# 1 Ranking of tree-ring based temperature reconstructions of the past millennium

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# Abstract

1 2 3 Tree-ring chronologies are widely used to reconstruct high- to low-frequency variations in growing season temperatures over centuries to millennia. The relevance of these timeseries in large-scale climate 4 5 reconstructions is often determined by the strength of their correlation against instrumental temperature data. However, this single criterion ignores several important quantitative and qualitative characteristics 6 7 of tree-ring chronologies. Those characteristics are (i) data homogeneity, (ii) sample replication, (iii) growth coherence, (iv) chronology development, and (v) climate signal. A new reconstruction-scoring 8 scheme, based on these 5 characteristics, is designed and applied to 39 published, millennial-length 9 temperature reconstructions from Asia, Europe, North America, and the Southern Hemisphere. By 10 providing a more comprehensive set of criteria by which to evaluate tree-ring chronologies we hope to 11 improve the development of large-scale temperature reconstructions spanning the past millennium. All 12 the reconstructions and their corresponding scores are provided at www.url-here.de.

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#### 15 Keywords

- 16 Paleoclimate, Climate change, Proxy data, Dendrochronology, Dendroclimatology
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#### 1 1. Introduction

2 3 Tree-ring chronologies (TRCs) are an important source of information in large-scale temperature reconstructions (IPCC 2013, St. George 2014). The latter are used to estimate temperature variability at 4 continental (Euro-Med 2k consortium 2015, Pages 2k consortium 2013, Trouet et al. 2013), hemispheric 5 (Christiansen and Ljungqvist 2012, D'Arrigo et al. 2006, Esper et al. 2002a, Ljungqvist 2010, Ljungqvist et al. 2012, Mann et al. 2008, Schneider et al. 2015, Shi et al. 2013) and global scales (Mann and Jones 6 7 2003) over the past 1000 years, enabling comparisons between climate variations during pre-industrial 8 and industrial periods. The importance of TRCs in these reconstructions arises from the precise annual 9 dating inherent to this proxy (Douglass 1941) and our mechanistic understanding of the influence of 10 temperature on tree growth (Fritts 1976). The relative significance of tree-ring series increases back in time, as the overall number of annually resolved proxies rapidly declines towards the early centuries of 11 12 the past millennium (Esper et al. 2004). 13

14 TRCs are typically composed of tree-ring width (TRW) or maximum latewood density (MXD) 15 measurement series from many trees (Fritts 1976). A TRC might extend back over the entire past 16 millennium if one or more individual trees is 1000 years or more in age. Such longevity, however, is 17 rare and restricted to only a few locations (OldList at: www.rmtrr.org/oldlist.htm). Most millennium-18 length TRCs are therefore produced by combining samples from living trees with older material from 19 archeological and historical structures (hereafter: historical samples), dead wood on the ground (remnant 20 samples), or wood buried under ground or preserved in lakes (sub-fossil samples). The successful 21 combination of samples from living trees with historical/remnant/sub-fossil samples improves when the 22 provenance of all samples is ecologically consistent. If not, sections of a millennium-length chronology 23 can have different growth rates and climate signals than those sections dominated by samples from 24 living trees (Tegel et al. 2010, Linderholm et al. 2014). For example, remnant samples from a sub-alpine 25 site in the Alps are ideally combined with samples from living trees growing on the same slope, at the 26 same elevation and aspect (Salzer et al. 2014b). Sub-fossil trees from a shallow lake are ideally 27 combined with information from living trees surrounding the lake (Düthorn et al. 2013, 2015). 28

29 Combining samples from living trees with historical/remnant/sub-fossil samples is not always 30 straightforward. Habitat homogeneity in a TRC derived from living trees and in-situ remnant or sub-31 fossil wood from the same location may be high, but their combination with historical material can be 32 more complicated. If, for example, the historical samples were obtained from an old building in a 33 mountain valley, it often remains unclear from which position in the surrounding forests the samples 34 were originally collected (Büntgen et al. 2006b). It is not uncommon that historical structures, 35 particularly in alpine environments, contain recycled material of unknown origin, as a consequence of 36 repairs and additions (Bellwald 2000, Kalbermatten and Kalbermatten 1997). Without detailed 37 construction histories the researchers ability to trace the origin of samples is limited (Büntgen et al. 38 2005, Wilson et al. 2004). The situation is further complicated if the samples combined in a TRC are 39 from multiple locations spread over a large region, and if this region extends over several hundreds of 40 kilometers. These problems, affecting the *Homogeneity* of a tree-ring dataset, are seemingly reduced in 41 chronologies from only living trees, sampled at a single site.

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43 Another important characteristic of millennium-length TRCs includes the number and temporal 44 distribution of TRW (or MXD) measurement series averaged in the mean chronology. Varying sample replication is often reported when describing a new TRC, but is usually disregarded in large-scale 45 46 temperature reconstructions. Typically, the number of measurement series included in a TRC declines 47 back in time, and might change from more than 100 living-tree samples in the 20th century to only a 48 handful of samples (perhaps from a single historical structure) at the beginning of the last millennium. 49 Acknowledging the effects of changing sample size by calculating temporally varying uncertainty 50 estimates, is not usually considered outside the tree-ring community (IPCC 2013). However, this 51 characteristic is important as the relevance an individual TRC in large-scale proxy networks is 52 commonly based on the strength of instrumental calibration of only the well-replicated 20th century 53 data, thereby overlooking any pre-instrumental replication changes.

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55 Similarly, the coherence among the TRW (MXD) series combined in a TRC, and temporal change 56 thereof, is not considered in the non-dendrochronological literature (Frank et al 2007). The inter-series

1 correlation among TRW series is an important characteristic of a mean chronology, and is commonly 2 3 computed to evaluate the temporal changes in the chronology's signal strength (Fritts 1976). It is rarely stable and can change, for example, at (i) the transition from living trees to series from 4 historical/remnant/sub-fossil material, or (ii) from a cluster of measurement series of a certain building 5 to another building, or (iii) by the proportion of juvenile, mature, and adult growth rings (Cook and 6 Kairiukstis 1990). Gradual trends in inter-series correlation, as well as step changes, are common in 7 long TRCs and bare important information on the reliability of dendroclimatic reconstructions during 8 pre-instrumental periods. Measures that assess the affect of changing sample size and inter-series 9 correlation include the Expressed Population Signal and Subsample Signal Strength (Wigley et al. 10 1984). However, these metrics are not widely recognized beyond the tree-ring community, and their combination with other uncertainties, e.g. from the unexplained variance of the calibration model or the 11 12 choice of the detrending model, remains challenging (Esper et al. 2007).

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14 Another important TRC characteristic is the degree to which a chronology retains the full spectrum of 15 pre-instrumental temperature variance, which is affected by the method used in chronology 16 development, and the age-structure of the underlying data (Cook et al. 1995). Recent assessments of 17 large datasets showed that instrumental meteorological measurements and tree-ring timeseries contain 18 different frequency spectra (Bunde et al. 2013, Büntgen et al. 2015, Franke et al. 2013, Zhang et al. 19 2015), and that TRCs are limited in capturing millennial scale temperature trends (Esper et al. 2012b). 20 To minimize the loss of long-term information, dendrochronologists apply detrending techniques that 21 are specifically designed to preserve low frequency variance. The preferred approach is the Regional 22 Curve Standardization (RCS) method, introduced to dendroclimatology by Briffa et al. (1992). 23 However, RCS demands a large number of TRW (MXD) measurement series, and requires the 24 underlying data to represent a combination of short segments (trees) distributed more or less evenly 25 throughout the entire chronology (Esper et al. 2003a). For example, if a TRC is composed of only very 26 old living trees, the chronology's biological age will steadily increases towards the present. This causes 27 the biologically younger rings to be concentrated at the beginning of the past millennium, and the older 28 rings in the modern period. This age structure limits the comparison of tree-rings of the same age over 29 time, which is the backbone of RCS and related tree-ring detrending techniques (Melvin and Briffa 30 2008). 31

32 The basic characteristics of TRC Homogeneity, Replication, Growth Coherence, and Chronology 33 Development are well known to dendroclimatologists. However, they are not usually recognized in the 34 multi-proxy paleoclimate community and rarely, if ever, considered in large-scale temperature 35 reconstructions derived from these data (IPCC 2013, Pages 2k consortium 2013). Promoting these 36 characteristics, and stimulating their consideration in addition to the classical calibration against 37 instrumental climate data, is the main objective of this article. We review the characteristics of a total 38 of 39 published temperature reconstructions reaching back to AD 1000 and use the results to rank the 39 timeseries.

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#### 42 **2. Data and Methods**

#### 43 2.1 Temperature reconstructions

44 A survey of tree-ring based temperature reconstructions, reaching back with a minimum replication of 45 three TRW (or MXD) measurement series to AD 1000, returned 39 records (Table 1). Fourteen records 46 are from Asia, 13 from Europe, 8 from North America, and 4 from the Southern Hemisphere (SH). The 47 reconstructions are not evenly distributed over the hemispheres, but are clustered in Fennoscandia, the 48 European Alps, northern Siberia, high Asia, the Rocky Mountains, southwestern US, southern South 49 America, and Tasmania/New Zealand. The TRCs are located in regions characterized by different 50 summer warming trends over the past 100 years (see the colored areas in Fig. 1). Compared to the 51 Northern Hemisphere (NH), the SH is clearly underrepresented with only four records.

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53 There are precedents of long TRCs with an inferred climate signal (e.g. LaMarche 1973, 1974), but the 54 first instrumentally calibrated, millennium-length record is the summer (previous-year December to 55 current-year February; pDec-Feb) temperature reconstruction from very old (living) *Fitzroya* 56 *cupressoides* growing in the Rio Alerce valley in southern Argentina (Villalba 1990; Table 1). Other records developed at that time were later updated by including new measurement series and/or reprocessed using new methods. A good example is the Torneträsk MXD chronology that was originally developed in the 1980s (Schweingruber et al. 1988), calibrated and reprocessed in the early 1990s (Briffa et al. 1990, 1992), updated in the early 2000s (Grudd 2008), and recently again updated and reprocessed (Melvin et al. 2013). In those instances where there are multiple versions of a reconstruction, we cite the most recently published account as it contains references to all previous work.

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8 The millennium-length temperature reconstructions are derived from various conifer species 9 representing nine genera, with Pinus (n=14 records) and Larix (n=12) being most common. Seven 10 reconstructions, including the early *Fitzrova cupressoides* record from Argentina (Villalba 1990), are produced from only living trees, whereas the majority of chronologies (n=32) are composed of tree-ring 11 12 series from living trees combined with series from historical samples (e.g. Lötschental TRC from 13 Switzerland; Büntgen et al. 2006a), remnant samples (e.g. Polar Ural TRC from Russia; Briffa et al. 14 2013), and sub-fossil samples (e.g. Oroko Swamp TRC in New Zealand, Cook et al. 2002). Some of 15 these chronologies are composed of samples collected in well-constrained, ecologically homogeneous 16 (Schweingruber 1996) sites (e.g. Dzehlo in Russia, Myglan et al. 2012b), whereas others combine data 17 from different sites (e.g. Yamal in Russia, Briffa et al. 2013), and even from several valleys within a 18 larger region (e.g. Karakorum in Pakistan; Esper et al. 2002b).

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20 All TRCs included in this survey have either been calibrated against instrumental climate data and 21 transferred into temperature units, or interpreted by the original authors as a temperature proxy. 22 Interestingly, the different methods used to transfer TRW and MXD data into temperature units (Briffa 23 et al. 1983, Cook et al. 1994, Esper et al. 2005) resulted in vastly different reconstructed temperature 24 ranges, varying by only a few tenths to several degrees Celsius over the past millennium (see the varying 25 ranges of the thin black curves in Figure 2; see Esper et al. 2012a for a regional example). Also, the 26 season of maximum response to temperature (e.g. June-August, May-September, etc.) and the 27 reconstructed climate target (e.g. mean, maximum, and minimum temperature) differ among the records 28 (last column in Table 1).

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30 Surprisingly, despite these differences in *(i)* location and regional 20th century temperature trends (Fig. 31 1), (ii) species composition and sample sources (historical/remnant/sub-fossil), (iii) seasonality of the 32 temperature signal, and *(iv)* transfer technique and reconstructed variance; the simple arithmetic mean 33 of each "continent" (acknowledging that the records do not spatially represent NH continents) coheres 34 astonishingly well over the past 1000 years (Fig. 2e). Correlations range from r = 0.42 between Asia 35 and North America to r = 0.48 between Europe and Asia, and increase at decadal resolution to 0.66 36 (Asia/N-America) and 0.82 (Europe/Asia). This large-scale coherence indicates that some common 37 external forcing affects this dendrochronological network (Fernández-Donado et al. 2013, Pages 2K 38 PMIP3 group 2015) and confirms the paleoclimatic significance of tree-ring data over the past 39 millennium.

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# 41 2.2 TRC characteristics and metrics

42 In this section, we describe the five basic TRC characteristics Data Homogeneity (2.2.1), Sample 43 Replication (2.2.2), Growth Coherence (2.2.3), Chronology Development (2.2.4), and Climate Signal 44 (2.2.5), commonly used by dendrochronologists to evaluate a chronology for climate reconstruction, 45 and explain how statistical metrics of these characteristics are used in an ordinal scoring scheme that is 46 understandable to non-specialists. In those instances when raw TRW and MXD data were publically 47 available or contributed by the authors (raw data at: www.url-here.se), we have re-calculated the metrics 48 of interest. When the original cross-dated measurements were not available (see last column in Table 49 7), we have estimated their scores based on information provided in the original articles. Such estimates 50 are highlighted in red in the tables that follow. The calibration scores, resulting from the TRC's correlation against temperature data (2.2.5 *Climate signal*), were taken from the original articles. In the 51 52 instances where no measure of calibration was detailed in the original article, we used nearby gridded 53 data to provide an estimate of climate calibration.

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For each characteristic (2.2.1 to 2.2.5) we used the ordinal scoring scheme to rank the reconstructions. To aid reconstruction comparison, results of the TRC scores are stratified into four groups: class-A (highlighted in green in Tables 2-7), class-B (light green), class-C (light blue), and class-D (blue).
Except for the first characteristic (2.2.1 *Data Homogeneity*), we highlighted the ten top-ranked TRCs in green (ranks 1-10), the TRCs ranking 11-20 in light green, the TRCs ranking 21-30 in light blue, and the TRCs ranking 31-39 in blue. This systematic was changed with *Data Homogeneity* (5 green, 9 light green, 16 light blue, 9 blue) to account for the larger number intermediate TRCs. The individual ranks for each characteristic (2.2.1 to 2.2.5) were finally summed into an overall score.

8 2.2.1 Data Homogeneity

9 Of the five characteristics introduced here, Data Homogeneity is the most descriptive, i.e. it is based on 10 a combination of qualitative traits rather than quantitative measures. Homogeneity integrates information on the (i) source of wood samples, (ii) type of chronology, (iii) number of species, (iv) 11 12 temporal clustering, and (v) a remark (results shown in Table 2). "Source" includes information on the origin of wood samples and the number of sampling sites. We use "Sub-fossil" for samples from lakes, 13 14 bogs, etc., "Remnant" for dead wood on the ground, and "Historic" for samples from old buildings 15 archaeological structures. The Homogeneity score also considers whether the samples originate from 16 one, several, or multiple sites, as far as this information could be obtained from the original publication 17 or via personal communication with the authors. "Chronology type" differentiates between "C" for 18 records composed of living plus relict (sub-fossil/remnant/historical) material, and "L" for records 19 composed of samples from only living trees. The "Number of Species" in a TRC is typically one, but 20 occasionally may be two. "Temporal clustering" refers to cases where the contribution of data from 21 distinct homogeneous sites dominates specific periods of the past 1000 years (a condition that might 22 require the application of multiple RCS runs, Melvin et al. 2013). Finally, we included a "Remark" 23 section summarizing specific features that are relevant to the Homogeneity score in support of the 24 reconstruction's ranking.

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#### 26 2.2.2 Sample Replication

27 The temporal distribution of TRW (or MXD) measurement series in the reconstructions differs 28 dramatically over the past millennium (Fig. 3). These changes are considered in the second metric by 29 combining information on (i) mean replication, (ii) maximum replication, (iii) minimum replication, and 30 (iv) the 11th/20th century ratio of measurement numbers. "Mean replication" is the average number of 31 measurement series (core samples or radii from disks) over the last millennium, considering all years 32 from AD 1000 to the most recent year of a reconstruction. "Maximum replication" and "Minimum 33 replication" refer to the maximum and minimum numbers of measurement series, which are typically 34 reached in the modern and the early periods of a reconstruction, respectively (see the black curves in 35 Fig. 3). The "11th/20th century ratio" acknowledges this exemplary replication curve shape, as well as 36 its significance in the reconstruction: all TRCs are calibrated over the well-replicated 20th century, but 37 the reconstruction period extends back to the, often weakly replicated, 11th century. The metric equals 38 the mean 11th century replication, divided by the mean 20th century replication, multiplied by 100. To 39 produce the final *Replication* score, the first three values are summed (i+ii+iii), and the resulting sum 40 multiplied by (*iv*). If the reconstruction is produced using MXD data the *Replication* score ((*i+ii+iii*)\**iv*) 41 is multiplied by 2 to account for MXD's increased signal strength and higher production costs. Note that 42 these choices, as well as those described below for the other TRC characteristics, are not statistically 43 validated, but made with the intention of combining descriptive measures commonly used in 44 dendrochronology into an ordinary scoring system that can be used to compare and rank reconstructions.

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# 46 *2.2.3 Growth Coherence*

47 Another important characteristic influencing the temporally changing skill of tree-ring based climate 48 reconstructions, is the correlation between the TRW (MXD) measurement series (Frank et al. 2007, 49 Osborn et al. 1997, Wigley et al. 1984). For those reconstructions where the raw data are available, we 50 calculated the inter-series correlation (abbreviated "Rbar" in the dendrochronological literature; Cook 51 and Kairiukstis 1990) for 100-year segments, sliding in 10-year steps along the chronology (Fig. 4). The 52 resulting timeseries reveal substantial differences among the reconstructions (the black curves in Fig. 53 4), as well as a minor tendency towards reduced values back in time, particularly in some of the records 54 from Europe and Asia. These characteristics are considered here in the Growth Coherence score by 55 summing the (i) average inter-series correlation over the past millennium (mean Rbar), (ii) maximum 56 inter-series correlation in a single 100-year period (max. Rbar), and (iii) minimum inter-series

correlation in a single 100-year segment (min. Rbar). The sum (*i*+*ii*+*iii*) is multiplied by (*iv*) the
11th/20th century Rbar ratio (in %).

# 4 2.2.4 Chronology development

5 A key component in the process of building a TRC is the detrending method used to remove tree-age 6 related growth trends from the raw measurement series (Bräker 1981, Cook and Kairiukstis 1990, Cook 7 et al. 1995). As detailed above, application of RCS (Esper et al. 2003a) is currently accepted as the 8 preferred method to preserve low frequency variance in TRCs. We acknowledge this view by (i) assigning TRCs produced using RCS a "1", and TRCs produced using individual-series detrending 9 methods (e.g. ratios from negative exponential curves or smoothing splines) a "2" (Cook and Peters 10 1997). However, RCS only works well if the underlying measurement series are derived from a 11 12 composite of living and relict trees, ideally including young and old tree-rings evenly distributed 13 throughout the past millennium (Esper et al. 2014). TRCs that are composed this way are characterized by age curves that are nearly horizontal over the past 1000 years (Fig. 5). In practice this is rarely the 14 15 case. The age curves of some TRCs composed of very old living trees in, for example, North America 16 and Asia are particularly steep (increasing-age towards present). In contrast, in Europe, where the 17 majority of reconstructions are derived from composite chronologies of historical and living-tree 18 samples, the mean age curves are relatively flat (the blue curve in Fig. 5e).

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We score these attributes by considering *(ii)* the maximum difference between the highest and lowest value in the age curve over the past millennium, and *(iii)* the slope of a linear regression fit to the age curve. We further consider *(iv)* the maximum retained low frequency information, ranging from multicentennial = 1, to centennial = 2, to decadal = 3. For the final *Chronology Development* score we multiply *(i)* the method score (1 for RCS, 2 for individual-series detrending), with *(ii)* the (square root of the) max.-min. age difference, *(iii)* the (absolute) slope of the linear regression (times 100), and *(iv)* the maximum retained low frequency score (1 to 3, for multi-centennial, centennial, and decadal).

28 2.2.5 Climate signal

29 This final score considers some of the classic metrics used in paleoclimatic research, such as the 30 correlation against monthly instrumental temperature data, averaged over the season of maximum 31 response (see the last column in Table 1). However, as the period of overlap between instrumental and 32 proxy data varies considerably among the reconstructions – largely due to the lengths of observational 33 data available to researchers – we score *Climate Signal* by (i) the square root of the number of years of 34 overlap between the TRC and instrumental record, multiplied by the residual between, (ii) the 35 correlation against climate data and (iii) a split calibration/verification difference. The latter metric is a 36 standard criterion in dendroclimatology used to benchmark the temporal robustness of the relationship 37 between proxy and instrumental data (Cook and Kairiukstis 1990). However, the split 38 calibration/verification differences are not always reported. In those instances, we estimated the split 39 calibration/verification difference based on our calculations using gridded temperature data. Finally, we 40 include an additional adjustment (iv) to account for a calibration period that was intentionally shortened 41 to avoid potential divergence issues (for details see Büntgen et al. 2008, D'Arrigo et al. 2008, Esper and 42 Frank 2009, Esper et al. 2010, Wilson et al. 2007). If such problems are reported in the original article, 43 and the calibration period was truncated, we used 0.5 as a multiplier (1 if no such problem was detected). 44 The final *Climate Signal* score was derived by: square root *i* \* (*ii* - *iii*) \* *iv*.

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# 47 **3. Results and Discussion**

# 48 3.1 Overall TRC ranking

Our assessment of 39 millennial-length TRCs' *Homogeneity*, *Replication*, *Growth Coherence*, *Chronology Development*, and *Climate Signal* is presented in Tables 2-6. The final ranking (Table 7), derived from the sum of these metrics, reveals that no reconstruction consistently dominates in the top group (class-A, dark green dots in the tables) in all five categories. Four records (N-Scan, E-Canada, Finland, Dzhelo) score high (class-A or class-B) in four out of the five categories, and one record (Yamal) scores high in three. However, each of these, overall best-ranked reconstructions, scores less

- well (class-C: light blue dot) on at least one criterion, mostly *Homogeneity* (four records).
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There are ten records (W-Himalaya, Tatra, Karakorum, Great Basin, S-Finland, Tien Shan, Jämtland, Wulan, Gulf of Alaska, French Alps) with weak scores (class-C or class-D) in four metrics. W-Himalaya is the only reconstruction scoring in class-D in four: *Homogeneity, Replication, Growth Coherence*, and *Chronology Development*. This low scoring, combined with the class-B rank in *Climate Signal*, places the W-Himalaya reconstruction at the bottom of table 7. However, the raw data are not available for this reconstruction (see the last column) and several scores had to be estimated (highlighted in red in tables 3-6). The same is true for several other reconstructions, and it seems advisable to emphasize the coarse categorization into four classes (A to D) rather than the precise ordering in our tables.

## 10 *3.2 Detailed TRC rankings*

Four reconstructions (Dzehlo, Tasmania, Rio Alerce, Oamdo, Mongolia) rank at the top in Homogeneity 11 12 (Table 2). The data used in these TRCs include samples from living trees, as well as remnant and sub-13 fossil material from a single site or valley (with one exception; Rio Alerce includes only living trees). 14 These top-ranked records are followed by a group of nine reconstructions that were sampled from 15 slightly less homogeneous conditions, including data at moist and dry micro-sites, from different 16 elevations, and measured using different techniques, for example. Despite their less than ideal 17 Homogeneity score, these reconstructions are still more homogenous compared to a number of TRCs 18 (n=16) that integrate data from multiple sites in different valleys, regions, and/or elevations (light blue 19 dots in Table 2). Such large differences in habitat can introduce substantial growth rate variations that 20 are difficult to differentiate from long-term temperature variations. In addition, the climate signal might 21 change between samples from different elevations and micro-sites. These potential biases are likely 22 most severe in the nine TRCs ranking last (dark blue dots in Table 2). The two end members (Karakorum 23 and W-Himalaya) are produced using living trees only, sampled from multiple sites, located in different 24 valleys and at different elevations, with distances up to 100 km between sites, and including two tree 25 species (Karakorum). Clearly these TRCs contain a less homogeneous sample composition compared 26 to the top-ranked records that include samples from only one, well-constrained site.

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28 The reconstructions scoring well in *Homogeneity* are not necessarily top-ranked in *Replication* (Table 29 3). To appear in the top group in *Replication*, it is necessary not only to include a large number of TRW 30 or MXD measurement series, but have these samples evenly distributed throughout the past millennium. 31 Bumps from very high to very low replications in certain periods, as well as large differences between 32 20th and 11th century replications, result in a lower score. Among the records doing well in *Replication* 33 are two TRCs from Central Asia (Mongun and Dzhelo) and one from New Zealand (Oroko Swamp). 34 These records score particularly well in the 11th/20th century ratio, reaching values >100%. Other 35 reconstructions, such as the Alps (larch) and Swiss/Austrian Alps TRCs include many samples (530 and 36 253 respectively over the past millennium), but contain a dramatic replication decline from the 20th to 37 the 11th centuries (down to 2% and 25%, respectively), limiting the skill of these timeseries in the early 38 period of the past millennium. The TRCs scoring weakest in Replication (Tatra, Boreal Plateau, Rio 39 Alerce, Tien Shan) are characterized by low minimum replications ( $n \le 5$  series) and small 11th/20th 40 century ratios  $\leq 15\%$ ). These records might perform well when calibrated against 20th century 41 instrumental temperature data, but there is considerable risk that this 20th century skill does not persist 42 over the past millennium, simply because the number of samples changes dramatically back in time.

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44 Since more than the sheer number of measurement series is important, we also considered the reconstructions' inter-series correlations (Table 4). The three TRCs scoring best in this category 45 46 (Indigirka, Yamal, Taimyr) are all located in northern Siberia, where growth variations among trees are 47 synchronized by harsh climatic conditions during a rather short growing season. These top-ranked 48 records are characterized by inter-series correlations that do not fall below Rbar=0.20 at any time over 49 the past millennium (minimum correlation in Table 4) and reach values >100% in their 11th/20th century 50 ratio. Other mid-ranked TRCs, such as Polar Ural (class-B) and Jämtland (class-C), display either a very 51 low minimum Rbar values (-0.20 in Polar Ural) or substantially decreasing Rbar values from the 20th 52 century back to the 11th century (42% in Jämtland). Another interesting example of a class-C TRC is 53 Oroko Swamp, which is characterized by only minor Rbar changes back in time (92%), but an overall 54 low mean inter-series correlation (Rbar=0.18). Finally, the TRCs scoring weakest (Tatra, S-Finland, 55 Central Alps) are characterized by severe correlation declines, down to  $\leq 20\%$  back in the 11th century, 56 and either a low mean Rbar values (0.20 estimated for S-Finland and Central Alps) or negative minimum Rbar values (-0.07 in Tatra). In these cases it seems advisable to anticipate substantial changes in the chronologies' signal strength over the past millennium, as the coherence among their constituent measurement series is extremely variable. If the inter-series correlation drops significantly, reductions in TRC variance, and a tendency towards the long-term mean are to be expected.

5

6 The three top-ranked reconstructions in the *Chronology Development* category are all from Northern 7 Europe (N-Scan, Finland, Torneträsk (TRW)), followed by records from the Alps (Lötschental) and 8 Canada (E-Canada) (Table 5). These reconstructions, as well as the other class-A and class-B TRCs 9 (green and light green in Table 5, total n=20), are all composed of a mixture of living trees and 10 historical/remnant/sub-fossil samples, facilitating the application of RCS for optimal conservation of low frequency variance (Autin et al. 2015, Briffa et al. 1992, Esper et al. 2003a). The top-scoring 11 12 Northern European records are, however, additionally characterized by small age ranges (<110 years) 13 and only minor (positive and negative) linear trends in the mean age curves. The top-ranked N-Scan 14 record is reported to contain millennial scale temperature variance (Esper et al. 2012b), a feature also 15 seen in the Taimyr reconstruction from Northern Siberia. The subsequent mid-ranked TRCs are 16 characterized by age ranges from  $\sim$ 150-300 years, as well as linear trend angles ranging from  $\sim$ 5-30 17 degrees. Some class-C records were standardized using individual detrending methods, including the 18 Swiss/Austrian Alps, Lauenen, and Mongolia reconstructions, an approach more commonly found in 19 the TRCs towards the bottom of table 5. The application of individual detrending methods has been 20 shown to systematically limit the low frequency variance retained in TRCs (Cook et al. 1995). This 21 limitation is reflected in the maximum frequency metric included here, indicating that six 22 reconstructions (Rio Alerce, Wulan, Gulf of Alaska, Mongun, S-Chile, Lauenen) maximally retain 23 decadal scale temperature variance. These records, as well as some of the individually detrended TRCs, 24 should not be used with the objective of reconstructing the full spectrum of temperature variance over 25 the past millennium (e.g. Mann et al. 2008).

26

27 By comparison to Homogeneity, Replication, Growth Coherence, and Chronology Development, 28 measures of climate signal strength are widely recognized in the paleoclimatic community. However, a 29 good correlation between tree-ring proxy and instrumental temperature data alone is a fairly incomplete 30 description of reconstruction skill. For example, if a TRC includes many more samples during the 20th 31 century (*Replication* metric), or the samples originate from different valleys (*Homogeneity*), or the mean 32 age curve declines severely back in time (Chronology Development), the 20th century calibration 33 statistics provide little information about the signal strength over past centuries. That being said, we 34 here assess climate signal strength based on the length of the calibration period, the correlation strength 35 with instrumental data, the calibration/verification difference and any, seemingly arbitrary, truncation 36 of the calibration period.

37

38 The reconstructions scoring best for *Climate Signal* are all from regions where sufficient instrumental 39 data are available to calibrate over periods of 100 years and longer (Table 6). The three top-ranked 40 records (Torneträsk (MXD), N-Scan, Alps (larch)) all correlate at ≥ 0.70 against instrumental temperature data, with only minor differences (<0.10) between calibration and verification periods. 41 42 Other reconstructions, with calibration period correlations ≥ 0.70, albeit over shorter periods (53 years 43 in Oamdo, 57 years in Taimyr), contain larger calibration/verification differences (0.18 in Taimyr) and 44 appear in class-B. These reconstructions certainly meet the criteria for a successful TRC calibration, but 45 they may contain a marginally verifiable climate signal. This is either because the 46 calibration/verification differences are large (e.g. 0.63 in Qilian), the calibration period was truncated 47 due to some inconsistency (e.g. Tatra, see the fourth column in Table 6), or the overall correlation is 48 low (e.g. 0.17 in Upper Wright Lakes). However, a weak calibration result does not necessarily mean 49 that a TRC contains no climate signal, but might indicate that the instrumental station record is too short 50 (Esper et al. 2010), of poor quality (Böhm et al. 2001, 2010, Parker et al. 1994), or too remote (Cook et 51 al. 2013).

52

Perhaps a good example, highlighting the importance of using several categories to evaluate a TRC, is
 the case of the Alps (larch) record. The Alps TRC correlates well (r=0.70) over 140 years of regional
 instrumental temperatures, and thus ranks #3 in the *Climate Signal* metric (calibration/verification

56 difference is 0.07, calibration period not truncated). However, these calibration statistics were obtained

over the period 1864-2003 during which the TRC's mean replication is 1379 series. Concurrently, the average number of TRW series in the 11th century reaches only 22, which produces an 11th/20th century ratio of 2% (see Table 3). Though certainly an extreme example, it nicely demonstrates how a largescale reconstruction produced focusing on robust 20th century climate signals, can result in an overestimation of statistical skill over the past millennium.

# 7 *3.3 Ranking implications*

8 Over the past decades a number of statistically valid methods have been developed to describe TRC's 9 signal strength. Examples include the *Expressed Population Signal* (Wigley et al. 1984), bootstrap 10 confidence intervals (Briffa et al. 1992), ensemble calibration technique (Frank et al. 2010), and reduced 11 sample calibration trials (Esper et al. 2012b). All of these dendro-specific statistics help estimate the 12 temporally varying skill of tree-ring based climate reconstructions, but the methods are largely 13 inapplicable to other proxy archives, and are not used in large-scale, multi-proxy reconstructions (Pages 14 2k consortium 2013).

15

16 By providing an assessment and ranking of the TRCs, we attempt to bridge the gap between the tree-17 ring, modeling, and multi-proxy communities. While some of the scores and metrics used here have not 18 been rigorously validated, we believe that the development of an intuitive ranking system that can be 19 universally applied to all TRCs will foster the judicious use of tree-ring data in large-scale 20 reconstructions. For example, if NH temperature variability during medieval times is of interest, it is not 21 meaningful to include TRCs with only a few samples during the 11th century, i.e. researchers might 22 want to avoid reconstructions with low *Replication* scores (Table 3). Similarly, if the full spectrum of 23 past temperature variability is of interest, one might want to include only those TRCs retaining 24 centennial to millennial scale variance, i.e. exclude records with low *Chronology Development* scores 25 (Table 5).

26

27 These arguments lead to a list of recommendations:

- R1 Avoid integrating TRCs that emphasize decadal scale variance (scoring low on "Maximum frequency") when intending to reconstruct low frequency (centennial to millennial scale)
   temperature variance.
- R2 Avoid overrating TRCs that average many TRW (MXD) measurement series in the 20th century,
   but only few series at the start.
- R3 Pay attention to the sample composition of millennium-length TRCs. Do data sources change
   through time (different sites, buildings, valleys, etc.)?
- R4 Consider TRC replication and Rbar changes when interpreting tree-ring based climate
   reconstructions.
- R5 Differentiate between composite TRCs that integrate data from varying sources
   (living/remnant/historical/sub-fossil) and TRCs that integrate data from only old living trees, and
   acknowledge potential biases due to changing tree ages over the past millennium.
- R6 Do not only focus on the calibration statistics from comparisons with instrumental climate data, as
   this perspective can give the false impression that reconstruction skill persists throughout the past
   millennium.

44 We acknowledge that some of the metrics presented here contain partly redundant information, e.g. 45 lower replication or reduced Rbar values typically result in weaker correlations with instrumental 46 climate. There are also other TRC characteristics that could be used to assess tree-ring based temperature 47 reconstructions, though these appeared difficult to quantify with simple metrics. Examples include the 48 TRC serial correlation (Meko 1981) and climate signal after trend removal (von Storch et al. 2004). For 49 instance, an assessment of serial correlation in both tree-ring and instrumental temperature data might 50 reveal a larger lag-1 autocorrelation in a TRC (likely due to biological memory effects; Esper et al. 2015), potentially indicating a coherence deficiency and reduced skill of a long-term climate 51 52 reconstruction. Similarly, an assessment of the climate signal after removing low frequency variance 53 (e.g. increasing 20th century temperature trend), from the instrumental and proxy data, increases the 54 degrees of freedom of the calibration statistics and supports the estimation of signal strength in the high 55 frequency domain. However, correctly evaluating these properties in a large network of millenniumlength TRCs, including several records for which the underlying measurement data are not available, is
not feasible.

4 Our review clearly indicates that solely focusing on the calibration statistics overlooks a number of 5 additional, important characteristics inherent to tree-ring based climate reconstructions. When 6 evaluating large TRC networks it is important to keep in mind that the 20th century instrumental data 7 (i) contain gaps, breakpoints, and biases (Hinkel et al. 2003, Landsberg 1981, Oke 2007), (ii) are of 8 substantially varying length depending on the study region (e.g. in Europe versus central Asia; Cook et 9 al. 2013), and (iii) are recorded at greatly differing distances from the tree-ring sampling sites. The 10 suitability of a station record is additionally influenced by the topography (flat or mountainous), the elevational difference between tree and station sites, and regional synoptic weather patterns. The use of 11 12 gridded climate data does not necessarily overcome these shortcomings as they rely on the same (Jones 13 et al. 1999) or even fewer (Krusic et al. 2015) station data.

14

15 Finally, it is important to acknowledge the seasonality of the climate signal (typically related to the 16 regional growing season, Table 1) and question the use of a seasonally restricted reconstruction, as 17 reported in the original literature, in attempting to reconstruct mean annual temperatures (Mann et al. 18 1999, 2008). While we acknowledge that the calibration/verification statistics (Table 6), as well as 19 means by which the chronology is transformed to measures of climate are also important (e.g. Bürger 20 et al. 2006, Christiansen 2011, Christiansen et al. 2009, Esper et al. 2005, Juckes et al. 2007, Lee et al. 21 2008, Smerdon et al. 2011, 2015, Rutherford et al. 2005, von Storch et al. 2004, Zorita et al. 2003), we 22 believe that a revised consideration of basic TRC characteristics as well as the development of new 23 TRCs is key to improving the extensive network of large-scale reconstructions and our understanding 24 of past climate variability.

25 26

# 27 4. Conclusions

28 Thirty-nine millennium-length temperature reconstructions are ranked based on a rating scheme that 29 considers basic TRC characteristics commonly considered by dendrochronologists. The TRC 30 characteristics were grouped into five composite scores: Homogeneity, Replication, Growth Coherence, 31 Chronology Development, and Climate Signal. It is argued that consideration of these characteristics, 32 beyond the tree-ring community, will improve the development of large-scale temperature 33 reconstructions that utilize tree-ring data from different regions and continents. Similarly, the rankings 34 produced for each score will support this objective, as they facilitate the selection process of TRCs when 35 addressing paleoclimatic objectives. For example, researchers might not want to include TRCs resting 36 on only a few trees during the 11th century, in a study addressing the magnitude and spatial extent of 37 warmth during medieval times. This, and other recommendations are expressed towards the end of this 38 review paper.

39

40 A systematic comparison of TRC characteristics permitted ranking the 39 millennium-length temperature reconstructions into four groups (class-A to class-D) for each of the five metrics. No 41 42 reconstruction scores top in all five metrics, but each record has its particular strengths and weaknesses. 43 Nevertheless, there are some reconstructions that overall perform better than others. These include N-44 Scan and Finland from Europe; E-Canada from North America; Yamal and Dzehlo from Asia. 45 Reconstructions performing less well include W-Himalava and Karakorum from Asia; Tatra and S-46 Finland from Europe; and Great Basin from North America. The rankings presented here can be used 47 to select and exclude particular records for producing hemispheric scale reconstructions. The fact that 48 some of the records appear more often towards the end of a ranking table does not mean they cannot be 49 used for climate reconstruction purposes, but indicates that the users of these data need to be aware of 50 potential weaknesses that may inadvertently affect their experiment. This review of millennial-length 51 TRCs will be updated as new reconstructions are produced. Updates will be published online at: 52 www.url-here.se 53

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#### **1 Table and Figure Captions**

Table 1 Millennium-length tree-ring based temperature reconstructions. Superscript \* indicates
 reconstructions developed using MXD (instead of TRW). The Icefield reconstruction contains both

- 4 MXD and TRW data. *Signal* specifies the seasonality of reconstructed temperatures, with *p* indicating
- 5 previous-year months. *T* is temperature, *Tmax* is maximum temperature, *Tmin* is minimum temperature.
- 6 **Table 2** *Homogeneity* scores. Chronology type *C* refers to reconstructions derived from a composite of
- 7 material from living trees, remnant, historical and/or sub-fossil wood. Type *L* refers to reconstructions 8 derived from only living trees. Temporal clustering (*Yes*) indicates records composed of data from
- 8 derived from only living trees. Temporal clustering (*Yes*) indicates records composed of data from
   9 distinct sites or species concentrated in discrete periods over the past 1000 years.
- **Table 3** *Replication* scores. The number of TRW (or MXD) measurement series included in the reconstructions. *11th/20th* is the ratio of the mean replication during the 11th century relative to the
- 12 mean replication during the 20th century. Values in red are estimates.
- Table 4 *Growth Coherence* scores. Mean, maximum, and minimum correlations among the TRW (or
   MXD) series included in the reconstructions. *11th/20th* is the ratio of the correlation during the 11th
   century relative to the 20th century correlation. Values in red are estimates.
- **Table 5** Chronology Development scores. Detrending method 1 = RCS (and Signal Free), and 2 =
- 17 individual detrending. Age range is the difference between highest and lowest mean age curve point 18 over the past millennium. Age trend is the slope of a linear regression fit to the mean age curve over the
- 18 over the past millennium. *Age trend* is the slope of a linear regression fit to the mean age curve over the 19 past millennium (times 100). *Maximum frequency* indicates the wavelength of lowest frequency
- past millennium (times 100). *Maximum frequency* indicates the wavelength of lowest frequency
   information retained in a reconstruction including 1 = multi-centennial, 2 = centennial, and 3 = decadal.
- 21 Values in red are estimates.
- 22 Table 6 Climate Signal scores. Length is the period of overlap with instrumental temperature data in
- 23 years. *Correlation* is the Pearson correlation coefficient between the tree-ring chronology and the
- 24 instrumental data over the calibration period. Calibration/verification difference indicates the
- correlation range between different periods of overlap with instrumental data. *Truncation* = 0.5 if the calibration period was shortened (e.g. due to divergence), *truncation* = 1 if this is not the case. Values in red are estimates.
- Table 7 Ranking of 39 tree-ring based temperature reconstructions based on the *Homogeneity*,
   *Replication, Growth Coherence, Chronology Development, and Climate Signal* scores. Last column
- 30 indicates which datasets are publicly available.
- 31
- Fig. 1 Location of millennium-length tree-ring based temperature reconstructions (circles). Colors
   indicate the June-August temperature change between the mean of the period 1964-2013 minus the mean
   of the period 1914-1963 using GISS 1200 km gridded data.
- **Fig. 2** Tree-ring based temperature reconstructions. Black curves are the 13 reconstructions from Europe
- 36 (a), 14 from Asia (b), 8 from North America (c), and 4 from the Southern Hemisphere (d) shown as
- 37 anomalies from their 20th century means. Note that the reconstructed temperature variance differs 38 substantially among records, largely as a result of the differing calibration schemes used in the original
- 38 substantially among records, largely as a result of the differing calibration schemes used in the original 39 publications. Colored curves are the arithmetic means calculated over the common period of all
- 40 reconstructions in each region. e, Comparison of the mean timeseries from Europe, Asia, and North
- 41 America.
- Fig. 3 TRC replication curves. Black curves show the changing numbers of TRW (or MXD)
  measurement series integrated in the temperature reconstructions from Europe (a), Asia (b), North
- 44 America (c), and the Southern Hemisphere (d). The replication curve of the Alps (larch) reconstruction
- 45 in (a) refers to the right axis. Colored curves are the arithmetic means calculated over the common
- 46 period covered by all reconstructions in each region. **e**, Comparison of the mean curves.
- 47 Fig. 4 TRC inter-series correlations. Black curves show the correlation coefficients among the TRW (or
- 48 MXD) measurement series integrated in the local temperature reconstructions from Europe (a), Asia
- 49 (b), North America (c), and the Southern Hemisphere (d). Correlations are calculated over 100-year
- 50 periods shifted in 10-year steps throughout the past millennium. The earliest value is centered on 1050,
- the most recent value on 1950. Colored curves are the arithmetic means calculated for each region, and dashed lines indicate the mean values over the millennium. **e**, Comparison of the mean inter-series
- 53 correlation curves.
- **Fig. 5** TRC age curves. Black curves show the mean tree age of the TRW (or MXD) data integrated in the temperature reconstructions from Europe (**a**), Asia (**b**), North America (**c**), and the Southern

- Hemisphere (d). Colored curves are the arithmetic means calculated over the common period covered by all reconstructions in each region. e, Comparison of mean replication curves.
- 2