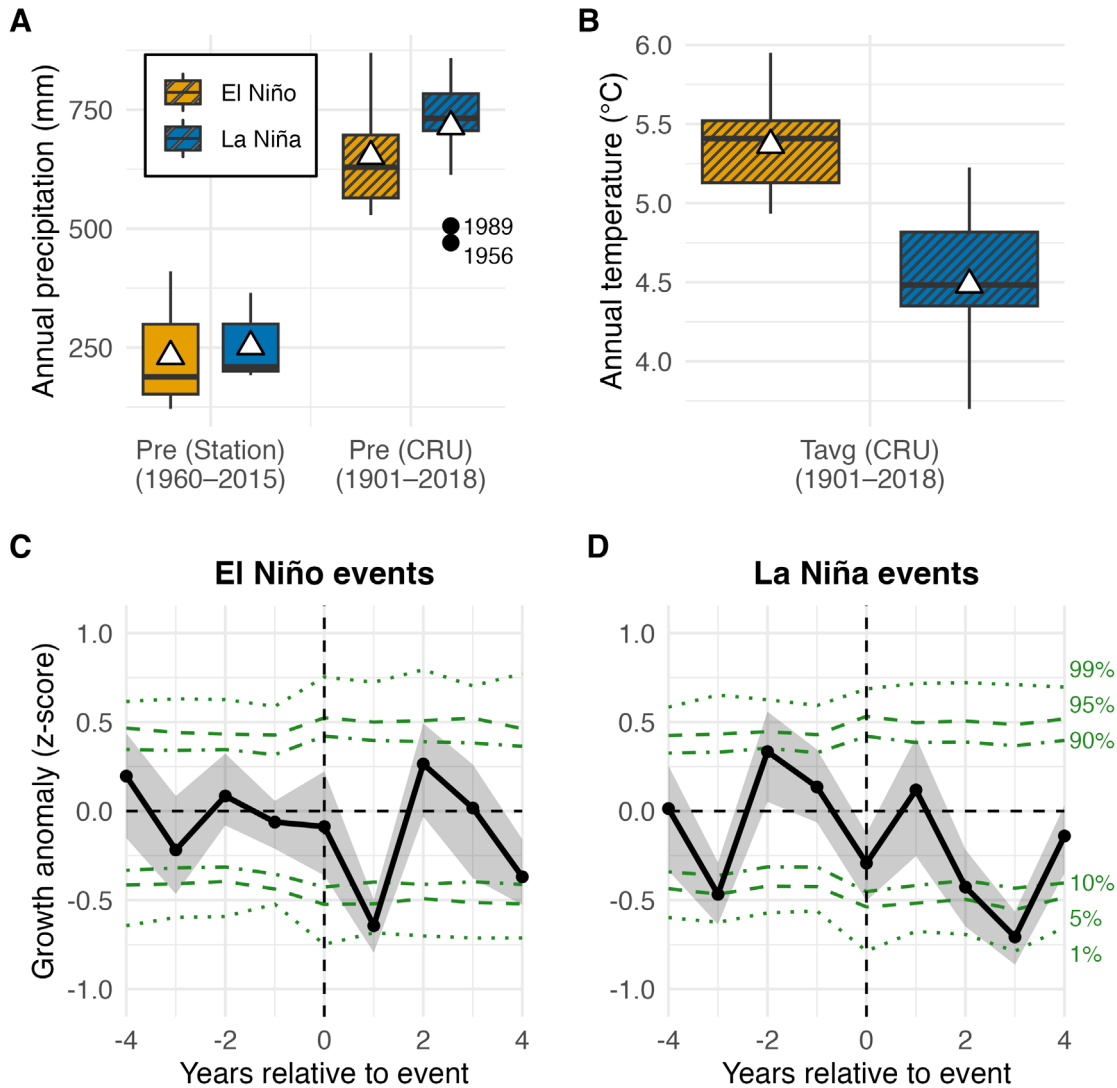
425

426 Some RW outliers identified in the residual RW chronology coincide with years of known ENSO events. (Fig. 2D) The largest
427 tree-ring observed in *P. pepei* occurred at the same time than an extreme DJF-La Niña event in 1950-51 while the smallest
428 growth ring in the chronology was formed after a DJF-El Niño event in 1905 (Table 1; Fig 2D). Overall, the inter-annual *P.*
429 *pepei* RW at the Andes-Amazon treeline has a lagged significant response to extreme DJF-ENSO conditions.

430 To visualize the significant increase of minimum temperatures near Keara in a long-term context, annual and seasonal diurnal
431 temperature range anomalies were calculated for the full CRU TS 4.08 calibration period 1901-2015 (Fig A5; DTR, Tmin,
432 Tmax). There was a significant decrease in DTR and an increase in minimum temperature anomalies since the mid-20th century
433 at this site. The largest decline in DTR was observed for the October-April season at a rate of $-0.197^{\circ}\text{C decade}^{-1}$, though annual
434 and July-August DTR declined at a similar rate (-0.191°C and $-0.194^{\circ}\text{C decade}^{-1}$ respectively, 1960-2015; Fig. A5).

435

436



437

438

439 **Figure 6: Boxplots of annual (A) precipitation and (B) temperature for the 26 years of extreme DJF-ENSO events,**
 440 **using both station and CRU data. Outlier years in the CRU precipitation data are represented as black dots (A; 1956,**
 441 **1989). Annual mean climate during El Niño-DJF is represented as orange colors, while La Niña is shown in blue ($n =$**
 442 **13 years per event). Superposed epoch analysis of the residual RW response 4 years before and 4 years after the DJF-**
 443 **ENSO events (C, D; black line). The uncertainty of the growth response is depicted as grey shading. Horizontal green-**
 444 **lines represent the two-tailed significance thresholds: (10-90%, 5-95% 1-99%) derived from stationary bootstrapping.**

445

446 4 Discussion

447 4.1. Climate sensitivity and Radial growth decline of a tropical treeline site in Bolivia

448 Here we have generated the first *P. pepei* tree-ring chronology from the MNP treeline in Bolivia and the longest annual growth
449 record for this species in South America spanning from 1850 to 2018 CE (3795-4400 m a.s.l; 14.75°S). These results provide
450 new information of tree growth and climate dynamics in an understudied biodiversity hotspot in the southwestern Andes-
451 Amazon corridor. We found that *P. pepei* RW is limited by prior-year minimum temperature (negatively) and precipitation
452 (positively) variability during the wet season, leading to larger RW in the subsequent growth year when it was colder and
453 wetter (Figs. 3 and 4). There was also a positive relationship between current-year maximum temperature and RW variability
454 between 1960-2015 at this treeline. The lagged RW-precipitation signal is consistent with one of the two *P. pepei* tree-ring
455 studies (lag=1; 1969-2004; 16°12S, 68°79W; 4130 m.a.s.l; Jomelli et al. 2012) and several *Polylepis tarapacana* investigations
456 that reported prior-year water availability as a useful predictor of RW (Argollo et al. 2004, Morales et al. 2004, Christie et al.
457 2009, Soliz et al. 2009, Crispín-DelaCruz et al. 2022, Rodriguez-Caton. et a. 2021). The second study of *P. pepei* found that
458 RW was positively correlated to current-year temperature variability (lag=0; 1941-1983; 17°S, 65°39W; 4100 m.a.s.l Roig et
459 al. 2001). However, the dominant and negative relationship between RW and lagged minimum temperature variability has yet
460 to be explored for treeline *P. pepei*.

461
462 The pronounced decline of radial growth in *P. pepei* since 1960 coincided with warmer and drier austral summers at this
463 treeline (~November-March, 1960-2015; Figs. 2C & 4AC). There was also a significant step change and subsequent decline
464 in raw RW after the 1993 growth-ring (Fig. 2A). It is possible that this area was affected by an 8.2 magnitude earthquake ~150
465 km east of the site was recorded in June 1994 (Myers et al. 1995). Although this region of Bolivia is sparsely populated,
466 landslides were reported in southern Peru and Chile, and thus it is possible that slope failures in 1994 occurred near this treeline
467 in Keara Bolivia (Blodgett et al. 1998). This decline also occurred 1 year before the broader study region of Madidi was
468 designated a National Park in September 1995 due to its extremely high levels of biodiversity and endemism for the Southern
469 Amazon. The results herein verify the sensitivity of *P. pepei* tree-ring width at this site to both local climate and environmental
470 disturbances. An altitudinal advance of this treeline is unlikely due to local geomorphologic constraints (Fig. 1D; Macias-
471 Fauria and Johnson, 2013). Further, species-specific thermal niches that are required for seedling establishment and recruitment
472 may be at risk due to the observed increasing temperature trends in recent decades (Kessler et al., 2014; Körner and Hoch,
473 2023). Although the spatial extent of this forest under increased warming and environmental change is uncertain, this study
474 has helped clarify the nature of past and current growth variability of *P. pepei* in this region.

475
476 Human activity in Andes-Amazon forests should be considered when evaluating the growth patterns at tropical treelines. Forest
477 fragmentation was observed in the lower elevation open-canopy forest (3795-4100 m.a.s.l.) and tree-ring samples showed
478 evidence of fire scars in the 1940s. This subpopulation is likely more threatened by small-scale cattle ranching in Keara than

479 the closed canopy forest located at higher elevations (4000-4400 m.a.s.l.; Fig. 1C, D). Herein, efforts were made to minimize
480 potential impacts of land use and disturbance in field sampling, but ecosystem disturbances in certain regions of the MNP were
481 nevertheless observed (mostly in areas below 2000 m.a.s.l.). Despite these observations, correlations between *P. pepei* RW
482 and regional temperature and precipitation variability were robust and may signify a response to larger-scale hydroclimate
483 patterns seen in tropical South America (Fig. 5).

484
485 The MNP treeline revealed an imprint of extreme ENSO-related drought events in *P. pepei* growth rings. Hot and dry
486 conditions related DJF-El Niño years had significant and negative impacts on radial growth (Fig. 6C). Although similar RW
487 responses to large-scale drought DJF ENSO events were selected for analyses because they occur during mature phase of the
488 summer monsoon (70% of annual rainfall), but the impacts of annual or broader seasonal ENSO extremes may be more
489 informative due to the sensitivity of RW to cumulative climate conditions (e.g. seasonal correlations Fig. A4). Overall local
490 and regional climate-growth analyses suggest this treeline is primarily limited by temperature-driven moisture stress during
491 the peak wet season (~November-March Figs 3,4, A3, A4). In the following sections we discuss both site-level and large-
492 scale hydroclimate conditions that may have contributed to the growth decline of treeline *P. pepei* since the 1960s.

493 494 **4.2. Local temperature and humidity changes at the *P. pepei* treeline in Bolivia**

495 Increases in minimum temperature may be reducing moisture availability in high-elevation Andean tropical forests. The mean
496 diurnal temperature range near Keara has significantly and negatively declined since 1960, signifying that the rate of minimum
497 temperature increase has surpassed that of maximum temperature (Fig. A5. CRU TS 4.08 ;1901-2015). Daily climate loggers
498 within the *P. pepei* site indicated the 2021-2022 wet season was significantly warmer and drier than in 2011-2014 ($p < 0.0001$
499 October-April). This decrease in relative humidity alongside an increase in minimum temperatures suggests that the capacity
500 of air to hold moisture has outpaced the actual moisture content, making the air drier despite higher temperatures. Although
501 the data logger records cover a short time window, they provide in situ, high-resolution data over a 10-year period in Keara,
502 and thus they are useful to confirm: i.) the warming trends observed in the long-term climate data (Fig. 4, A5), and ii.) that
503 the increase in minimum temperatures coincided with a decrease in humidity at this site (Fig. A6) .

504
505 *P. pepei* RW was significantly and negatively impacted by lower wet-season precipitation and higher minimum temperatures
506 even after linear trends in climate were removed (Fig. A4; residual correlations). Minimum temperatures influence convection,
507 as rainfall in the Andes-Amazon ecotone largely occurs in the afternoon/night when radiative cooling drives cold air downslope
508 and converges with rising warm moist air from the tropical lowlands (Garreaud, 2009; Junquas et al., 2018; Romatschke and
509 Houze, 2010; Rosales et al., 2022). Known as ‘orographic precipitation’, this process is key for rainfall distribution across the
510 Andean foothills (Arias et al., 2021; Chavez and Takahashi, 2017). According to the hourly in situ data loggers in Keara, daily
511 RH peaked at 3:00 P.M. on average, likely reflecting afternoon cloud formation at this site. In the region of the MNP,
512 maximum precipitation totals occur between 1000-1300 m.a.s.l., with a sharp decrease in moisture transport above 3000

513 m.a.s.l. (Chavez and Takahashi, 2017). If minimum temperatures increase, the radiative cooling effect weakens, which may
514 result in warm, moist air converging at lower elevations (e.g. below our *P. pepei* site)(Romatschke and Houze, 2010). In terms
515 of tree-growth in this area, one interesting observation is the declining *P. pepei* RW trends observed at the MNP treeline (above
516 4000 m.a.s.l.) diverge from the increasing RW trends observed in lower-elevation humid forests in the MNP (e.g. *Juglans*
517 *boliviana* ~1300 m.a.s.l) (Oelkers et al. 2023).

518
519 One possible explanation for this pattern is that warmer minimum temperatures alter the local balance between temperature,
520 humidity, and upslope moisture transport in the Andes–Amazon ecotone. Because precipitation at these elevations depends
521 strongly on orographic processes and on the interaction between moist lowland air and cooler mountain conditions, weaker
522 nighttime cooling could reduce the efficiency of moisture convergence at the elevation of the treeline. We cannot demonstrate
523 that mechanism directly here, but the observed increase in minimum temperature, decline in DTR, and reduction in relative
524 humidity in Keara are all consistent with a shift toward locally dry conditions that may limit *P. pepei* growth (Figs 4, A5, A6).
525 In this sense, the results point to moisture stress amplified by warming, rather than temperature alone, as the most plausible
526 explanation for the long-term decline in ring width at the MNP treeline.–Future research needs to be conducted on elevational
527 moisture transport in the Andes-Amazon and these diverging tree growth trends observed within the MNP.

528
529 Besides inhibiting moisture convergence and transport from lower elevations, warmer minimum temperatures could also affect
530 the tree water-balance in Keara. A global analysis of tropical tree longevity found that tree mortality is increasing in all tropical
531 biomes due to heat-related water stress and increased evaporative demand at the leaf level (Locosselli et al., 2020). From an
532 ecophysiological perspective, respiration rates increase with warmer temperatures. Excessive warming at night (i.e. increased
533 T_{min}) without the process of photosynthesis may increase the amount of tree-level carbon (respiration) and soil-water content
534 released to the atmosphere (Körner et al. 2012). Temperature and precipitation near the *P. pepei* site are inversely correlated
535 during peak wet season (~November-March, $p < 0.05$). Thus, if there is less cloud cover (and precipitation), there is higher solar
536 irradiance and temperatures, which may limit the photosynthetic capacity of trees especially in tropical moist treelines (García-
537 Núñez et al., 2004; Hoch and Körner, 2005; Jaramillo, 2015). Further research must be conducted to determine the
538 physiological response of *P. pepei* at this treeline to diurnal and seasonal changes in orography and soil-water availability.

539
540 The positive relationship between *P. pepei* and February-May mean and max temperatures may demonstrate ecophysiological
541 link between tree-ring size (i.e. xylogenesis) and cooling observed at the end of the wet season (lag=0; FMAM T_{max} Figs.
542 3BD, A3BD). Although it is possible the mechanisms controlling primary and secondary growth may not occur at the same
543 time (i.e. photosynthesis vs. wood formation), the RW-climate correlations indicate growth rings are smaller when there are
544 cooler conditions in late austral summer (or vice versa; Figs. 3, 4). It is interesting to note that an extreme cold period observed
545 near Keara during the late 1950s corresponds with a growth suppression observed in raw, standard and residual *P. pepei* RW
546 chronologies, emphasizing the coupled relationship between RW and mean and maximum temperatures during this period

547 (Figs. A5B, C and 2A, C). The exact growing season for *P. pepeii* in Keara estimated to be during the wet season (~ October-
548 April), but the phenology of *P. pepeii* in the MNP and elsewhere is largely unknown. The use of point-dendrometer bands that
549 record high resolution variations of stem circumference may reduce uncertainty regarding the onset and cessation of tree-
550 growth during the hydrological year and its relationship temperature variability between February to May.

551

552 **4.3. Large-scale climate variability impacting tree growth in tropical Andean treelines**

553 In addition to site-level conditions at this treeline, the recent decline in annual growth may also be linked to broader scale
554 hydroclimate patterns observed in tropical south America. In the central Andes treeline, negative RW trends in *P. tarapacana*
555 since the 1970s were attributed to increasing drought conditions in southern Peru and northern Chile (>17°S; 4657-4800
556 m.a.s.l.) (Morales et al. 2023). Increasing drought frequency has been linked to delayed wet season onset (Espinoza et al., 2016;
557 Fu et al., 2013; Marengo et al., 2011). One possible explanation is that the intensification of the atmospheric Hadley Circulation
558 during the 20th century, driven by warming SST in the tropical Atlantic, has led to a weakening of zonal moisture transport
559 and in increase in subsidence during transitional dry-to-wet seasons in the central Andes (Beveridge et al., 2024; Espinoza et
560 al., 2021, 2019; Sierra et al. 2022; Yoon and Zeng, 2010. In that context, the decline observed in *P. pepeii* may be part of a
561 wider pattern of increasing hydroclimatic stress at high elevations in the tropical Andes. At the same time, the Keara site differs
562 from drier *Polylepis* treelines such as *P. tarapacana* (Morales et al. 2023;), so annual growth-comparisons should be interpreted
563 cautiously.

564

565 In contrast to a drier start to the wet-season, a recent increase in precipitation at the end of the wet season has been observed
566 in lowland regions below 2000 m.a.s.l. in the northern Andes and Amazon Basin (Arias et al., 2021; Espinoza et al., 2021,
567 2019; Malhi et al., 2008; Zanin and Satyamurty, 2020). These studies argue that the warming of the Atlantic Ocean, land-
568 surface changes in the Amazon, and wind anomalies are the primary factors contributing to changes in circulation and specific
569 humidity in tropical south America overall (see references in (Beveridge et al., 2024)). Contrasting precipitation trends have
570 been observed at various latitudes and elevations within Bolivia as well. tree-ring oxygen isotopes ($d^{18}O$) from *Cedrela odorata*
571 in the lowland Bolivian Amazon reflected an increase in November-March precipitation between 1980-2010 (10°5'S, 66°18'W;
572 106 m.a.s.l.), while treeline *P. tarapacana* $d^{18}O$ from the Bolivian Altiplano recorded a decrease in December-March
573 precipitation between 1992-2012 (22.3°S, 67.23°W; ~4600 m.a.s.l.; Baker et al., 2016; Cintra et al., 2025; Rodriguez-Caton et
574 al., 2024). Even though these hydroclimate trends are complex and spatially variable, having *in situ* high-resolution climate
575 proxies at the Andes-Amazon treeline such as this *P. pepeii* may be useful in understanding long-term changes in orography.

576

577 Overall, growth variability at this *P. pepeii* treeline is modulated by temperatures and moisture available at the site, which may
578 be influence by small-scale climate dynamics driven by orography, but may also reflect broader hydroclimatic changes across
579 the tropical Andes and adjacent Amazon basin mentioned above. *Rather* than implying a shared mechanism across all Andean
580 treelines, our results suggest that distinct *Polylepis* treelines may be responding to a common regional trend toward warmer

581 and, in some cases, effectively drier conditions during the peak wet-season near the Andes-Amazon treeline (~November-
582 March). Overall, this newly-generated record has provided in situ evidence of a long-term, and negative growth response of a
583 tropical treeline to increased temperature trends in Northwestern Bolivia between 1960-2015 (14°S, 4400 m.a.s.l).

584 **5 Conclusions**

585 We report a significant decline in radial growth in *Polylepis pepeii* at a tropical treeline in the southwestern Amazon Basin
586 (~4400 m a.s.l.). Our results indicate that this decline is associated with increasingly warm and dry conditions since the 1960s,
587 particularly through reduced precipitation and higher minimum temperatures during the wet season preceding ring formation.
588 Together, these changes suggest increasing moisture stress at the site. Rising minimum temperatures, especially when not
589 matched by equivalent increases in maximum temperatures, may also reduce moisture availability by lowering relative
590 humidity, altering local moisture transport, and increasing nighttime respiratory carbon losses. More broadly, warmer
591 conditions are likely to enhance evaporative demand and further constrain tree growth. We also found a tendency for narrower
592 rings in the year following warm and dry El Niño events, whereas the influence of La Niña remained less clear, suggesting an
593 asymmetric ENSO effect that deserves further study. Expanding the *P. pepeii* network at Keara could provide the basis for the
594 first tree-ring-based climate reconstructions for northern Bolivia, extending the short instrumental record back to the mid-
595 nineteenth century. Overall, our findings provide new insight into the vulnerability of tropical treelines to ongoing warming
596 and hydroclimatic change, and they highlight the value of *P. pepeii* for future dendrochronological and ecophysiological
597 research in the tropical Andes. Future work should thus focus on quantifying leaf respiration, temperature sensitivity, and the
598 timing of wood formation using ecophysiological measurements and point dendrometers, to better understand the mechanisms
599 linking climate change and growth decline in this species

600 **Author contribution**

601 Co-authors LAH, DRC, MRC, MEF, RO, ET, AF collected tree samples in Keara. CM and AF organized field campaigns and
602 provided essential information regarding Madid National Park. RDD and LAH assisted in initial experimental design and
603 editing. Tree-ring dating, measuring, and analyses was conducted by corresponding other RO. HTTN, aided in correlation
604 analyses and concept. APS assisted in wood anatomical methods. Editorial review provided by all co-authors. Site images for
605 Figs. 1 and A2 were taken by RO.

606 **Competing interests**

607 The authors declare that they have no conflict of interest

608 **Acknowledgements**

609 This work was made possible by the following funding sources from the U.S. National Science Foundation: AGS-1702789,
610 AGS-1903687, AGS-2303524, and OISE-1743738. We acknowledge the NSF AGS-1903690 grant at UC Irvine KCCAMS
611 facility for seven radiocarbon measurements. E.T. received funding from the Comunidad de Madrid program Atracción Talento
612 “César Nombela” grant number 2023-T1/ECO-29118. We thank Marco Tedesco’s Lab at LDEO for making the Echo
613 microscope camera available for the anatomical images. This work is dedicated to Renaud, his family and the community of
614 Keara for their hospitality, knowledge, and assistance in the field. Special thanks the Nacional Herbario in La Paz, including
615 Alfredo Fuentes and Freddy “Zen” Ruiz, for their guidance during the 2012 and 2019, 2023 Bolivian field campaigns.

616 **Data Availability**

617 The *Polylepis pepei* RW chronology from Keara will be made publicly available on the NOAA International Tree-ring
618 Database.

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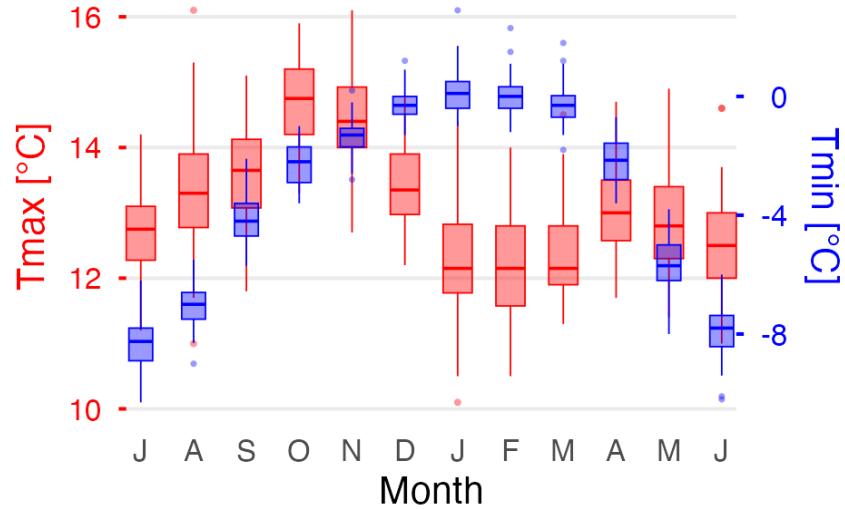
El Niño DJF	La Niña DJF
1983	1974
1998	1917
1973	1956
1931	1976
1992	2011
1966	1989
1919	1910
1926	1971
1958	1951
1897	2000
2010	1918
1987	2008
1942	1943

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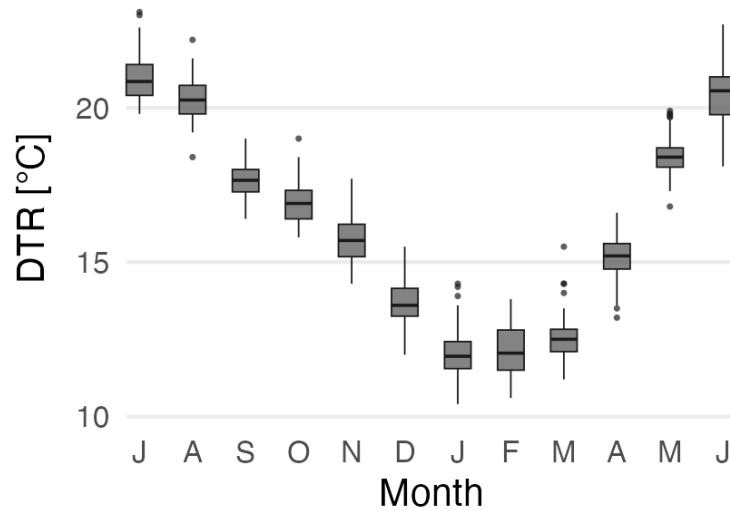
641 **Table A1. NOAA-PSL list of 26 top ranked DJF-ENSO events (13 each) which were selected for SEA analyses of *P.***
642 ***pepei* RW. The years provided for the DJF seasons are centered on January. The remaining 22 top ranked events (48**
643 **total) are available on the website (https://psl.noaa.gov/enso/past_events.html).**

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A
Diurnal temperature range 1960–2015



B



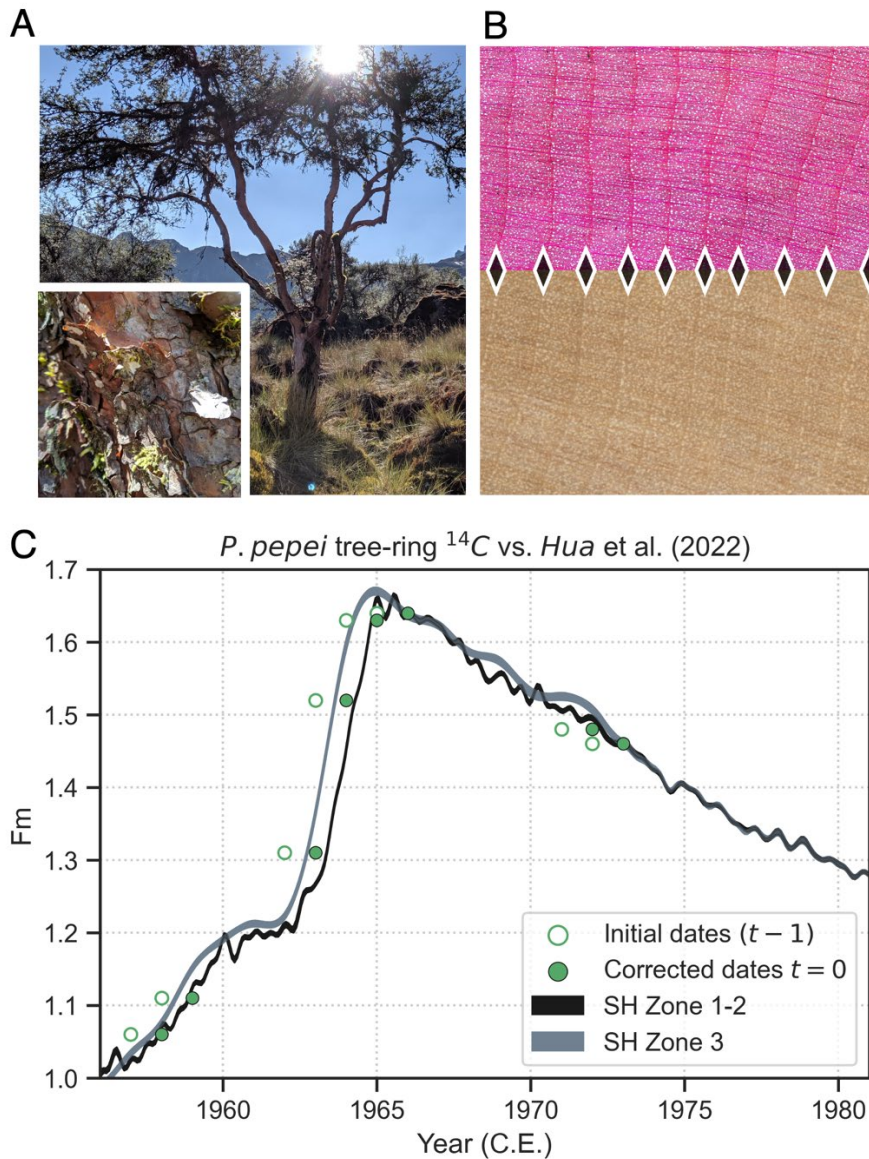
645

646 **Figure A1: (A) Boxplots of the mean monthly distribution of maximum and minimum temperatures and (B) the diurnal**
 647 **temperature range (grey;DTR) for the study region between 1960–2015 using the nearest CRU 4.08 grid point data**
 648 **(14.75,69.25). Boxplots include the temperature median (horizontal line), 1.5x the inter-quartile range (whiskers), and**
 649 **outliers (colored dots).**

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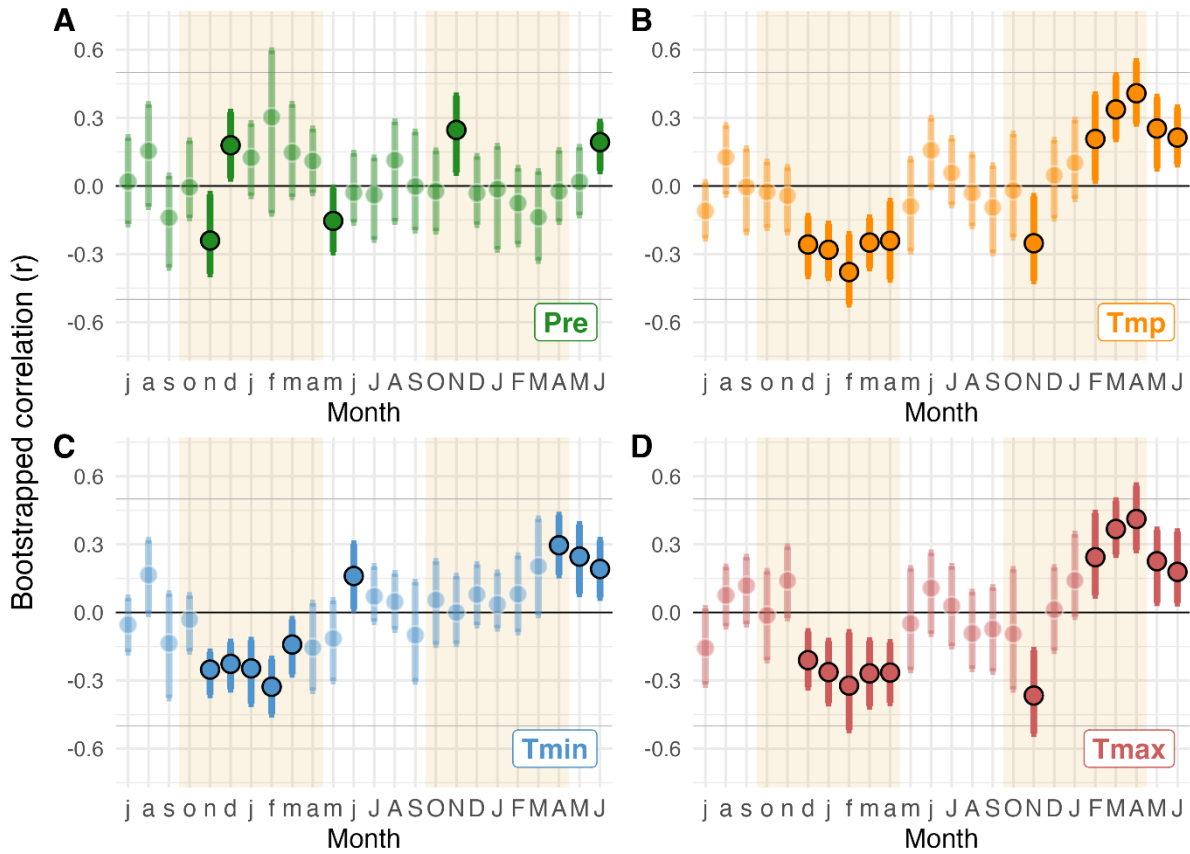
651 **Appendix A2. Photo of *Polylepis pepeii*, Wood anatomy and Radiocarbon Analyses**

652 At early stages of RW chronology development for the *Polylepis pepeii* network, initial radiocarbon results (Fig. A2.C) showed
653 that ^{14}C measurements of SP20X were offset by 1 year in relation to the SH Zone ^{14}C curves. In the original *P. pepeii* RW
654 timeseries generated with material collected in October 2012, the first ring behind the bar was not measured as it was
655 considered an incomplete growth year. Upon a recent inspection of these samples, we observed that the final ring behind the
656 bark did not always correspond to a partial ring as some trees had not yet started wood formation for the 2012 growth year (at
657 least for the side of the stem where the core was sampled). Therefore, the calendar year assigned to the last complete ring for
658 the KEPP RW samples was corrected to 2011. This date-adjustment on the samples from the 2012 campaign was confirmed
659 after cross-dating RW from additional living trees collected in 2019 from the Waca-cocha closed-canopy forest. The final RW
660 chronology shown included 31 trees from 2019 (combined mean correlation, $r= 0.50$). In summary, the traditional
661 dendrochronological crossdating techniques of RW measurements, wood anatomical cuts, and radiocarbon results confirmed
662 *P. pepeii* in Keara formed annual rings.



663
 664 **Figure A2. (A) Photo of a *P. pepei* tree in Keara during the dry season in July 2019. The trees have evergreen foliage,**
 665 **and can appear shrublike, with twisted, and at times multiple, stems. The bark consists of thick layers of compressed**
 666 **flakes that are red and brown in color. (B) *P. pepei* wood anatomy for a histological slice (40X magnification Echo**
 667 **microscope camera) and scanned image of a tree-core (3200 dpi). The direction of radial growth is from left to right**
 668 **(pith to bark) while diamond shapes indicate the latewood/earlywood boundary between annua tree-rings. *P. pepei***
 669 **tree rings feature large, uniformly distributed vessels in the earlywood while the latewood includes both solitary**
 670 **vessels and thicker, fiber-like tracheid cells. The vessel lumen area in the latewood appears to taper tangentially in**
 671 **size before each subsequent growth-ring boundary. (C) Radiocarbon measurements (green circles) from the alpha-**
 672 **cellulose of selected rings in a cross-section is plotted with the Hua et al. (2022) reference curves SH Zone 1-2 and SH**
 673 **Zone 3.**
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Residual Monthly Climate vs. Residual RW



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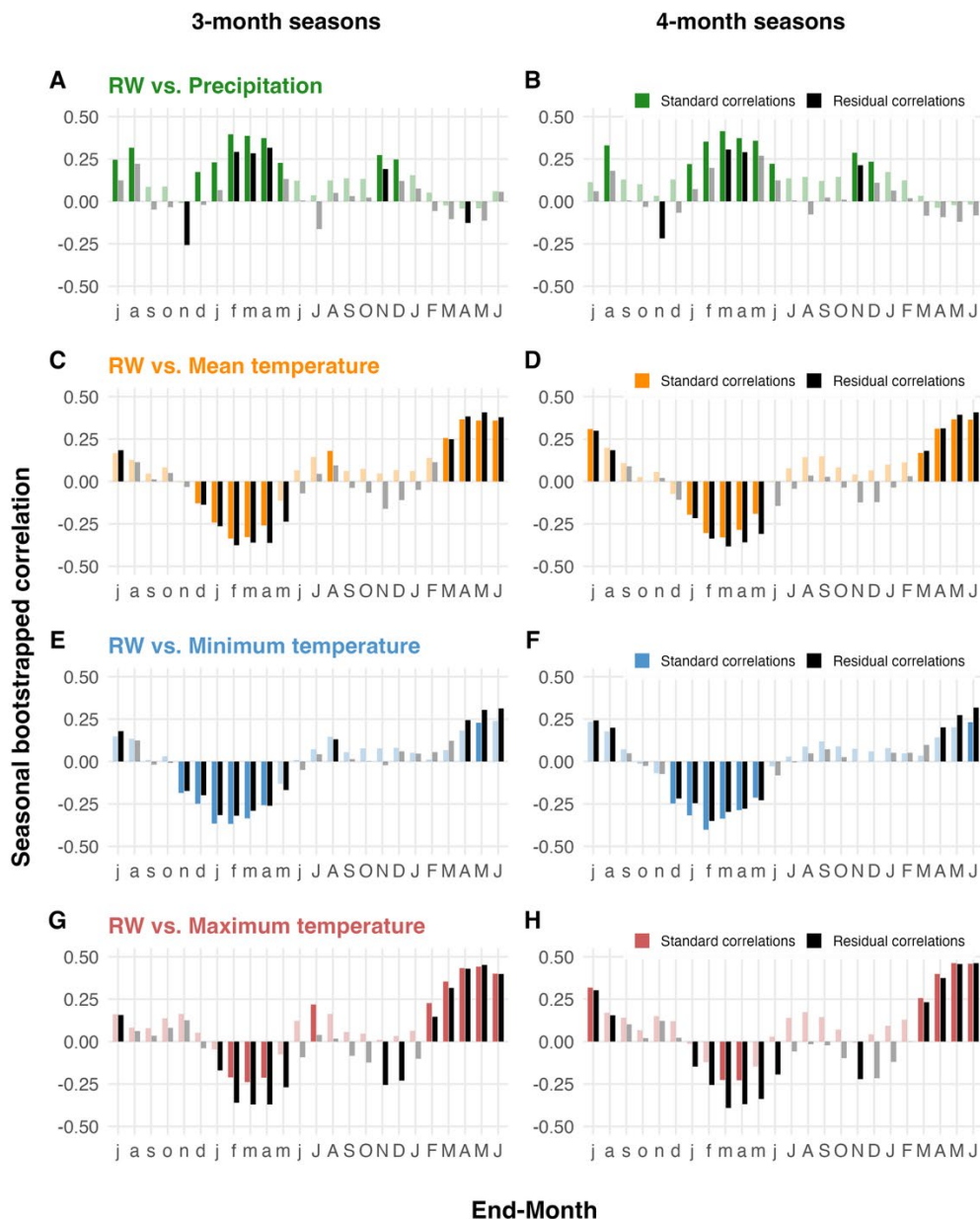
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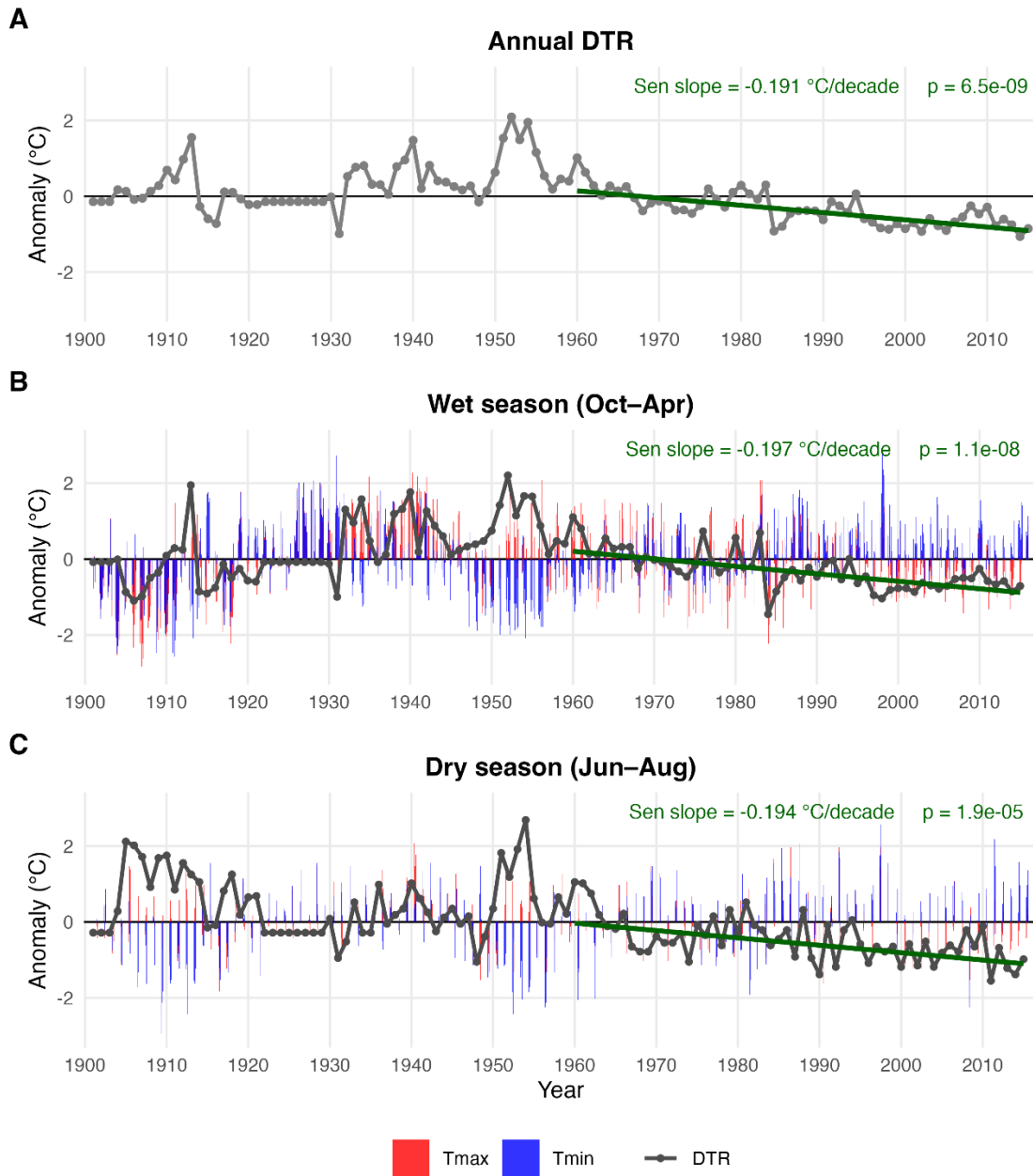
Figure A3. Residual *P. pepei* RW correlations with Residual monthly climate data 1960-2015. The residuals from a linear regression were applied to the local monthly precipitation (A), and mean, minimum, and maximum temperature series for the site (B-D) to evaluate the interannual relationships between annual RW and monthly climate variables. Lowercase letters on the x-axis represent prior-year climate (lag=1). Correlations are reported as the median r estimated using stationary bootstrapping. Significance was defined from 95% bootstrapped confidence intervals and are denoted in solid-colored circles outlined in black. Tan-shading represents the estimated ‘wet-season’ for the treeline site which extends between October-April for this period.



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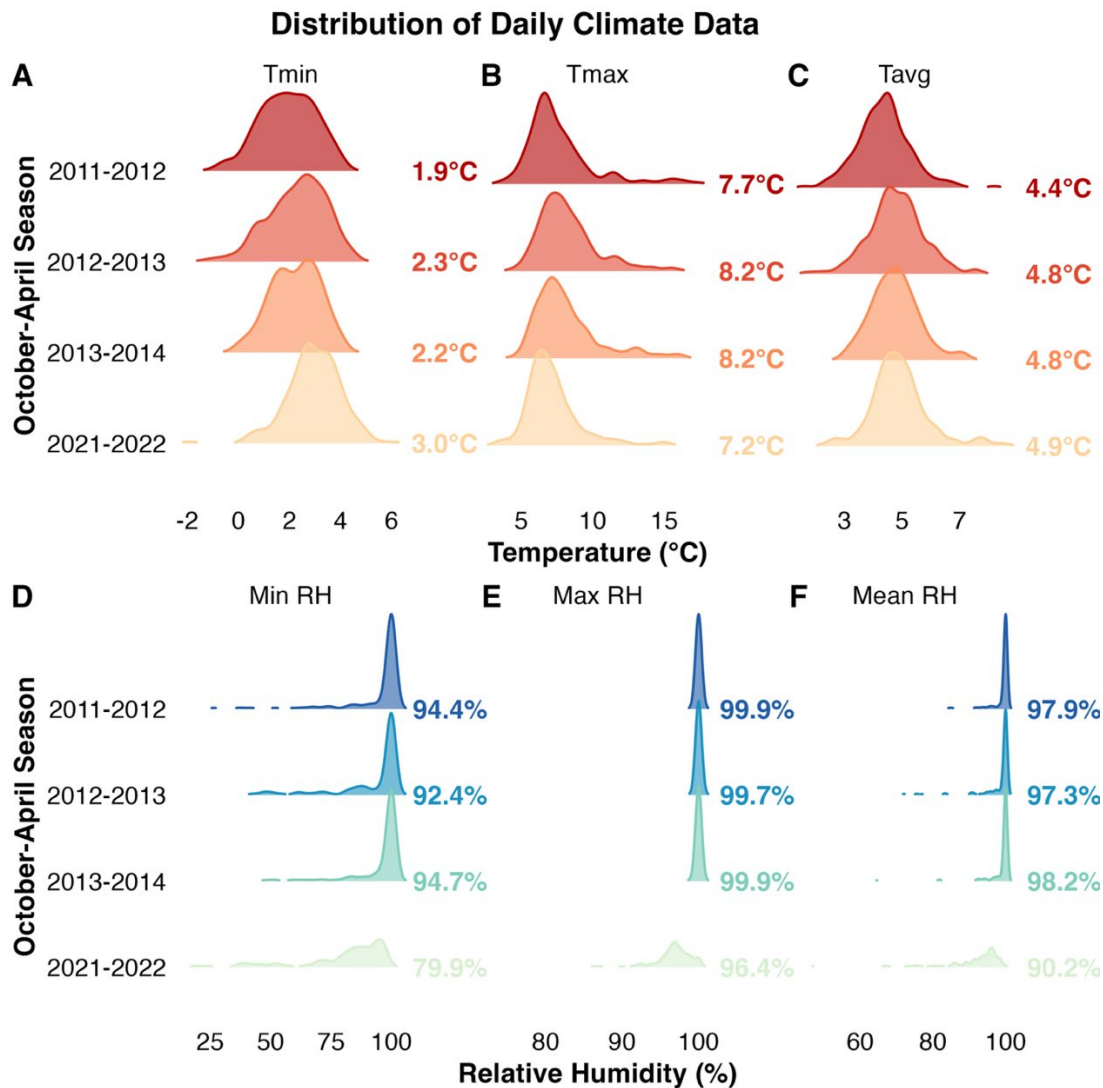
684 **Figure A4. Running seasonal climate correlations with *P. pepei* RW for 3-month (left column) and 4-month (right column)**
 685 **seasonal averages using local temperature and precipitation data (1960-2015). ‘Residual correlations’ are**
 686 **represented as black bars (i.e. detrended climate vs. residual RW), while colored bars represent ‘standard correlations’**
 687 **(i.e. mean climate vs. standard RW). The x-axis represents the ‘end-month’ of the season, for example, lowercase ‘f’ in**
 688 **panel (A) represents correlations between RW and prior-year December-February precipitation (lag=1). Significance**
 689 **was inferred from 95% bootstrapped confidence intervals. Non-significant seasonal correlations are faded, while**
 690 **significant correlations are solid-colored bars.**

1901–2015 Temperature anomalies near the Keara treeline



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692 **Figure A5. Long-term temperature anomalies for the Keara site (1901–2015). The grey timeseries represents annual**
693 **(A) and seasonal (B–C) diurnal temperature range for the nearest gridpoint from CRU TS 4.08. 1901 is the earliest year**
694 **of available data for CRU. Linear trends in DTR anomalies were estimated using Sen’s slope. The significance of the**
695 **1960–2015 trend was evaluated using a two-tailed Mann-Kendall test. The mean slope is reported as the average change**
696 **in DTR per decade (°C). Seasonal anomalies for Tmax (red) and Tmin (blue) are represented as vertical bars for the**
697 **1901–2015 period (B–C).**



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Figure A6. Ridgeline distributions of daily temperature and relative humidity between October–April for the 2011–2012, 2012–2013, 2013–2014, and 2021–2022 seasons at the site. This data was resampled from hourly measurements recorded *in situ* from dataloggers installed near the Keara open-canopy site in 2011 (see section 2.1). The top row (A–C) represents daily minimum (Tmin), maximum (Tmax), and mean (Tavg) temperature. The bottom row (D–F) displays minimum, maximum, and mean relative humidity (RH; bottom row). Colored ridges represent daily observations within each season, and bold values at the right margin indicate seasonal means for each year.

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