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Age-related drought sensitivity of Atlas cedar in the Moroccan Middle Atlas forests

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Abstract

Investigations of age-related tree responses to climate change are still poorly understood at the individual tree level. In this paper, we seek to disentangle the relative contribution of tree age to growth decline and growth-climate relationships in Atlas cedar (*Cedrus atlantica*) forests at the Middle Atlas Mountains, northern Morocco. Dendrochronological methods were applied to quantify growth-climate relationships using tree-ring width indices (TRWi) calculated for cedars of two contrasting age groups (old trees, age ≥ 150 years; young trees, age < 150 years). TRWi-climate relationships were assessed at the site and tree levels by using response functions and linear mixed-effects models, respectively. Growth of the studied Atlas cedars was negatively affected by recurrent droughts and by the steep temperature rise since the 1970s. Response functions and mixed-effects models indicated that the decline in tree growth was mainly explained by diminishing precipitation. The negative association between cedar growth and temperature was stronger in old than in young trees, whereas growth variance explained by precipitation was slightly higher in young than in old trees. Vulnerability to temperature-induced drought stress in old cedar trees may lead to an impending growth decline. We argue that the age dependence of growth sensitivity to drought must be quantified and considered at the individual tree level when predicting the future dynamics and persistence of cedar forests in the Moroccan Middle Atlas.

Keywords: *Cedrus atlantica*, dendroecology, drought, growth-climate relationships, linear mixed-effects models, standardized precipitation index.

50 **Introduction**

51 Drought-related tree growth decline and associated increased mortality risk are recurrent
52 phenomena that have been reported in a variety of forest communities around the world
53 (e.g., Dobbertin, 2005; van Mantgem et al., 2009; Allen et al., 2010; Linares and
54 Camarero, 2011). Drought may intensify physiological stress on long-lived woody
55 vegetation, leading to sudden growth reduction (Bigler et al., 2004; McDowell et al.,
56 2008; Linares and Camarero, 2011; Hereş et al., 2011; Lloret et al., 2011). Growth
57 decline in response to severe droughts may trigger widespread mortality which can
58 transform pine, fir or cedar landscapes to more open stands dominated by other woody
59 species more resistant to drought, including oaks and junipers, at regional to local scales
60 (Galiano et al., 2010; Gonzalez et al., 2010; Linares et al. 2011).

61
62 Radial growth can be used as a surrogate of the whole-tree carbon budget (e.g., Litton et
63 al., 2007; Zweifel et al., 2010; but see also Rocha et al., 2006). Such use implies that
64 tree-ring data provide valuable information about the effects of environmental change
65 on forest productivity. Dendrochronology has long used tree-rings to better understand
66 climate-growth relationships in trees within a site or across a region, with tree age often
67 considered to contribute unwanted noise to the signal (Cook, 1985; Osborn et al., 1997;
68 Vaganov et al., 2006). Thus, age-dependent effects often have not been taken into
69 account in investigations of the interactions between climate change and tree growth
70 decline (Voelker, 2011).

71
72 The most common approach in dendroecological studies to deal with age growth trends,
73 i.e. the tendency of tree rings to get narrower as trees get bigger and older, is to detrend
74 tree-ring width series by calculating ratios or differences between the raw data and fitted

curves (e.g., negative linear or exponential curves) to obtain site mean chronologies averaging the indexed values across trees and for each year (Cook and Peters, 1981; 1997). Therefore, such pooled tree-ring indices do not account for the variance in climate-growth responses among trees. Moreover if the age growth trends of some trees covary with the climatic trend, detrending would also remove part of the environmental signal of interest. Here, we combined both standard detrending methods, widely used among dendrochronologists, with linear mixed-effects models by multi-model selection criteria, to model, at an individual tree scale, age-dependent growth responses to local temperature and precipitation. We provide a novel advance for analysing the effects of both climatic stress and intrinsic age effects on growth, using linear mixed-effects models to reproduce changes in tree-ring width indices at the individual tree level.

We focused on Atlas cedar (*Cedrus atlantica* Manetti) because it is an endemic and very long-lived tree species, which is considered to be highly vulnerable to climatic warming and the related decrease in soil-water availability (Cheddadi et al., 2009; Linares et al., 2011). Since the early 1980s, severe droughts have been related to Atlas cedar growth decline and mortality (Bentouati, 2008; Mokrim, 2009; Allen et al., 2010), mainly in the mountainous xeric areas near the Sahara Desert (Chenchouni et al., 2008). The Atlas cedar growth decline seems to affect mainly old dominant trees, which suggests age/size-dependent tree responses to long-term changes in water availability (Linares et al., 2011).

We hypothesised that: (i) growth response to drought increases is modulated by inherent age-dependent sensitivity to temperature, precipitation and drought; and (ii) age affects the vulnerability of tree species to climate-induced stress. To test these hypotheses, we

investigated local climatic trends (temperature and precipitation) for the 20th century, and we measured tree-ring width and calculated tree-ring width indices from *C. atlantica* trees sampled in the Middle Atlas, northern Morocco. Our specific aims were: (i) to quantify the extent to which temperature, precipitation and drought drive the *C. atlantica* climate-growth patterns in trees of different ages, and (ii) to evaluate whether age-dependent responses to climate could explain recent Atlas cedar growth decline. The fulfillment of these aims may shed light on the nature of age-dependent links between climate change and growth in trees. Furthermore they could allow evaluating whether variables such as tree age are appropriate to forecast drought-induced forest decline.

Methods

Study species

Currently, the genus *Cedrus* includes three extant species native from the Mediterranean mountains (*C. atlantica* from Algeria and Morocco; *Cedrus libani* Rich. in Asia Minor; *Cedrus brevifolia* (Hooker fil.) Henry in Cyprus) and another one in the Himalaya (*Cedrus deodara* Don; Farjon, 2008). The Atlas cedar forests cover an area of over 130,000 ha (Cheddadi et al., 2009) distributed over Morocco (Rif, Middle Atlas, and north-eastern High Atlas) and Algeria (Ouarsenis, Aurès and Djurdjura; Linares et al., 2011). *C. atlantica* occurs at elevations of 1300 to 2600 m a.s.l., where the amount of annual rainfall ranges from 500 to 2000 mm and the minimum temperature of the coldest month ranges between -1° C and -8° C (Benabid, 1994; Mhirit, 1994). The Middle Atlas in northern Morocco contains about the 80 % of the world's Atlas cedar forest area (ca. 100,000 ha; Benabid, 1994; Linares et al., 2011). The Atlas cedar has relatively wide tolerances with regard to climate and soil type. Middle Atlas cedar

forests contain several evergreen (holm oak, *Quercus rotundifolia* Lam.; prickly juniper, *Juniperus oxycedrus* L.; European holly, *Ilex aquifolium* L.) and deciduous (*Acer opalus* Mill., *Crataegus laciniata* Ucria) tree and shrub species. The most abundant tree species, beside Atlas cedar, in the stands studied was *Q. rotundifolia* (Linares et al., 2011).

Climate data

To quantify temperature and precipitation trends over the second half of the 20th century (period 1950-2006) we used local climatic data (monthly mean temperature and total precipitation) from the Ifrane meteorological station located about 14 km apart from the studied plots (33° 32'N, 5° 07' W, 1630 m, period 1958-2003 and estimated the other values through linear correlations using gridded regional data from the CRU TS 3.0 dataset produced by the Climate Research Unit (CRU 2008). Although all correlations between local and regional data were highly significant, we only used the local dataset to model tree growth in further analyses (see Linares et al., 2011). The annual water budget was estimated by the Palmer drought severity index (PDSI) obtained from gridded regional data from the CRU TS 3.0 dataset (CRU 2008). Mean temperature and PDSI were standardized by subtracting the mean and dividing by the standard deviation:

$$Z_i = \frac{(x_i - \bar{x})}{\sigma} \quad (1),$$

where Z_i expresses the x_i score distance from the x average ($\bar{x}$) in standard deviation units (σ).

To determine the severity and the rarity, in statistical terms, of extreme drought events, precipitation data were standardized by calculating the Standardized Precipitation Index (SPI; McKee et al., 1993). The SPI accounts for the frequency distribution of

precipitation. Conceptually, the SPI also represents a *z*-score of an event, i.e. the number of standard deviations above or below the mean of that event, as has been defined in Eq.1 for temperature and PDSI values. However, the SPI performs a pre-adjustment to this formulation because precipitation is usually positively skewed (Bordi et al., 2001). To adjust for this statistical bias in rainfall data, the precipitation data was transformed to a more normal distribution by applying the gamma function (see Appendix 1 for a detailed report about SPI computation).

Dendrochronological methods

To quantify the growth patterns of *C. atlantica*, a total of 53 dominant and co-dominant trees were selected based on an extensive field survey within an area delimited by the following coordinates: latitude 33°24'40"-33°24'60"N and longitude 5°03'00"-5°08'40"W. The elevation varied from 1830 to 1890 m, i.e. near the lowermost elevation limit of the distribution of this species in the Middle Atlas (Benabid, 1994; see also Linares et al., 2011). All trees were located on limestone substrates. We selected and specifically sampled trees showing regular and straight boles, and we avoided those with asymmetrical or eccentric growth. We then extracted from two to four cores per tree at breast height (1.3 m) and always in a direction perpendicular to the terrain maximum slope using 40-cm or 60-cm long Pressler increment borers, depending on the diameter of the trees sampled. Bark and core length were measured in the field (as well as tree diameter). When the core did not reach the pith (ca. 10 % of the cored trees), total tree age at coring height was thereafter estimated by linear regressions assuming an averaged growth rate for the missing innermost rings.

All cores were sanded with sandpapers of progressively finer grain until tree rings were clearly visible under a binocular microscope, and then were visually cross-dated. Tree-ring widths were measured to the nearest 0.001 mm using a LINTAB measuring device (Rinntech, Heidelberg, Germany), and cross-dating quality was checked using the COFECHA program (Holmes, 1983).

The trend due to the geometrical constraint of adding a volume of wood to a stem of increasing radius was corrected by converting tree-ring widths into indices following the standard dendrochronological procedures based on detrending. For each individual tree, standardised tree chronologies were built from cross-dated ring-width series using the ARSTAN program by using a two-step detrending procedure (Cook, 1985). Negative exponential functions were first fitted to the series, and then a cubic smoothing spline with a 50% frequency response of 100 years was fitted. We acknowledge the fact that the applied detrending will limit our findings to inter-annual and decadal fluctuations. The detrended series were then standardized and transformed into dimensionless indices by dividing the observed by the expected values given by the functions. Then, autoregressive modelling was performed on each series to remove the serial correlation between tree-rings and to obtain residual series of tree-ring width indices (TRWi). We calculated chronologies at the individual level by averaging TRWi values for each tree. These data were used for individual climate-growth analyses (see section *Linear mixed-effects models of TRWi at the individual tree level*). To detect which were the main climatic drivers of growth at the population level we obtained a mean chronology of all trees by averaging the indexed values for each year using a bi-weight robust mean. We also obtained mean chronologies for the two age groups analysed (old trees, age ≥ 150 years; young trees, age < 150 years).

199 We related the mean site residual chronology calculated considering all trees to
200 standardized monthly climate data (mean temperature, SPI, PDSI) for the period 1950-
201 2006 using response functions. Response function coefficients were based on
202 bootstrapped stepwise multiple regressions computed on the principal components of
203 climatic variables (Fritts 1976). The bootstrapped response function is a robust
204 approach to test the significance of the regression climate-growth coefficients within a
205 specific time period and avoids the spurious effects caused by the existence of
206 collinearity among climatic variables. We calculated the response coefficients from
207 September of the previous year (year $t-1$) up to September of the year of tree-ring
208 formation (year t). Mean regression coefficients were regarded as significant ($p < 0.05$) if
209 they were at least twice their standard deviation after 999 bootstrapped iterations.
210 Bootstrap response functions were calculated using the program Dendroclim2002
211 (Biondi and Waikul, 2004).

212

213 Several dendrochronological statistics were calculated for the common period 1950-
214 2009 to characterize the raw (MW mean tree-ring width; SD, standard deviation; AC,
215 first-order autocorrelation) and residual chronologies of the two age groups considered
216 (ms_x , mean sensitivity; r_{bt} , mean between-trees correlation; EPS, expressed population
217 signal; E1, percentage of variance explained by the first principal component; Fritts,
218 1976). The first-order autocorrelation of ring-width series (AC) measures the year-to-
219 year growth similarity, whereas the mean sensitivity (ms_x) of residual ring-width series
220 quantifies the relative year-to-year variability in width of consecutive tree rings. Finally,
221 the mean between-trees correlation (r_{bt}) is a measure of the similarity in growth among
222 trees, the expressed population signal (EPS) is a measure of the statistical quality of the
223 mean site chronology as compared with a perfect infinitely replicated chronology, and

the percentage of variance explained by the first principal component (E1) is an estimate of the common variability in growth among all trees.

Linear mixed-effects models of TRWi at the individual tree level

We fitted linear mixed-effects models using the *nlme* package in R software (R Development Core Team, 2011) to explain TRWi variability using mean monthly temperature, SPI, and PDSI from September of the previous year ($t-1$) up to September of the year of tree-ring formation (t) as explanatory variables. Climate variables were included as fixed effects, and each tree was included as a random effect based on the following equation:

$$TRWi = I + (\text{Temperature, SPI, PDSI}) + \varepsilon \quad (5)$$

where I is the intercept; Temperature, SPI and PDSI represent the standardized monthly data for these variables; and ε is the error component (Pinheiro et al., 2011). The covariance parameters were estimated using the restricted maximum likelihood method, which estimates the parameters by minimising the likelihood of residuals from the fitting of the fixed-effects portion of the model (see further details in Pinheiro et al., 2011). We used an information-theoretic approach for multi-model selection, based on the Akaike information criterion (AIC; Akaike, 1974) corrected for small sample sizes (AICc). The AIC combines the measure of goodness of fit with a penalty term based on the number of parameters used in the model. Significant variables were selected using a backward step-wise procedure based on AICc. We considered models with substantial support to be those in which the ΔAIC (i.e., the difference of AICc between models) was < 2 (see Burnham and Anderson, 2002).

Growth-climate linear mixed-effects models were fitted using all trees, i.e. not differentiating age classes. Thereafter, we explored the age distribution of the sampled trees and we defined two ages classes accounting for a reasonable compromise between adequate sample size and sufficient data discrimination. Differences between age classes were computed by one-way ANOVA comparing variance explained for each individual tree.

Results

Climate trends

Linear regression analyses performed over the mean temperature series showed a significant increase ($p < 0.001$; Figure 1), which accounted for a warming of $+1.6\text{ }^{\circ}\text{C}$ on average for the 1958-2003 span, being highest from March to June (data not shown). The annual SPI showed a marginally significant ($p = 0.06$) negative trend, which accounted for a decrease of -364 mm in the raw annual precipitation for the 1958-2003 span (i.e. an approximate 28% reduction in the total annual precipitation). The greatest precipitation decrease was found in early spring (March and April, data not shown). Extreme drought events were more pronounced towards the end of the 20th century (Figure 1). PDSI showed a notable decrease ($p < 0.001$, Figure 1) as a result of the above indicated trends corresponding to increasing mean temperatures and decreasing total precipitation.

Figure 1

Growth-climate relationships

The highest and significant response coefficients were obtained for current May SPI and April to August PDSI, whereas the previous September PDSI showed a negative association with TRWi (Figure 2). The model with highest explanatory value, based on temperature variables, included a negative effect for previous September and current May and June, while previous October and current April and September mean temperature showed a positive effect (Table 1). This model accounted for 12.5% of the TRWi variance. PDSI accounted for 24.7% of the TRWi variance, with previous October and current March and September showing a negative effect on TRWi (i.e. higher TRWi as drought increases during these months), while current May and August showed a positive effect (i.e. higher TRWi as drought decreases during these months). SPI variables showed always positive effects on TRWi and accounted for 24.8% of growth variance. SPI-based models rendered slightly higher explanatory value of TRWi variability PDSI-based models (Table 1). SPI of previous November and current January, March, April and May were selected in the model with highest explanatory value.

Figure 2

Table 1

Age-related drought sensitivity of Atlas cedar

Year-to-year serial correlation (AC) was higher in old than in young trees, whereas the reverse was observed in the case of mean ring-width (Table 2). The relative year-to-year variability in width of consecutive tree rings (ms_x) and the similarity in growth among trees (r_{bt}) were similar between the two ages classes.

Trees were grouped into cedars younger (young; n=30) and older (old; n=23) than 150 years, respectively (Figure 3). Young and old trees had mean $\pm$ standard deviation diameters at 1.3 m of 43.14 ± 9.64 cm and 108.11 ± 48.19 cm, and mean ages of 86 ± 18 and 243 ± 72 years, respectively.

Figure 3

Table 2

Atlas cedar showed severe growth reductions during the past century in 1979-1981, 1995, 1999 and 2005 (Figure 4). The year-to-year TRWi variability increased in the 1990s and 2000s as compared with previous decades and the most extreme growth indices observed in these decades were more frequent in young than in old trees (Figure 4). Old trees showed higher responsiveness to temperature than the young ones (18% of explained variance vs. 9%, respectively; Figure 4). At the individual tree level, the variance explained by temperature was significantly higher for old trees than for young ones (Figure 5), while young trees showed slightly higher variance explained by precipitation than old ones (marginally significant; $p=0.09$), and no significant differences among age classes were found for PDSI.

Figure 4

Figure 5

Discussion

Drought induction of Atlas cedar growth decline

We hypothesised that tree vulnerability to climate change could be modulated by age-dependent growth sensitivity to drought, providing valuable information to infer the underlying mechanisms leading to growth decline at the individual tree level. Overall, our results support the contention that rising spring and summer temperatures and decreasing rainfall from winter to early summer will magnify Atlas cedar growth decline (Aussenac and Finkelstein, 1983; Till, 1987; Mokrim, 2009). This conclusion may be also applied to other drought-triggered growth declines which have been reported in other Mediterranean mountain conifers such as *Pinus sylvestris* (Martínez-Vilalta et al., 2008; Hereş et al., 2011), *Pinus nigra* (Linares and Tiscar, 2010), *Abies alba* and *Abies pinsapo* (Camarero et al., 2011; Linares and Camarero, 2011). However, we also found that the effects of rising temperature on *C. atlantica* growth decline are age-dependent, i.e. that the negative effects of temperature warming on growth are stronger in old than in young trees.

In the Middle Atlas, about 23% of the area covered by cedars showed growth decline and dieback symptoms (defoliation, increased mortality) after recent severe droughts (Bentouati, 2008; Mokrim, 2009; Allen et al., 2010). Indeed, we found severe growth reductions during the past century in 1979-1981, 1995, 1999 and 2005. In a longer temporal context, several reconstructions suggest that during the late 20th century Morocco underwent some of the most severe droughts since at least the mid 15th century (Till and Guiot, 1999; Esper et al., 2007; Touchan et al., 2008).

In the Middle Atlas, growth reductions of cedars occurred mainly since 1978, and were related to severe droughts and temperature rise as reflected in the available local climatic records (see also Mokrim, 2009). Growth and monthly precipitation showed

significant positive correlations, based on linear mixed-effects models, from November of the previous year to the current May, in agreement with response function and previous growth-climate analyses, illustrating how precipitation is the main limiting factor for the growth of Atlas cedar (Aussenac and Finkelstein, 1983; Till, 1987), as in other cedar species such as *C. libani* (Akkemik, 2003), *C. brevifolia* (Touchan et al., 2005), and *C. deodara* (Borgaonkar et al., 1999). However, our growth-temperature correlations contrast with those of Till (1987), who reported a positive effect of January mean temperature for Atlas cedar based on response functions computed for the period 1941-1970. It should be noted that the Till (1987) study included Atlas cedar populations from the Rif, Middle Atlas and High Atlas Mountains at elevations ranging from 1400 to 2600 m. Our analyses cover the second half of the 20th century, for which the temperature trend was characterized by a sharp rise, mainly since the 1970s onwards. In addition, our study focuses on Atlas cedar forests near the lowermost elevation limit of the distribution of this species in the Middle Atlas (Benabid, 1994). We expect that low-elevation cedar stands exhibit a lower positive effect of winter-to-spring temperatures on growth than high-elevation stands subjected to cold conditions.

In conifers, severe drought stress affects processes related to cambium activity and xylogenesis, including carbon sink-source trade-offs mediated by xylem and phloem formation (Vaganov et al., 2006). Warming-induced drought stress may alter the partitioning of newly fixed carbon in such a way that less carbon is diverted to secondary growth and wood formation thus negatively affecting hydraulic conductivity (McDowell et al., 2008). Our results agree with the first reasoning since we found a negative association between secondary growth and May-June temperature, the period when Atlas cedar presents maximum radial-growth rates (Aussenac and Finkelstein,

1983). In addition, the negative effect on growth observed for prior September temperature could be related to stomatal control of water loss in late summer to early autumn of the year before tree-ring formation, which may reduce carbohydrate synthesis and storage, increase carbon losses through enhanced respiration, and decrease wood formation in spring (McDowell et al., 2008).

The pronounced decrease in growth of the studied Atlas cedar stands since the early 1980s coincided with the local rise of air temperatures, and the occurrence of severe drought events, and it is also in agreement with long-term drought reconstructions in Morocco (Esper et al., 2007). As stated above, this apparently drought-triggered growth decline may be linked to the dieback of cedar forests in the Moroccan Middle Atlas (Linares et al. 2011). However it should be also noted that, besides drought stress, other factors may also explain the observed growth patterns. For instance, local anthropogenic disturbances (overgrazing, pruning, logging), crown defoliation by processionary caterpillars (*Thaumetopoea bonjeani* and *T. pityocampa*) and cedar bark stripping by Barbary Macaques (*Macaca sylvanus*) seem to be interacting with climate-driven factors on recent Atlas cedar decline (Camperio Ciani et al., 2001; Linares et al., 2011). Therefore they may be involved in the occurrence of alternating phases of growth reduction and recovery in Atlas cedar forests.

Age-related drought sensitivity of Atlas cedar

We demonstrate here that Atlas cedar growth-climate relationships are age-dependent, as has been shown in other tree species growing in harsh environments such as subalpine forests (see for instance Carrer and Urbinati, 2004). This finding suggests the existence of age-related changes in endogenous parameters possibly linked to the

396 hydraulic status and the tree carbon balance. Our results support the idea that rising
 397 spring to summer temperatures become more limiting for growth as trees become older.
 398
 399 The higher xylem hydraulic safety found in old trees as compared with young ones
 400 (Magnani et al., 2000; Domec and Gartner, 2002) suggests the former display lower
 401 sensitivity to water deficit than the latter ones (Ryan et al., 1997). However, increased
 402 size and structural complexity augment maintenance respiration costs (Hunt et al., 1999)
 403 and reduce the efficiency of the hydraulic pathway, which may explain the observed
 404 growth reduction and the increase in temperature sensitivity (Ryan and Yoder, 1997).
 405 Such ontogenetic changes affect xylogenesis (Rossi et al., 2008) and modify growth-
 406 climate relationships (Carrer and Urbinati, 2004). Additionally, old and usually large
 407 and dominant trees should display improved growth conditions by lowering sensitivity
 408 to competition despite increasing tree height leads to decreasing hydraulic conductance
 409 (Domec and Gartner, 2002). Nevertheless, the stronger growth response to temperature
 410 in old than in young trees might be due to a difference in canopy temperature and thus
 411 vapour pressure deficit given that usually canopies of old trees are exposed to more
 412 light than canopies of young trees (Bond, 2000).
 413
 414 Increases in hydraulic resistance could reduce the supply of water for transpiration and
 415 induce earlier stomatal closure to prevent xylem dysfunction through cavitation (Ryan
 416 et al., 2006), which in turn would decrease photosynthetic rates in old trees as compared
 417 with younger trees (Yoder et al., 1994; Bond, 2000). If old trees lose hydraulic
 418 conductivity throughout their trunks, they would need a larger sapwood area to maintain
 419 the same water flux for a given water potential gradient. Then, increased sapwood area
 420 would mean a larger number of parenchyma cells and, hence, higher respiration rates

and also increased temperature sensitivity (Domec and Gartner, 2002; Phillips et al., 2002).

Although PDSI did not show a significant age effect, as Esper *et al.* (2007) also noticed, young Atlas cedar trees showed slightly higher precipitation sensitivity than old ones (but the difference was only marginally significant), and could be more susceptible than the old ones to increased water deficit if the summer drought season becomes longer or more intense due to their limited root system (i.e., poor access to deeper water) and inability to induce stomatal closure and conserve water under moderate levels of vapour-pressure deficit (Wharton et al., 2009). Consistent with this hypothesis, the old trees may be slightly less sensitive than young ones to water availability due to contrasting wood anatomy and xylem function (lower hydraulic conductance and higher xylem safety in older trees; see Domec and Gartner, 2002; Kolb et al. 2007). However, other authors have found few evidences supporting age effects on the climate signal portrayed by tree rings and thus they recommended increasing sample replication when comparing young and old trees (see Esper et al., 2008).

Conclusions

The growth of cedars has progressively declined as aridity has increased in the Moroccan Middle Atlas. Temperature-induced drought stress appears to become increasingly limiting as cedars age, whereas decreasing precipitation exerts a great impact on radial growth in both young than in old trees. This contrasting age-dependent climatic sensitivity may significantly impact future forest dynamics in the Middle Atlas at regional scales, since both a warming trend and a precipitation decline have been forecasted for upcoming decades. Our findings bear implications for Atlas cedar

persistence in the Middle Atlas because ongoing decline of low-density cedar stands may accelerate soil erosion and their replacement by more drought-resistant species as *Quercus rotundifolia*. We argue that the age dependence of growth sensitivity to drought must be explicitly considered at the individual tree level for understanding and predicting the future responses of forests to climate warming.

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Tables

Table 1. Selection criteria for linear mixed-effects models of tree ring-width indices (TRWi) considering all Atlas cedars trees (young and old trees) together (n=4731 rings; 53 trees). A null model considering TRWi as a constant was also tested. Abbreviations: k , number of parameters included in the model (number of explaining variables plus one constant plus the error); Δi , difference in AICc (Akaike information criterion corrected for small samples) with respect to the best model; $L(gi/x)$, Likelihood of a model gi given the data y ; nc, non-computed value. The symbols (+) and (–) indicate positive and negative effects, respectively, for the following climatic variables: SPINp standardized precipitation index (SPI) of previous November (+); SPIJA, January SPI (+); SPIMR, March SPI (+); SPIAP, April SPI (+); SPIMY, May SPI (+). POp, previous October Palmer Drought Severity Index (PDSI) (–); PMR, March PDSI (–); PMY, May PDSI (+); PAU, August PDSI (+); PS, September PDSI (–). TSp, previous September temperature (–); TOp, previous October temperature (+); TAP, April temperature (+); TMY, May temperature (–); TJN (–), June temperature (–); TS, September temperature (+).

Model	K	AICc	Δi	$L(gi/x)$	Variance explained (%)
SPINp+SPIJA+SPIMR+SPIAP+SPIMY	7	363.50	0.00	1.00	24.8
POp+PMR+PMY+PAU+PS	7	366.40	2.90	0.23	24.7
TSp+TOp+TAP+TMY+TJN+TS	8	1072.72	709.22	0.00	12.5
null model	2	1363.89	1000.40	0.00	nc

Table 2. Dendrochronological statistics calculated for the ring-width chronologies of each cedar group for the common period 1950-2009.

Group	No. trees/radii	Raw tree-ring data			Residual chronology			
		MW (mm)	SD (mm)	AC	ms _x	r _{bt}	EPS	E1 (%)
Young	30 / 60	1.89a	0.99	0.76a	0.20	0.51	0.89	56.38
Old	23 / 50	1.09b	0.92	0.87b	0.17	0.49	0.86	40.67

Abbreviations: MW, mean ring-width; SD, ring-width standard deviation; AC, first-order autocorrelation (a measure of the year-to-year growth similarity); ms_x, mean sensitivity (a measure of the year-to-year variability in width of consecutive rings); r_{bt}, mean between-trees correlation; EPS, expressed population signal (a measure of the statistical quality of the mean site chronology as compared with a perfect infinitely replicated chronology); E1, variance explained by the first principal component (an estimate of the common variability in growth among all trees at each site).

Comparisons for selected statistics were based on one-way ANOVA and different letters represent significant ($p < 0.05$) differences between young and old cedars.

Figures legends

Figure 1. Evolution of standardized mean annual temperature (T), standardized precipitation index (SPI) and standardized Palmer Drought Severity Index (PDSI) in the study area for the period 1901-2006. We used local climatic data from the Ifrane meteorological station for the period 1958-2003 and estimated the other values through linear correlations calculated between local and gridded regional data (CRU TS 3.0 dataset). Although all correlations between local and regional data were highly significant ($p < 0.001$), we only used the local dataset to model tree growth in further analyses.

Figure 2. Response function computed on tree-ring width indices based on 53 Atlas cedar trees. Significant response coefficients ($p < 0.01$) are indicated by filled symbols located above bars. The tested monthly climatic variables are the mean monthly temperature, the monthly standardized precipitation index (SPI) and the monthly Palmer drought severity index (PDSI). The analysed temporal window includes the period from September (-1) of the previous year up to September of year of tree-ring formation.

Figure 3. Frequency distribution of age estimated at 1.3 m (25-year classes) of sampled Atlas cedars (*Cedrus atlantica*) from the Moroccan Middle Atlas. The vertical dotted line separates the two age groups analysed, i.e. trees younger and older than 150 years, respectively. These two groups were used for further comparisons in growth-climate relationships.

Figure 4. Observed and predicted changes in tree-ring width index (TRWi) of cedars younger (young; $n=30$ trees) and older (old; $n=23$ trees) than 150 years from the Moroccan Middle Atlas. Standardized mean annual temperature (Temp. Model, ●), standardized precipitation index (SPI model, ▲) and standardized Palmer Drought Severity Index (PDSI model, ◆) were tested as predictors by using linear mixed-effects models.

Figure 5. Tree ring-width index variance explained by the following standardized climatic variables: mean annual temperature, precipitation index (SPI) and Palmer Drought Severity Index (PDSI) computed by linear mixed-effects models. The

variances explained for cedars younger (young; $n=30$ trees) and older (old; $n=23$ trees) than 150 years were compared by using ANOVA tests. In the box-plot figure, error bars represent the 5th and 95th percentiles; boxes represent the standard errors; solid lines represent the median; dashed lines are the means ; and points are outliers. F test values and significance levels (p) are indicated for each climatic variable.

Figure 1.

Figure 2.

Figure 3.

Figure 4.

Figure 5.

Appendix 1. Standardized Precipitation Index (SPI)

To determine the severity and the statistical rarity of extreme drought events, precipitation data was standardized by calculating the *Standardized Precipitation Index* (SPI) for the Ifrane meteorological station. The spatial and temporal dimensions of drought events create problems in generating a drought index because not only must an anomaly be normalized with respect to location, but the anomaly must also be normalized in time if it is to produce a meaningful estimate of drought. The SPI accomplishes both. The SPI is normalized to a station location because it accounts for the frequency distribution of precipitation as well as the accompanying variation at the station. Additionally, the SPI is normalized in time because it can be computed at any number of time scales, depending upon the impacts of interest to the analyst. Conceptually, the SPI represents a z-score, or the number of standard deviations above or below that an event is from the mean:

$$Z_i = \frac{(x_i - \bar{x})}{\sigma} \quad (1),$$

where Z_i expresses the x_i score's distance from the x average ($\bar{x}$) in standard deviation units (σ).

However, the SPI perform a pre-adjustment to this standard formulation due to precipitation is typically positively skewed. To adjust for this mathematical rainfall feature, the precipitation data is transformed to a more normal distribution by applying the gamma function (McKee *et al.* 1993; Edwards and McKee, 1997; Bordi *et al.* 2001).

The gamma distribution is defined by its frequency or probability density function:

$$g(x) = \begin{cases} 1 \\ \beta^\alpha \Gamma(\alpha) x^{\alpha-1} e^{-x/\beta} \text{ for } x > 0 \end{cases} \quad (2)$$

where:

$\alpha > 0$, is a shape parameter; $\beta > 0$, is a scale parameter; $x > 0$, is the precipitation amount;

$\Gamma(\alpha)$ is the gamma function:

$$\Gamma(\alpha) = \int_0^{\infty} y^{\alpha-1} e^{-y} dy \quad (3)$$

Computation of the SPI involves fitting a gamma probability density function to a given frequency distribution of precipitation totals for a station. The alpha and beta parameters of the gamma probability density function are estimated for each station, for each time scale of interest (seasonal, annual, etc.), and for each month of the year. The maximum likelihood solutions are used to optimally estimate α and β as follows:

$$\hat{\alpha} = \frac{1}{4A} \left(1 + \sqrt{1 + \frac{4A}{3}} \right) \quad (4)$$

$$\hat{\beta} = \frac{\bar{x}}{\hat{\alpha}} \quad (5)$$

where:

$$A = \ln(\bar{x}) - \frac{\sum \ln(x)}{n} \quad (6)$$

n = number of precipitation observations.

The resulting parameters are then used to find the cumulative probability of an observed precipitation event for the given month and time scale for the station in question. The cumulative probability is given by:

$$G(x) = \int_0^x g(x) dx = \frac{1}{\hat{\beta}^{\hat{\alpha}} \Gamma(\hat{\alpha})} \int_0^x x^{\hat{\alpha}-1} e^{-x/\hat{\beta}} dx \quad (7)$$

Letting $t = x / \hat{\beta}$, this equation becomes the incomplete gamma function:

$$G(x) = \frac{1}{\Gamma(\hat{\alpha})} \int_0^t t^{\hat{\alpha}-1} e^{-t} dt \quad (8)$$

Since the gamma function is undefined for $x=0$ and a precipitation distribution may contain zeros, the cumulative probability becomes:

$$H(x) = q + (1 - q)G(x) \quad (9)$$

where q is the probability of a zero. If m is the number of zeros in a precipitation time series, q can be estimated by m/n .

The cumulative probability, $H(x)$, is then transformed to the standard normal random variable Z with mean zero and variance of one (see Eq. 1) as follows:

$$Z = SPI = -(t - \frac{c_0 + c_1 + c_2 t^2}{1 + d_1 t + d_2 t^2 + d_3 t^3}) \text{ for } 0 < H(x) \leq 0.5 \quad (10)$$

$$Z = SPI = +(t - \frac{c_0 + c_1 + c_2 t^2}{1 + d_1 t + d_2 t^2 + d_3 t^3}) \text{ for } 0.5 < H(x) < 1.0 \quad (11)$$

where:

$$t = \sqrt{\ln(\frac{1}{(H(x))^2})} \text{ for } 0 < H(x) \leq 0.5 \quad (12)$$

$$t = \sqrt{\ln(\frac{1}{(1.0 - H(x))^2})} \text{ for } 0.5 < H(x) < 1.0 \quad (13)$$

$c_0=2.515517$, $c_1=0.802853$, $c_2=0.010328$, $d_1=1.432788$, $d_2=0.189269$ and $d_3=0.001308$.

Appendix References.

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