Historical Forest Management Practices Influence Tree-Ring Based Climate Reconstructions

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Tree-ring widths (TRW) of historical and archeological wood provide crucial proxies, frequently used for high-resolution multi-millennial paleoclimate reconstructions. Former growing conditions of the utilized trees, however, are largely unknown. Potential influences of historical forest management practices on climatic information, derived from TRW variability need to be considered but have not been assessed so far. Here, we examined the suitability of TRW series from traditionally managed oak forests (Quercus spp.) for climate reconstructions. We compared the climate signal in TRW chronologies of trees originating from high forests and coppice-with-standards (CWS) forests, a silvicultural management practice widely used in Europe for most of the common era. We expected a less distinct climate control in CWS due to management-induced growth patterns, yet an improved climate-growth relationship with TRW data from conventionally managed high forests. CWS tree rings showed considerably weaker correlations with hydroclimatic variables than non-CWS trees. The greatest potential for hydroclimate reconstructions was found for a large dataset containing both CWS and non-CWS trees, randomly collected from lumber yards, resembling the randomness in sources of historical material. Our results imply that growth patterns induced by management interventions can dampen climate signals in TRW chronologies. However, their impact can be minimized in well replicated, randomly sampled regional chronologies.

Keywords: Quercus spp., hydroclimate sensitivity, forest management, tree-rings, coppice-with-standards, climate reconstruction

INTRODUCTION

Medieval societies in Europe developed the “Coppice-with-standards” (CWS) silvicultural system to ensure a sustainable supply with fuelwood and timber for the growing population. This two-story forest structure combines an understory of even-aged coppice harvested in short rotation for fuelwood and tanning with an uneven-aged partial upper story of standard trees growing at wide spacing for timber production (Mosandl et al., 2010). The earliest historical evidence for this practice dates back to the thirteenth century CE (Hausrath, 1982; Hasel and Schwartz, 2006). Dendroarchaeological evidence for CWS management in central Europe during the first millennium...
CE suggests an even longer tradition, at least since the early medieval period (i.e., 500–1000 CE) (Muigg et al., 2020b). The CWS forest management system was used all over medieval and early modern Europe (Short and Hawe, 2012) until it disappeared in most regions by the mid-twentieth century, when fossil fuels had replaced wood as the primary energy source and modern forestry focused primarily on timber production (Groß and Konold, 2010). A small region in southern Germany (Franconia) (Figures 1A,B) provides one of the few places left in central Europe, where traditional CWS forest stands are still maintained (Albrecht and Abt, 2014).

The long tradition of CWS forest management in Europe led to various studies from the fields of forest history and dendroarchaeology (e.g., Bernard et al., 2014; Szabó et al., 2015; Vandeckerkhove et al., 2016). As wood samples originating from archaeological excavations or historical constructions lack information on site and stand conditions, it is likely that, across Europe, a substantial part of oaks originates from CWS managed forests.

Historical tree-ring chronologies are unique archives of past climate variability and as such are frequently used for multi-centennial to multi-millennial paleoclimate reconstructions with annual resolution (e.g., Büntgen et al., 2011; Tegel et al., 2020). However, forest management interventions that might influence tree growth and its sensitivity to climate variability are typically unknown for such samples. So far, possible influences of management-induced alterations in tree-growth patterns on the suitability of CWS standards for reconstructing climate have not been studied. The aim of this study was to compare climate-growth relationships in CWS standards and trees from nearby high forests (HF) (forests consisting mainly of large, tall mature trees with a closed canopy), to critically evaluate their potential for climate reconstructions. We expected a weaker climate signal in CWS standards compared to oak trees from nearby lumber yards in Bad Mergentheim (BAME; 210 individuals) and Wittighausen (WITT; 233 individuals) and a veneer company in Lohr (62 samples). The stem disks (CWS, HF, combined and random) used in this study were collected close to the lower crown height following the random sampling approach (Tegel et al., 2010). The HF dataset consisted only of those trees (128 in total) that did not show cyclic and temporarily aligned growth release events (cf. Muigg et al., 2020b) and therefore presumably originated from high forests. Additionally, we merged the CWS and HF datasets (combined dataset) to assess the hydroclimate signal of a hypothetical chronology containing a relatively large proportion of trees (ca. 50% in this case) with periodic management-induced growth patterns. Finally, we assembled a “random dataset” (see Tegel et al., 2010) by merging the full TRW datasets of WITT, BAME, and Lohr (in total 505 oaks of unknown management) to simulate the random composition of samples from archeological and historical material (Figure 1A).

All samples were prepared with standard dendrochronological methods (Speer, 2010). TRW was measured with a precision of 0.01 mm using a movable object table (Megatron) and recorded using the program Berlin Muehle 4 1.0 (developed by Tobias Heussner). The TRW series were crossdated with PAST software (Knibbe, 2008). Three different detrending methods were tested on the combined dataset after removing auto-correlation (AR), including a 30-year and a 100-year spline (Cook and Peters, 1981; Büntgen et al., 2012) and a modified negative exponential model. Finally, four average TRW chronologies were developed using the bi-weight robust mean (Supplementary Figure S1). The datasets used in this study include oaks with individual tree ages over 200 years. However, we limited our analyses to the period 1902–2013 due to the availability of instrumental climate data for the study region (starting in 1902) and the availability of tree ring records (ending in 2013 for the WITT, BAME, and Lohr datasets).

The quality of the developed chronologies (Figure 1C) as well as the relationships between hydroclimate variables and growth (Supplementary Figure S1) were tested on the raw TRW data and on different detrending methods after removing auto-correlation. The strongest correlations with hydroclimate variables were obtained when applying a 30-year cubic spline detrending. Hence, this detrending method was chosen for further analyses. The quality of the detrended chronologies was assessed by calculating the descriptive statistics EPS (expressed population signal), rbar (mean interseries correlation) and SNR (signal to noise ratio). EPS is an indicator of how well a chronology represents a theoretical infinite population (Wigley et al., 1984) and is used as an indicator of common

**METHODS**

**Study Design and Development of Chronologies**

Four datasets of oak (Quercus robur L. and Quercus petraea (Matt.) Liebl.) TRW series were utilized for the purposes of this study (Figure 1A), comprising (a) oak standards from two actively managed CWS forests (CWS dataset), (b) TRW series of oaks, originating from high forests (HF dataset), (c) a dataset containing both previous (CWS and HF) datasets (combined dataset), and (d) a larger dataset of randomly sampled oak trees (random dataset) (see also Supplementary Figure S2).

The CWS dataset consisted of TRW series from northern Bavaria (Germany), sampled by Muigg et al. (2020b). Stem disks were collected from two actively managed CWS forest stands near Weigenheim (WEIG) and Welbhausen (WELB). To detect signals of CWS in the individual TRW series, we first detected growth releases by using the adapted growth averaging method as described by Muigg et al. (2020b). Following the methods proposed in the study by Muigg et al. (2020b), we applied a standardized scanning for growth releases at an average chronological interval of 26–36 years with less than 5 years standard deviation (SD) on the TRW data from WEIG and WELB to select 135 trees that showed the strongest CWS signal (hereinafter referred to as CWS dataset).

The rest of the datasets (HF, combined and random) that were used here for comparison consisted entirely (or partly in the case of the combined dataset) of oak trees (505 trees in total) from nearby lumber yards in Bad Mergentheim (BAME; 210 individuals) and Wittighausen (WITT; 233 individuals) and a veneer company in Lohr (62 samples). The stem disks (CWS, HF, combined and random) used in this study were collected close to the lower crown height following the random sampling approach (Tegel et al., 2010). The HF dataset consisted only of those trees (128 in total) that did not show cyclic and temporarily aligned growth release events (cf. Muigg et al., 2020b) and therefore presumably originated from high forests. Additionally, we merged the CWS and HF datasets (combined dataset) to assess the hydroclimate signal of a hypothetical chronology containing a relatively large proportion of trees (ca. 50% in this case) with periodic management-induced growth patterns. Finally, we assembled a “random dataset” (see Tegel et al., 2010) by merging the full TRW datasets of WITT, BAME, and Lohr (in total 505 oaks of unknown management) to simulate the random composition of samples from archeological and historical material (Figure 1A).

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The quality of the developed chronologies (Figure 1C) as well as the relationships between hydroclimate variables and growth (Supplementary Figure S1) were tested on the raw TRW data and on different detrending methods after removing auto-correlation. The strongest correlations with hydroclimate variables were obtained when applying a 30-year cubic spline detrending. Hence, this detrending method was chosen for further analyses. The quality of the detrended chronologies was assessed by calculating the descriptive statistics EPS (expressed population signal), rbar (mean interseries correlation) and SNR (signal to noise ratio). EPS is an indicator of how well a chronology represents a theoretical infinite population (Wigley et al., 1984) and is used as an indicator of common
variability in detrended tree-ring chronologies. Rbar is the mean correlation between series within a chronology and indicates the common signal strength in chronologies (Speer, 2010). Finally, SNR is a measure of the desired signal in each chronology versus the amount of unwanted information and random variation (Speer, 2010). All four chronologies displayed high EPS, ranging from 0.97 for the CWS and HF chronologies, 0.98 for the combined to 0.99 for the random chronology. Rbar was found...
to be lowest for the combined (0.283) and highest for the CWS (0.325) chronology.

Climate, Soil Moisture Content, and Drought Indices
We obtained gridded monthly temperature and precipitation data for the study region from the Royal Netherlands Meteorological Institute (KNMI) climate explorer,¹ which were available at a 0.5 × 0.5° resolution (CRU TS.03). Based on these data, we calculated the Standardized Precipitation Evapotranspiration Index (SPEI) using the SPEI package in R (Vicente-Serrano et al., 2010). Although we calculated different accumulation periods of the SPEI index (1, 3, 6, and 12 months) we selected the SPEI index calculated over an accumulation period of 6 months since it was found to correlate best with growth of oaks in the region. In addition, we acquired monthly data of modeled soil moisture content in the upper 100 cm of the soil from the NOAA-CIRES-DOE Twentieth Century Reanalysis V3 datasets (Slivinski et al., 2019). Twentieth Century Reanalysis data accessed by the NOAA/OAR/ESRL PSL, Boulder, Colorado, United States, from their Website on 14/3/2021. All the hydroclimatic variables used in the study were available for the period 1902–2013.

Climate-Growth Relationships and Evaluation of Reconstruction Skills
Hydro-climate sensitivity of the four developed chronologies was assessed by computing bootstrapped Pearson correlation functions between the detrended chronologies and monthly as well as seasonal means for all hydroclimatic variables (average monthly temperature, monthly precipitation sums, SPEI-6 and monthly soil moisture index). Correlation functions were computed for the individual months from April to September and seasonal means for spring (March–May), summer (June–August), as well as for the vegetation period (March–September), the period between April and September (AMJJAS) and the whole year. This analysis was performed for the period between the years 1902 and 2013 for which all the hydroclimatic variables were available. Once the strongest relationship between hydroclimatic variables and oak growth in the region was identified, we applied the commonly used split period (early and late, 1902–1957 and 1958–2013, respectively) ordinary least squares (OLS) regression approach to calibrate the reconstructions. The strength of the relationship between reconstructed and observed values was evaluated by the reduction of error (RE) and coefficient of efficiency (CE), with positive values for these statistics indicating that the reconstruction contains useful information (Fritts, 1976; Cook et al., 1994). The uncertainty in the reconstructions was assessed by calculating the root mean square error (RMSE). We used the functions implemented in the treeclim package in R (Zang and Biondi, 2015) to assess the relationships between annual growth and hydroclimate variability as well as to evaluate the reconstruction skills of each developed TRW chronology. All the analyses were performed using R software (R Core Team, 2013).

¹http://climexp.knmi.nl

RESULTS
The average tree age in the four TRW chronologies ranged from 120.4 (SD = 35.6) years for the HF dataset and 155.0 years (SD = 35.3 years) for the CWS dataset. Average tree age of the combined and random chronologies was 138.2 (SD = 39.4) and 140.1 (SD = 36.8) years, respectively.

Climate-growth correlations for the period 1902–2013 revealed not significant correlations with temperature (Supplementary Figure S1B) but significant positive correlations with all hydroclimate variables that were considered in this study (precipitation, SPEI-6 and soil moisture content) in annual and different monthly and seasonal time windows (Figure 1D). Climate growth correlations were overall strongest for the random chronology, followed by HF and combined chronologies and weakest for CWS. Note that correlations with annual precipitation were only significant for the HF and random chronologies, while mean precipitation in the months from April to September showed significant positive correlations with all chronologies except for the CWS. SPEI-6 showed insignificant correlations with growth of CWS for the early months of the vegetation period (April, May) and soil moisture content only showed significant correlations with CWS for June, July and the summer season. A significant positive correlation was observed between average soil moisture content in the spring months and the random chronology. September soil moisture showed insignificant correlations with all four chronologies. The strongest correlations for all four chronologies were found with June precipitation. Hence, this parameter was used to test the reconstruction skills of the developed chronologies for the period 1902–2013 (Figure 2).

To compare the skills of the four chronologies in reconstructing past climate variability, we performed reconstructions of June precipitation over the period 1902–2013 and compared the reconstructed values against instrumental records from the region (Figure 2). All four models (OLS regressions) used for reconstructions displayed a significance level of 99.9% (p < 0.001). Robustness over time was tested for all reconstructions by splitting the period 1902–2013 into two equal 56-year sub-periods for calibration (1958–2013) and verification (1902–1957). The strongest correlation for the full period (1902–2013) was obtained for the random chronology (Figure 2A; r = 0.47). The correlation coefficient was slightly lower for the calibration period (1958–2013; r = 0.42) than for the verification period (1902–1957; r = 0.51) but still statistically significant. The skill of the reconstruction in capturing precipitation variability is demonstrated with positive RE and CE of 0.26 and 0.25, respectively. Similar results were obtained for HF and combined chronologies (Figures 2B,C), with full-period correlation coefficients 0.44 and 0.42, respectively. In both cases, the correlation coefficients were lower for the calibration period (r = 0.40 and 0.38, respectively) than for the verification period (r = 0.48 and 0.46, respectively). RE and CE were slightly higher for the combined (0.22 and 0.21) than for the HF chronology (0.21 and 0.20). The lowest correlation coefficient for the full period was observed for the CWS chronology (Figure 2D; r = 0.36). Again, correlations remained significant over time,
Different regional June precipitation reconstructions for the period 1902–2013 (split calibration/verification periods 1902–1957 and 1958–2013, respectively) and comparison with instrumental climate data (gray). (A) Random chronology (black), (B) high forest (HF) chronology (blue), (C) combined chronology (orange), and (D) standards (CWS) chronology (green). Highlighted gray areas denote positive and negative hydroclimatic extremes that were used for visually assessing the accuracy of the reconstructions in years with extremely low and high June precipitation.

showing lower values for the calibration period \(r = 0.32\) than the validation period \(r = 0.40\). RE (0.16) and CE (0.15) were considerably lower compared to the other reconstructions but still larger than zero and therefore confirming the skills of the reconstruction.

DISCUSSION

Dataset Composition and Sample Size

Four TRW chronologies were created based on datasets, which contained differing proportions of trees with periodic management-induced growth releases (sustained growth increases). The CWS dataset consisted entirely of trees (135 in total) from actively managed CWS stands (see also Muigg et al., 2020b for further information). In a CWS forest, the reduced competition for light, nutrients and water after periodic understory coppice causes an increased growth in the remaining standards, commonly referred to as release (Müllerová et al., 2016). The absence of such periodic and temporally aligned release events in the HF dataset suggests the absence of substantial alterations in the canopies or understories of target trees, indicative of growing conditions in a high forest (Nowacki and Abrams, 1997; Bergès et al., 2000). To obtain such release-free TRW series a large dataset of 505 oak trees sampled in lumber yards and a veneer company in the same region was scanned for the absence of characteristic growth releases, providing a high forest (HF) dataset with a sample size of 128 trees and therefore equally well-replicated as the CWS dataset. The combined dataset, including all CWS and HF trees, consisted of 263 trees. With half of the trees originating from CWS forests we expected a strong influence of management interventions also in this dataset. For the random dataset, the highest replication of
505 samples was achieved by including all available oak trees, randomly sampled from regional lumber yards and sawmills to simulate the sampling conditions characteristic in historical and archeological samples, i.e., lack of stand information. Note that the lack of information on historical management practices also applies for trees from old-grown stands, where different and changing silvicultural measures in the past cannot be ruled out, especially before the onset of modern forestry, i.e., before the second half of the nineteenth century (Hausrath, 1982; Hasel and Schwartz, 2006).

**Climate Reconstruction**

Not significant climate-growth correlations found with temperature for all four chronologies are explicable with the prevalence of hydroclimate, primarily controlling oak growth in low elevation sites (Büntgen et al., 2010; Tegel et al., 2020). The strong correlations found for June precipitation for all four chronologies, are in accordance with previous studies on the climatic sensitivity of oak growth (e.g., Büntgen et al., 2010). It is, however, striking that the random chronology showed the weakest correlations with hydroclimate variables and demonstrated a lower performance in reconstructing June precipitation than the HF chronology (128 samples). These results further demonstrate the importance of testing individual TRW series for the presence of strong management signals in TRW series (for instance, see Muigg et al., 2020b).

The reconstruction based on the random chronology was also more accurate in years with extremely low and high precipitation sums than the rest of the chronologies. Several years illustrate that the reconstruction reflected the actual precipitation sums best based on the random chronology, successively less in the HF and combined chronologies, while the least accurate was the CWS chronology. Exemplary years, which are also confirmed by other studies (highlighted in Figure 2) include 1915, 1947, 1976, and 2000 (Brázdil et al., 2016; Hänself et al., 2019). It is, however, striking that June precipitation in 2003 (highlighted in Figure 2) is most clearly reflected in the reconstruction based on the CWS chronology (Figure 2D) and successively less pronounced in the reconstructions based on the combined, HF and random chronologies (Figures 2A–C).

Large amounts of archeological findings of oak wood demonstrate the preference of the species for various constructions throughout past epochs and enabled the development of long and well replicated oak tree-ring chronologies in Europe, providing one of the most important proxy records for high-precision multi-millennial climate reconstructions. During the recent decade, several oak tree-ring based paleoclimate reconstructions have been published by various authors (Büntgen et al., 2010, 2011; Cooper et al., 2013; Wilson et al., 2013; Dobrovolný et al., 2014; Cook et al., 2015; Prokop et al., 2016; Muigg et al., 2020a; Tegel et al., 2020). Several authors have stressed the importance of enhanced understanding of past climate to predict future climate variability (Hegerl and Russon, 2011; Woodhouse et al., 2016; Tierney et al., 2020). As oak wood has been the preferred raw material for various purposes in historical times, particularly for construction timber, it was regarded as a valuable forest resource and hence, oak forests have been extensively managed (Haneca, 2011). The CWS management system has been applied on oak forests for at least the past 1,500 years (Muigg et al., 2020b), most likely with varying intensities in different regions and periods (Short and Hawe, 2012; Short and Campion, 2014). Other forms of historical forest management have been practiced all over Europe, e.g., simple coppice management (Stajic et al., 2009; Unrau et al., 2018). Samples from archeological excavations and historical buildings lack information on the individual tree stands (Tegel et al., 2010). As the growing conditions of such samples remain unknown, potential effects of historical forest management cannot be ruled out. However, they can be compensated for by massive sample replication (cf. Büntgen et al., 2010). Having said this, uncertain silvicultural treatment in the early phases of modern oak trees, which are commonly used to calibrate climate-proxy relationships, could easily lead to under- or overestimation of past climate variability and therefore need to be considered as a potential bias in future studies. Similar to the CWS system, other management interventions, such as thinning can influence...
climate-growth relationships (see for example Cescatti and Piutti, 1998; Pérez-de-Lis et al., 2011; D’Amato et al., 2013) and should be also considered in tree-ring based reconstructions. Based on the results of this study, scanning of modern reference datasets for management-induced growth patterns provides an important tool to improve future tree-ring based climate reconstructions.

CONCLUSION

Our results imply that the influence of historical forest management on TRW does affect their suitability for climate reconstructions and should be considered in future paleoclimate reconstructions. It also shows, however, that the negative effects of historical forest management can be mitigated with sufficient sample replication, compensating for the management-induced effects of CWS management and presumably other forms of historical forest management practices. This information is all the more important for future paleoclimate reconstructions in regions with long historical significance and constant human impact on the cultural landscape. Further consideration regarding previous forest management should be given when selecting modern reference material for climate-growth calibration. Historical forest management practices and their influence on tree-ring based reconstructions should be considered in future studies. The field of forest history might provide substantial information for combined studies (Brázdil et al., 2016).

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DATA AVAILABILITY STATEMENT

The datasets generated for this study are available on request to the corresponding author. Requests to access these datasets should be directed to GS, georgios.skiadaresis@waldbau.uni-freiburg.de.

AUTHOR CONTRIBUTIONS

GS and BM designed the study, performed the analyses, and developed the datasets. BM, GS, and WT sampled the modern CWS stands. BM and GS wrote the manuscript with input from WT. All authors provided critical discussion, helped writing the manuscript, and approved its submission.

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SUPPLEMENTARY MATERIAL

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