

Review and test of reproducibility of subdecadal resolution palaeoenvironmental reconstructions from microfossil assemblages

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Abstract

Very high, even annual, resolution quantitative palaeoenvironmental reconstructions have been generated from microfossil assemblage data using transfer functions. Evidence of the utility of some of these reconstructions is given by the high correlations between the reconstructions and instrumental records. Such performance is despite several ecological and taphonomic issues that can become more problematic at subdecadal resolution than at multidecadal resolution. This paper explores these timescale dependent issues and tests the reproducibility of the several of published reconstructions. It demonstrates that the previously reported exceptionally good performance of several published chironomid-based reconstructions is not reproducible.

Highlights

- Microfossil assemblage-based palaeoenvironmental reconstructions with subdecadal resolution that are validated against instrumental records are reviewed.
- Ecological and taphonomic processes and chronological uncertainty mean that only in favourable circumstances will precise reconstructions be possible.
- Conditions are favourable for reconstructing recent changes in lake nutrient status with diatoms.
- Conditions are not favourable for subdecadal chironomid-temperature reconstructions.
- Serious problems with several of the subdecadal chironomid-temperature reconstructions are discussed.
- Chrysophyte-based reconstructions are difficult to evaluate due to a dearth of archived data.

Keywords: Microfossils; Quantitative palaeoenvironmental reconstructions; Transfer functions; Reproducibility; Anthropocene; Palaeoclimatology; Palaeolimnology; Global; Data treatment, data analysis

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1. Introduction

Palaeoecologists, amongst others, value high-resolution palaeoecological data (PAGES2k Consortium, 2013; Perga et al., 2015; von Gunten et al., 2012). Of the 17 very high resolution palaeolimnological time series in the PAGES2k Consortium’s (2017) compilation of temperature-sensitive proxy records, most are varve thickness, which can be difficult to interpret (Blass et al., 2007), but two series are derived from microfossil assemblages using transfer functions. Evidence for the utility of these high, even annual, resolution microfossil-based palaeoenvironmental reconstructions comes from comparisons of these reconstructions, typically from varved sediments, with instrumental records. As such records require large investments of time (Green, 1983), it is important to understand the circumstances in which they are likely to yield skilful reconstructions. All validated subdecadal microfossil assemblage based reconstructions identified by a literature search are listed in Table 1. Lotter (1998) and Alefs and Müller (1999) reconstruct slowly varying total phosphorus concentrations from annually-resolved, mainly planktic, lacustrine diatom assemblages. High-resolution chrysophyte stomatocyst stratigraphies have been used to reconstruct winter conditions as reflected by the date of spring mixing (de Jong and Kamenik, 2011) and the duration of cold water (Hernández-Almeida et al., 2015a). Since there are no long instrumental time series of these limnological variables, the reconstructions are correlated with winter air temperatures. Chrysophyte stomatocysts have also been used to reconstruct summer calcium concentrations (Hernández-Almeida et al., 2015b), which might reflect summer mixing and so were correlated with summer wind. High-resolution chironomid head capsule stratigraphies have been used to reconstruct summer air (e.g., Larocque et al., 2009a; Luoto and Ojala, 2016; Zhang et al., 2017a) and water (Medeiros et al., 2012) temperature. High-resolution pollen records have been used to reconstruct summer air temperature (Kamenik et al., 2009).

Table 1: Location, proxy, resolution, and correlation with the instrumental record of the validated reconstructions derived from microfossil stratigraphies. Where two numbers are given for the correlation, the first is for a calibration-in-space model, the second a calibration-in-time model.

	Study	Proxy	Variable	Site	Resolution	Instrumental correlation
3	Lotter (1998)	Diatoms	Total phosphorus	Baldegersee	Annual	“compares well”*
	Alefs and Müller (1999)	Diatoms	Total phosphorus	Ammersee	Annual	0.91*
				Starnberger See	Annual	0.77*
	De Jong and Kamenik (2011)	Chrysophytes	Date of spring mixing	Lake Silvaplana	Annual (3 year smooth)	-0.58 vs Oct.–May air T*
	Hernández-Almeida et al. (2015b)	Chrysophytes	Calcium concentration	Lake Żabińskie	Annual (3 year smooth)	0.5 vs May–Oct. zonal wind velocity*
	Hernández-Almeida et al. (2015a)	Chrysophytes	No. consecutive days with water < 4°C	Lake Żabińskie	Annual (3 year smooth)	0.35 vs Jan.–March air T*
	Larocque and Hall (2003)	Chironomids	July air-T	Lake Njulla	2–7 years	0.39
				Lake 850	2–6 years	0.365
				Alanen	1–5 years	0.35
				Laanijavri		
				Vuoskkujavri	1–6 years	0.37

Study	Proxy	Variable	Site	Resolution	Instrumental correlation
Larocque et al. (2009a), Larocque-Tobler et al. (2011a)	Chironomids	July air-T	Lake Silvaplana	Near annual	0.65/0.53*
Larocque-Tobler et al. (2011b)	Chironomids	July air-T	Seebergsee	3–8 years	0.64/0.73
Larocque-Tobler et al. (2016a, 2015)	Chironomids	August air-T	Lake Żabińskie	1–5 years	0.76*
Medeiros et al. (2012)	Chironomids	Mid-summer water T	Baker Lake	1–8 years	“corresponded well” vs annual air T
Luoto and Ojala (2016)	Chironomids	July air-T	Nurmijärvi	1–11 years	0.38/0.51*
Zhang et al. (2017a)	Chironomids	July air-T	Tiancai Lake	1–4 years	0.45
Lang et al. (2017)	Chironomids	July air-T	Speke Hall Lake	~8 years	0.62
Kamenik et al. (2009)	Pollen	April–November air T	Mauntschas Mire	1–5 years	–/0.66

Several of the correlations shown in Table 1 are surprisingly good given the ecological sensitivity of the proxy, taphonomic issues, and the inherent chronological uncertainty that occurs even in the best-dated palaeoecological records. Some are comparable with the best tree ring-width–climate correlations where the 99th percentile of statistically significant ($p < 0.05$) correlations is 0.7 for winter precipitation and 0.63 for summer temperature (St. George, 2014). Although the tree ring-width–climate relationship has its complexities (D’Arrigo et al., 2008), it is much simpler than the relationship between microfossil assemblages and climate, and usually has no chronological error or taphonomic complications. Section 2 of this paper explores these complications affect subdecadal resolution reconstructions, focusing on the chironomid studies as these are the most numerous of the studies in Table 1, are from diverse settings, and several have exceptional performance. The strongest correlation between chironomid-inferred temperature and the instrumental record is the August air temperature reconstruction from Lake Żabińskie (Larocque-Tobler et al., 2015), so this study is used to exemplify the issues with high-resolution palaeoenvironmental reconstructions.

Given that the exceptional performances shown in Table 1 are used to justify the development of millennium-long temperature reconstructions (Hernández-Almeida et al., 2017; Larocque-Tobler et al., 2010, 2012), which have been used in regional climate compilations (Trachsel et al., 2012), and could be used in hemispheric or global compilations with policy relevance (Masson-Delmotte et al., 2013), it is critical to ascertain that the results are reproducible. Pre-publication peer-review is not a sufficient guarantee of reproducibility, not least because the raw data are rarely made available to reviewers. Recent replication studies in other disciplines have shown low levels of reproducibility (Fidler et al., 2017). Full direct replication of palaeoecological studies would be costly and time consuming, but re-analysis of archived data is an important alternative for testing for reproducibility (Fidler et al., 2017). Fidler et al. (2017) argue for a systematic evaluation of the reproducibility of the evidence base in ecology. Papers with surprisingly good performance are an obvious target for evaluation. Section 3 of this paper attempts to reproduce the chironomid studies from Table 1, for which some data (or digitised data) are available for all the studies.

2. Challenges for subdecadal resolution reconstructions

This section explores the ecological, chronological, and taphonomic challenges faced by subdecadal resolution reconstructions, with a focus on chironomid inferred air temperature reconstructions, and considers other reconstructions when their attributes differ in important ways. Many of these challenges discussed below will also affect sedimentary geochemical proxies to some extent.

2.1. Ecological sensitivity

With the exception of the diatom-total phosphorus reconstructions, all the work cited in Table 1 attempts to reconstruct climate for part of a year, most

often air temperature from a single month in summer. It is possible to conceive of a biological proxy that is sensitive to, and only to, a single month's temperature. Such a proxy would need to have a short generation time, so it can rapidly respond to temperatures in that month, and to live mainly within that month and be dormant throughout the remainder of the year. It is not clear that any proxy meets these ideal standards, but bloom-forming algae such as chrysophytes might, especially in lakes with a short ice-free period. Chironomids would seem to fall far short of these ideals, with the larvae taking months, or even years, to mature (Tokeshi, 1995).

Larocque-Tobler et al. (2015) reconstruct August air temperature because this variable explains the most variance in a Canadian calibration set (Larocque et al., 2006); several other papers reconstruct July air temperature (Lang et al., 2017; Larocque and Hall, 2003; Luoto and Ojala, 2016). While it is widely believed (Eggermont and Heiri, 2011) that summer air temperature is a key variable affecting chironomid assemblages, there is no particular reason to suspect July or August temperatures to be uniquely important. August is not, for example, the warmest month in Lake Żabińskie (Bonk et al., 2015a). The phenology of chironomid emergence (Tokeshi, 1995) suggests that many chironomids will have emerged before August, precluding August temperatures of their assigned year from having any influence, a problem recognised by Larocque-Tobler et al. (2012). At Polish latitudes, many of the chironomids in shallow water are likely to be bivoltine (Tokeshi, 1995) with two generations a year. The first generation of adults, along with some univoltine species from deeper water, will emerge in spring (Tokeshi, 1995) and so might be influenced by the previous year's August temperatures, introducing a lagged response. The second generation, emerging in late summer/autumn (Tokeshi, 1995), will be affected by August temperatures but also temperatures throughout the summer and by the original species composition of the eggs. Lags can also be expected for subdecadal pollen-climate reconstructions: van der Knaap et al. (2010) used annual pollen trap data to show that the pollen accumulation rate for many taxa tends to depend on the conditions of the previous summer.

With a few exceptions (Kamenik et al., 2009; Larocque-Tobler et al., 2011a, 2011b; Luoto and Ojala, 2016), transfer-function models for quantitative palaeoclimate reconstructions are calibrated on a spatial calibration set of sites with paired microfossil and climatological normals that span a wide climatic range (Birks et al., 2010). Over a spatial calibration set, the correlation between air temperature in different months can be expected to be high. For example, in the Polish-Canadian chironomid calibration set used in the Lake Żabińskie study, the Pearson correlation between June and August mean air temperatures (Fick and Hijmans, 2017) is 0.97 (Fig. 1). Hence, it makes little difference if we calibrate the model against June, August or mean summer temperatures, as the model performance statistics will be very similar (Fig. 2a-b) and the reconstructions will all have a similar shape (Fig. 3). Any of these months can be reconstructed, provided the assumption that the relationship between monthly temperatures is invariant across time (Juggins, 2013). It makes sense to try to reconstruct temperature for the most ecologically important time of year, but the calibration

143 set may give little information on when this is.

144 While the calibration set is calibrated against climate normals, the target
 145 for validating the reconstruction is a time series of weather data. At an annual
 146 resolution, the correlations between months in the time series are low (Fig.
 147 1): a warm August does not necessarily follow a warm July. Consequently,
 148 the choice of calibration target has a large impact on the correlation between
 149 the reconstruction and the instrumental data. If a perfect reconstruction is
 150 validated against the wrong month, the performance will seem poor; a good
 151 correlation can only be expected for the portion of the year that the organisms
 152 are sensitive to. Surprisingly, despite the presumed sensitivity of chironomids
 153 to summer temperatures, the air-temperature reconstruction from the Lake
 154 Żabińskie chironomid stratigraphy is high only in August (Fig. 2c). The
 155 correlation between June and August air temperatures, at an annual resolution,
 156 in a composite of climate stations in the Lake Żabińskie region (similar to the
 157 composite used in Larocque-Tobler et al. (2015)) in the period 1896-2010, is only
 158 0.17. The correlation between mean summer air temperature and August air
 159 temperature in the same composite is 0.74. This is as good a correlation as can
 160 be expected for predicting August temperature from mean summer data collected
 161 with a thermometer. This problem is time-scale dependent: if temperature is
 162 averaged over several decades, as would be typical for a low- to medium-resolution
 163 palaeoecological analysis, the correlation between June and August temperatures,
 164 or August and the whole summer, increases because radiative forcing or long-term
 165 internal variability such as the Atlantic Multidecadal Oscillation (Schlesinger
 166 and Ramankutty, 1994) tends to warm the whole season and weather-related
 167 variability is suppressed (Fig. 1).

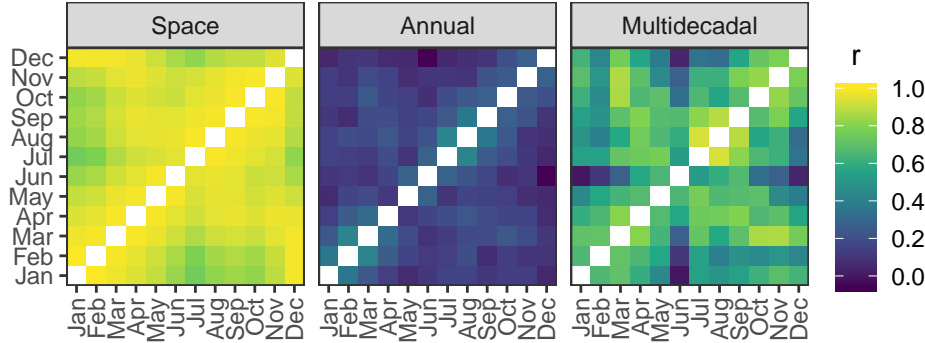


Figure 1: Spatial correlations of mean monthly temperatures in the Canadian-Polish calibration set (left), and temporal correlations of monthly temperatures in the central England temperature series (Parker et al., 1992) in single years (centre) and 30-year means (right).

168 Chironomids are probably more sensitive to water temperature than air
 169 temperature directly (Eggermont and Heiri, 2011); transfer-function models
 170 reconstructing air temperature depend on the strong relationship between air
 171 and water temperature. Air temperature is the most important determinant of
 172 water temperature (O'Reilly et al., 2015) but other factors such as wind speed

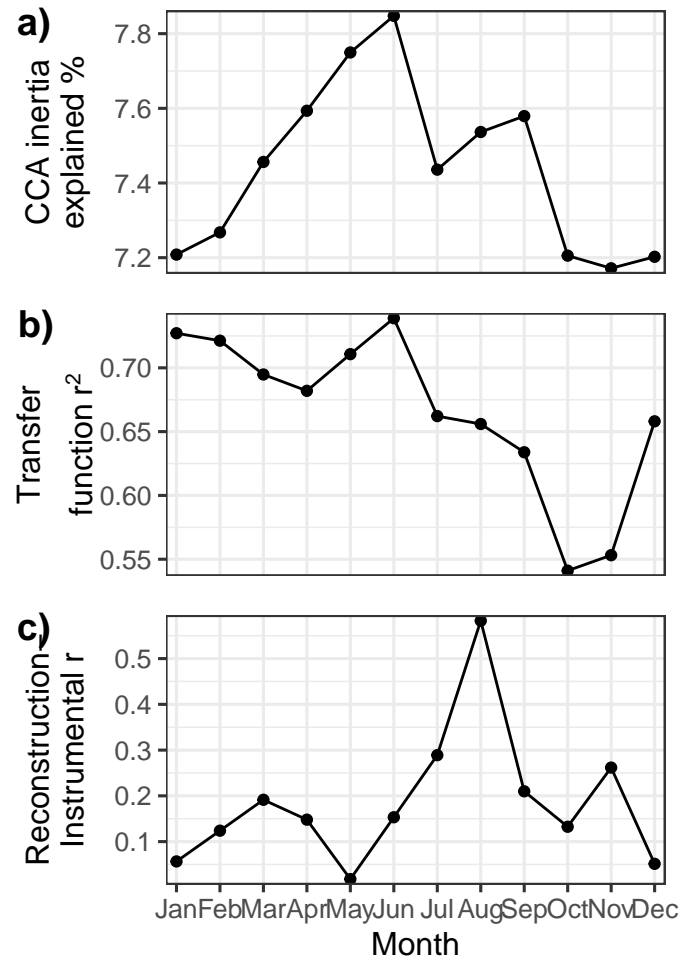


Figure 2: By month, a) the percent of inertia in the Canadian-Polish calibration set explained in a canonical correspondence analysis (CCA) constrained by temperature, b) transfer function leave-one-out cross-validation r^2 calibrated against temperature, c) correlation between the reconstructed temperature from Lake Żabińskie and the instrumental temperature

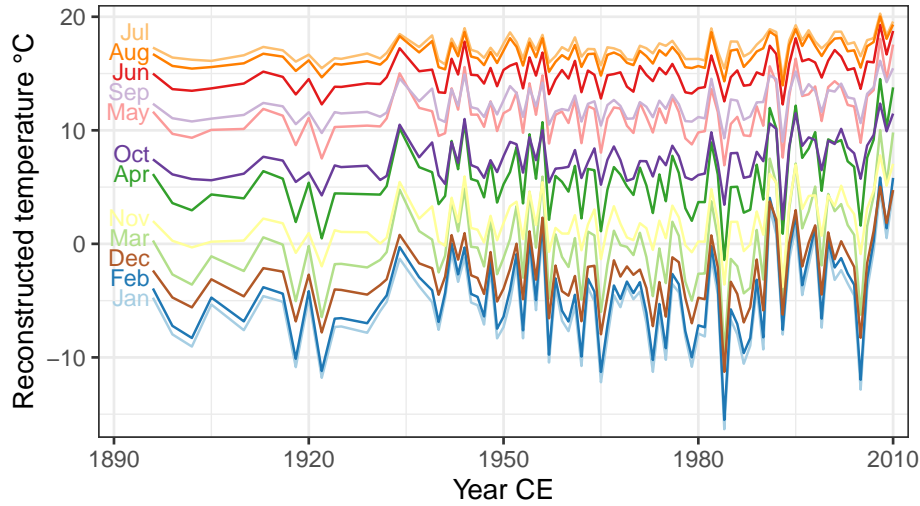


Figure 3: Temperature reconstructions for different months at Lake Żabińskie

and cloudiness also have an influence, so the inter-year correlation between air and water temperature is less than one. In Lake Zurich, this correlation is 0.67 in July (Livingstone and Lotter, 1998), varies between 0.59 and 0.76 for mean summer temperatures in eight Austrian lakes (Livingstone and Dokulil, 2001), and is about 0.75 throughout the summer in the small Plußsee in northern Germany (Rösner et al., 2012). An analysis of summer lake and air temperatures compiled by Sharma et al. (2015) shows that for small (area < 2km²), deep (mean depth > 5m), mid- to high-latitude lakes, the median correlation is 0.8 (n = 41). This correlation is as good as should be expected when comparing a thermometer in the lake with the instrumental air temperature.

While temperature undoubtedly has a strong effect on chironomid physiology, growth, and community structure, at least some of the apparent effect of temperature on chironomid assemblages in the calibration set is due to indirect effects of temperature (Eggermont and Heiri, 2011) acting through, for example, the relationship between temperature and productivity (Velle et al., 2010) or dissolved organic carbon (Larocque et al., 2001). These indirect effects will tend to strengthen the relationship between chironomid assemblages and temperature in the calibration set, but cannot be expected to fully operate on a monthly basis. As such, the cross-validation performance of the transfer function will tend to overestimate the model's predictive power on short time-scales.

Conversely, ecologically important variables other than the variable being reconstructed may have changed over time, which will bias the reconstructions (Juggins, 2013). For example, nutrient status (Amann et al., 2014; Witak et al., 2017) and anoxia (Bonk et al., 2015a; Wacnik et al., 2016), both important predictors of chironomid assemblages (Eggermont and Heiri, 2011), changed independently of August temperatures at Lake Żabińskie, and the naturally

199 oligotrophic Lake Silvaplana became eutrophic due to human impact before
200 recovering (Bigler et al., 2007).

201 2.2. Taphonomic processes

202 A strong correlation between an annual-resolution reconstruction and instru-
203 mental data can only be achieved if most of the microfossils in the sediment
204 attributed to a given year, lived in that year. This could be achieved if the
205 organisms lived at the core site (which might prevent the preservation of varves),
206 or if direct deposition of planktic organisms dominated the record.

207 Varved sediments can capture the seasonal succession of diatom communities
208 (Bonk et al., 2015b; Simola, 1977), although seasonal mixing can transport more
209 diatoms to the centre of the lake (Raubitschek et al., 1999). This suggests that
210 the largely planktic diatom record from the varved Baldeggersee (Lotter, 1998),
211 marked by abrupt changes in the assemblages, might be an example of the more
212 promising case of direct deposition of much of the assemblage.

213 By analogy with Lake De Waay (van Hardenbroek et al., 2011), the seasonally
214 anoxic hypolimnion (Bonk et al., 2015a) of Lake Żabińskie will be inhabited by
215 few chironomids (perhaps *Chironomus plumosus* which Nagell (1978) showed
216 can withstand long periods of anoxia, but without feeding), and the chironomid
217 assemblage in the centre of the lake will have been transported and redeposited
218 from the oxygenated shallows by wave-induced turbulence or currents during
219 seasonal mixing (Eggermont and Heiri, 2011). Lake Żabińskie experiences mixing
220 in spring and late autumn (Bonk et al., 2015a): complete mixing results in a
221 large sediment flux to the centre of the lake (Bonk et al., 2015a), but this does
222 not happen every year.

223 The importance of intense sediment focusing in Lake Żabińskie is evident
224 in the high ^{137}Cs and ^{210}Pb inventories (Tylmann et al., 2016). The diffused
225 ^{137}Cs peaks (Tylmann et al., 2016) indicate that not all the sediment in one
226 varve was generated in that year. The five year spread in the Chernobyl ^{137}Cs
227 peak might give an indication of the degree of smoothing that can be expected
228 in redeposited proxies. This is only a problem for very high resolution analyses.

229 The importance of sediment reworking in transporting chironomids to the
230 centre of Lake Żabińskie is demonstrated by the high proportion (73% on average)
231 of littoral taxa in the samples analysed (Larocque-Tobler et al., 2015). In Lake
232 Silvaplana, another lake with strong correlation between the chironomid-inferred
233 reconstruction and the instrumental record (Table 1), an average of 50% of the
234 assemblage is littoral taxa (Larocque et al., 2009a).

235 In lakes with oxic hypolimnia, there are other taphonomic and ecological pro-
236 cesses that will degrade the proxy record. Bioturbation and other processes will
237 mix the top few centimetres of sediment, acting as a low-pass filter. The smooth
238 changes in chironomid assemblages at Baker Lake (Medeiros et al., 2012) may
239 indicate the importance of such processes in this large deep lake. Stratigraphy
240 can potentially be further compromised as chironomids can burrow several cen-
241 timetres into the sediment (Charbonneau and Hare, 1998). Chironomids living
242 in the hypolimnion are isolated from summer temperature by the thermocline
243 (Livingstone and Lotter, 1998), and cannot be expected to respond directly to it.

244 2.3. Chronological precision

245 Reliable chronologies are essential for any proxy-proxy or proxy-instrumental
 246 data comparison (Trachsel and Telford, 2017). The impact of chronological
 247 error will depend on the magnitude of the chronological error relative to the
 248 persistence in the environmental variable being reconstructed. In cases such as
 249 the diatom reconstructions of total phosphorus (Alefs and Müller, 1999; Lotter,
 250 1998), where the target environmental variable rises and declines over decades,
 251 a small chronological error will have little effect on the agreement between the
 252 reconstruction and the instrumental records. If, however, the target variable
 253 has little or no autocorrelation, as for example with the Lake Żabińskie regional
 254 August air-temperature composite which has a lag-1 autocorrelation coefficient of
 255 0.16, then even an error of a single year would seriously degrade the agreement.

256 The chronologies supporting the subdecadal resolution reconstructions listed
 257 in Table 1 are either ^{210}Pb or varve based, with the exception of Kamenik et al.
 258 (2009) who use bomb- ^{14}C to date the peat sequence from Mauntschas Mire and
 259 report a dating uncertainty of 1–2 years. ^{210}Pb chronologies typically have an
 260 uncertainty of 5–10 years on sediment 100 years old (Appleby and Piliposian,
 261 2006). Varved chronologies are usually more precise, with a typical uncertainty
 262 of 1–3% of the age (Ojala et al., 2012) and allow the sediment from a single year
 263 (or known number of years) to be analysed. The error in varved chronologies
 264 tends to accumulate, becoming higher further back in time.

265 The impact of chronological uncertainty on the validation of an \sim annual
 266 resolution reconstruction can be illustrated with Lake Żabińskie. The varve-based
 267 chronology for Lake Żabińskie is supported by ^{14}C dates, a ^{137}Cs profile, and a
 268 microtephra (Bonk et al., 2015b). The counting uncertainty on the chronology
 269 is low: there are two possible extra varves in the mid-1960s and five possible
 270 missing varves between 1926 and 1899 CE. The impact of this uncertainty can
 271 be explored with a Monte Carlo procedure. Assuming that the probability of
 272 making a wrong decision at each uncertain varve is 1/3 and zero otherwise, the
 273 probability that the chronology is entirely correct is 0.06. If the reconstruction
 274 was perfect, chronological error is expected to reduce the correlation between
 275 the reconstruction and the instrumental record from 1 to 0.8. This reduction
 276 is relatively small because most of the uncertainty occurs in the lower half of
 277 the record, so any error affects fewer years. Furthermore, because chironomid
 278 concentrations were low, samples before 1939 span two to five years which
 279 reduces both the number of samples and their sensitivity to a single year’s error.
 280 The expected correlation in the pre-1939 section of the record for a perfect
 281 reconstruction is 0.69. This is lower than the reported correlation between the
 282 reconstruction and the instrumental record (0.81) for this period. Where the
 283 chronological uncertainty is higher than at Lake Żabińskie, for example at Lake
 284 Silvaplana where the uncertainty of the varve chronology is estimated to be 15%
 285 (Blass et al., 2007), the effect of chronological uncertainty will be greater.

286 There are some possible solutions to chronological uncertainty. Von Gunten
 287 et al. (2012) recommend smoothing the proxy data to allow for chronological
 288 errors. The three-year running-mean used by de Jong and Kamenik (2011) to

smooth their annually resolved chrysophyte data will reduce the impact of a one year dating error, and slightly reduce the impact of a two year error, however the loss of independence between consecutive years complicates the calculation of the significance of the correlation between the reconstruction and the instrumental data (von Gunten et al., 2012). Since the smoothed data have fewer degrees of freedom, they will have less statistical power if the chronology is correct. Smoothing annual resolution will generally have more statistical power than sampling the microfossil data at a lower resolution, but at the cost of having more samples to process and count. Either smoothing or sampling at a lower than annual resolution are valid strategies when the sediments are not laminated and single years cannot be precisely sampled.

An alternative strategy for dealing with chronological errors in varved sediment would be to identify which of an ensemble of possible chronologies are most plausible given the relationship between the proxy and the instrumental record (Werner and Tingley, 2015).

2.4. Calibration-in-time

Calibration-in-time models are routinely used in dendroclimatology (Shepard, 2010). They have been used in palaeolimnology where calibration-in-space models are not appropriate because the models need tailoring to a specific site (von Gunten et al., 2012). For example, they have been used to calibrate varve thickness against climate, especially for reconstructing summer temperature from proglacial lakes (Leemann and Niessen, 1994; Thomas and Briner, 2009); mass accumulation rate against winter precipitation (Elbert et al., 2012); and reflectance spectroscopy measurements against temperature (Saunders et al., 2013; Trachsel et al., 2010; von Gunten et al., 2009).

Kamenik et al. (2009) introduced calibration-in-time to microfossil data, using it to reconstruct April-November temperatures from the pollen stratigraphy from Mauntschas Mire, Switzerland. Subsequently, calibration-in-time models have been used to reconstruct summer temperatures from chironomid stratigraphies (Larocque-Tobler et al., 2011a, 2011b; Luoto and Ojala, 2016).

Calibration-in-time models are particularly sensitive to chronological errors and ecological or taphonomic lags as these affect both the calibration set and the fossil data to which the transfer function is applied.

Calibration-in-time models need to be checked against all the usual transfer function and reconstruction diagnostics. So far, this aspect of model development has been neglected. Because of the limited range of environmental conditions in the calibration set, there is a risk of non-analogue conditions downcore. Temporal autocorrelation needs to be checked for, as it may seriously bias transfer function performance statistics (Telford and Birks, 2009, 2005). The pollen stratigraphy in Kamenik et al. (2009) is strongly autocorrelated, partly because of the persistence of vegetation in the landscape, partly due to mixing of pollen in the acrotelm (Joosten and de Klerk, 2007), and partly due to linear interpolation and the three-year triangular filter used to smooth the pollen and climate data. Leave-one-out cross-validation cannot be expected to give credible performance statistics when adjacent samples have so little independence. With

334 h -block cross-validation (Trachsel and Telford, 2017), a scheme that is robust to
335 autocorrelation, or split sampling, the performance statistics would be worse,
336 but more realistic.

337 2.5. Promising and less promising proxies

338 Some proxy-environment systems appear more promising for subdecadal to
339 annual resolution reconstructions than others. Some attributes of lakes, proxies
340 and environmental variables associated with promising or unpromising systems
341 are given in Table 2.

Table 2: Aspects of proxies, sites and environmental variables leading to favourable or unfavourable reconstructions, with a brief explanation

Aspect	Favourable	Unfavourable	Explanation
Proxy sensitivity to target variable	High	Low	Better reconstruction
Habitat	Planktic	Benthic	Simple taphonomy
Proxy generation time	Hours-Days	Months-Years	Rapid response
Ecological lags	None	Proxy responds to previous year	Lagged proxy response
Target variable resolution	Seasonal	Monthly	More ecologically relevant
Other environmental change	Minimal	Substantial	Minimise secondary environmental gradients
Dominant frequency in target variable	Trends and low frequencies	High-frequencies	Less sensitive to lags and chronological error
Sediment	Varved	Not-varved	No bioturbation, demarcate one year's sediment
Chronology	Varves	Other	More precise dating

342 The diatom-total phosphorus reconstructions (Alefs and Müller, 1999; Lotter,
343 1998) derived from planktic diatoms hold the most promise. The planktic diatoms
344 have a short generation time; nutrient availability has a strong control on diatom
345 communities (Tilman, 1981); the taphonomy of sinking diatoms is relatively
346 simple; and the slow changes in the target environmental variable mean that
347 small chronological errors are inconsequential (conversely, this might make it
348 difficult to demonstrate skill on subdecadal rather than on decadal scales).

349 The chrysophyte-cyst climate system shares several of the advantages of the
350 diatoms, but is less promising because the target climate variable has strong
351 high-frequency variability so any chronological error or taphonomic mixing

will degrade the reconstruction. The three-year running mean used in all the chrysophyte studies will partially mitigate the high-frequency variability. The apparently successful reconstruction of summer wind speed (Hernández-Almeida et al., 2015b) and winter severity (Hernández-Almeida et al., 2015a) from the same stratigraphy is surprising given that these two climatic variables cannot be expected to be highly correlated. If both variables are ecologically important then the assumption that secondary environmental variables are unchanging is violated (Birks et al., 2010; Juggins, 2013), as each is a secondary environmental variable for the other, and good reconstructions cannot be expected. The mean summer wind reconstruction, relying on wind-induced upwelling of calcium to change species composition, seems particularly fragile, and is physically dubious as it seeks to find a relationship with mean wind velocity rather than mean wind speed. No data have been archived for any of the chrysophyte-cyst studies so their reproducibility cannot be verified.

Chironomids are even less promising for subdecadal reconstructions. In addition to the high-frequency variability in the target climatic variable, a chironomid’s life-span is long relative to the month-long target variable, the taphonomy is complex as most head capsules need to be transported from the littoral zone, and chironomid phenology means that many chironomids will have emerged before the summer so cannot be affected by summer temperatures in the year they emerge.

Doubts about the plausibility of subdecadal resolution chironomid reconstructions do not imply that reconstructions on longer timescales (Brooks et al., 2012) are flawed (but see Velle et al. (2010)). At multidecadal to centennial time scales, most of the issues discussed above are mitigated: single month and whole summer temperatures correlate well with each other as climate forcings rather than weather dominate; secondary variables such as dissolved organic carbon might correlate with climate change in time as they do in space and exert some influence on the assemblages; and lags due to taphonomic processes and phenology become largely irrelevant. It is not yet clear at what time scale the signal in the chironomid-inferred reconstruction can be expected to emerge from the noise.

While subdecadal resolution reconstructions from microfossil assemblages can only be expected to correlate well with instrumental data in favourable circumstances, very high-resolution palaeoecological analyses from varved sediments can be extremely valuable (Lotter and Anderson, 2012). For example, Allison et al. (1986) and Peglar (1993) use subdecadally (even annually) resolved pollen data to show the mid-Holocene *Tsuga* (Hemlock) and *Ulmus* (Elm) declines in North America and Europe, respectively, took just a few years.

3. Reproducibility issues with high-resolution chironomid reconstructions

Given the high correlations between chironomid-inferred temperatures and air temperature shown in Table 1 despite the severe problems such high-resolution reconstructions face, it is necessary to examine each reconstruction to try to

determine the reasons for these apparently good performances. Below, I show there are multiple problems with several of these reconstructions or their validation. Minor issues in some of the papers that do not have a large effect on their interpretation are omitted.

3.1. Lake Żabińskie

One might expect that a reconstruction as excellent as the Lake Żabińskie chironomid-temperature reconstruction would have excellent reconstruction diagnostics (Juggins and Birks, 2012). Indeed, the Żabińskie paper reports that all the fossil samples had good or fair analogues in the modern calibration set. However, using the methods described in the Żabińskie paper with the archived data (Larocque-Tobler et al., 2016b), 39 of the 89 fossil samples have no analogues (distance $> 95\%$ of calibration-set distances to nearest neighbour; Fig. S1). Another reconstruction diagnostic, the squared residual distance (Juggins and Birks, 2012), shows that 19 samples have a very poor fit ($> 95\%$ of calibration-set squared residual distances; Fig. S2), and a test of reconstruction significance (Telford and Birks, 2011) is not statistically significant ($p = 0.1$). This lack of significance is reflected in the relatively weak correlation ($r = 0.36$) between principal component axis 1 of the fossil data and August air temperature reported in the Żabińskie paper.

Since both the assemblage and instrumental temperature data are available, more direct tests can be applied than the reconstruction diagnostics used above. A canonical correspondence analysis fitted to the fossil assemblage data with August temperature as the sole predictor explains 2.47% of the variance in the assemblage data. This is statistically significant ($p = 0.001$) but the ratio of the constrained eigenvalue to the first unconstrained eigenvalue is far below one ($\lambda_1/\lambda_2 = 0.38$) suggesting that August temperature is not the most important axis of variability in the assemblage data and hence precise reconstructions cannot be expected (Juggins, 2013).

The surprisingly poor reconstruction diagnostics are not the only problematic aspect of the Żabińskie paper. The instrumental temperature series used by the paper was miscalculated for the period before 1939, the period in which chironomid samples were merged because of low abundances (Larocque-Tobler, pers comm). Rather than using the mean temperature for the interval corresponding to each sample, the temperature of the first year of the interval was used. Despite this, the pre-1939 section of the reconstruction has a better correlation ($r = 0.81$) with the incorrectly calculated temperature data than with the correct temperature data ($r = 0.53$).

The two ordinations in the Żabińskie paper cannot be reproduced from the archived data (seven substantially different versions of the fossil data were archived without explanation before the current version). Figure 2 in the Żabińskie paper is a principal component analysis of the calibration set. It shows only 103 of the 121 lakes in the calibration set (the paper reports 122): 18 lakes have been omitted, including all four lakes warmer than 23°C (Fig. 4a). The omission of these lakes makes the relationship between the chironomid

440 assemblages and temperature appear to be stronger than it is. In the pre-
 441 publication versions of this figure, several lakes had been moved rather than
 442 omitted.

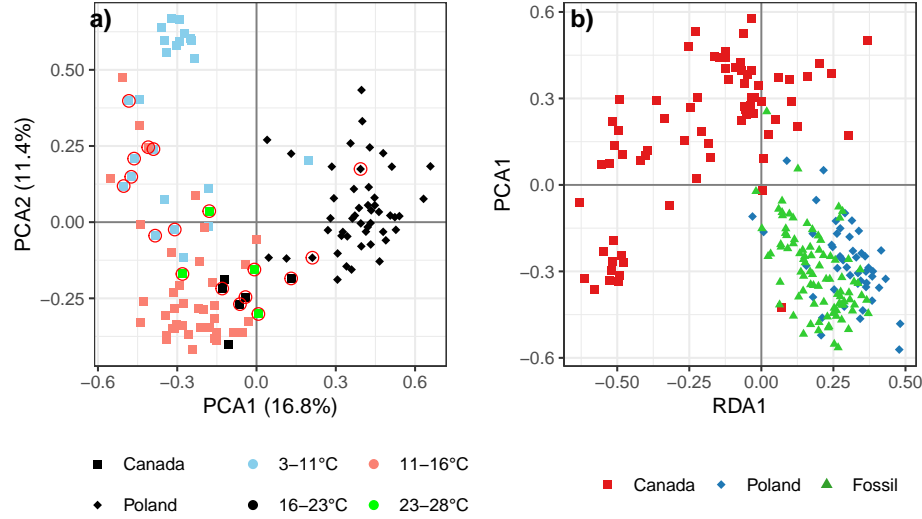


Figure 4: a) Attempt to reproduce the principal component analysis shown in Figure 2 from the Żabiński paper. Lakes enclosed by a red circle are missing from the published figure. Axis labels indicate the percent of variance in the calibration set explained (the Żabiński paper reports 22.3% and 17.5% for the first two axes). b) Attempt to reproduce the redundancy analysis shown in Supplementary Data Figure 1 from the Żabiński paper. The original figure has six more Canadian lakes in the lower right quadrant.

443 Conversely, Supplementary Data Figure 1 in the Żabiński paper, a redun-
 444 dancy analysis of the calibration set constrained to temperature with the fossil
 445 data added passively, has more lakes than expected (Fig. 4b). There are 73
 446 Canadian lakes in the calibration set (the paper reports 72) but the figure shows
 447 78. The extra Canadian lakes are mostly in the same quadrant as the Polish
 448 lakes and falsely suggest “a similarity between the assemblages in warmer lakes
 449 in Canada and those which cover the same temperature gradient in Poland”
 450 (Larocque-Tobler et al., 2015).

451 The Żabiński paper reports that the count sums in both the fossil and
 452 modern data sets were at least 50 chironomid head capsules. In a corrigendum,
 453 Larocque-Tobler et al. (2016a) subsequently acknowledged that many samples
 454 had lower count sums and removed nine of the 48 Polish lakes from the calibration
 455 set and reported that the count sums for the fossil data are as low as 19. Eighty
 456 eight percent of fossil samples have a count sum lower than that initially reported.
 457 It is well known (Heiri and Lotter, 2001; Larocque, 2001; Quinlan and Smol, 2001)
 458 that low count sums are associated with increased uncertainty in reconstructions.
 459 The magnitude of the increase in uncertainty depends on the interaction between
 460 species diversity and the differences in optima between taxa. Here, I estimate the
 461 uncertainty due to low counts by finding the standard deviation of the transfer

function predictions for the 59 lakes in the calibration set which appear to have more than 100 chironomids when resampled to count sums spanning the range observed in the fossil data. For each fossil sample from Lake Żabińskie, I estimate the reconstruction uncertainty from the mean of these standard deviations. The expected uncertainty in the reconstruction given the count sums is 1.15°C. This is slightly larger than the observed standard deviation of the residuals, 0.95°C, which is due to all sources of error.

The chironomid counts from Lake Żabińskie are unusual. Typically the rarest species in any assemblage is represented by a single individual. In the Lake Żabińskie counts, 38% of samples have a minimum abundance greater than one. In one sample, all taxa occur at least five times, in another, all counts are multiples of four.

The calibration set also has problems. Lac A and Lac H have identical assemblages, except that the percentages in Lac H have been multiplied by 1.79 such that they sum to 175%. Errors of up to 6.5°C in the August temperatures of some of the Canadian lakes were corrected by Bajolle et al. (2018).

3.2. *Abisko lakes*

Larocque and Hall (2003) report reconstructions of July air temperature for four lakes in northern Sweden. The correlations of the reconstructions with instrumental data, all between 0.35 and 0.39, are surprisingly similar. Using a simulation analogous to that used by Simonsohn (2013), the probability of obtaining a standard deviation of the correlations as low as that observed is < 0.001.

Of the four correlations, only one matches the correlation of reconstruction and instrumental data digitised from the authors' figure 4 – Alanen Laanijavri. For the other lakes, the correlations are 0.04, 0.45, and 0.69 for Lake Njulla, Lake 850, and Vuoskkujavri, respectively (Fig. S3). No assemblage data have been archived, so the reconstructions cannot be verified, but there are some discrepancies between the published stratigraphies and the reconstructions. Lake 850 lacks an assemblage for 1999 CE but has a reconstruction. Similarly, Lake Njulla has a reconstruction but no assemblage for 1923 CE and the converse for 1949 CE.

3.3. *Lake Silvaplana*

The chironomid-based July temperature reconstruction from the proglacial Lake Silvaplana for the period 1850–2001 CE is first presented in Larocque et al. (2009a). Subsequently, Larocque-Tobler et al. (2009) and Larocque-Tobler et al. (2010) published 420 and 1000 year long reconstructions, respectively, which both include the 1850–2001 CE period. The three reconstructions for the last 150 years should be identical, or at least nearly so, however they are very different. Furthermore, archived reconstructions for the 150-year reconstruction (Larocque et al., 2008) and the 420-year reconstruction (Larocque et al., 2009b) are different from their associated papers, giving five different reconstructions for this time interval. None of the papers include any explanation of these differences. The

505 overlapping portions of the archived 150- and 420-year reconstructions can largely
 506 be reconciled by aligning by sample number rather than age, which stretches
 507 the time series by up to 42 years (Fig. 5) – far more than the chronological
 508 uncertainty of the varve record (Blass et al., 2007). Remaining differences
 509 between the aligned reconstructions are nearly all integer $^{\circ}\text{C}$ differences.

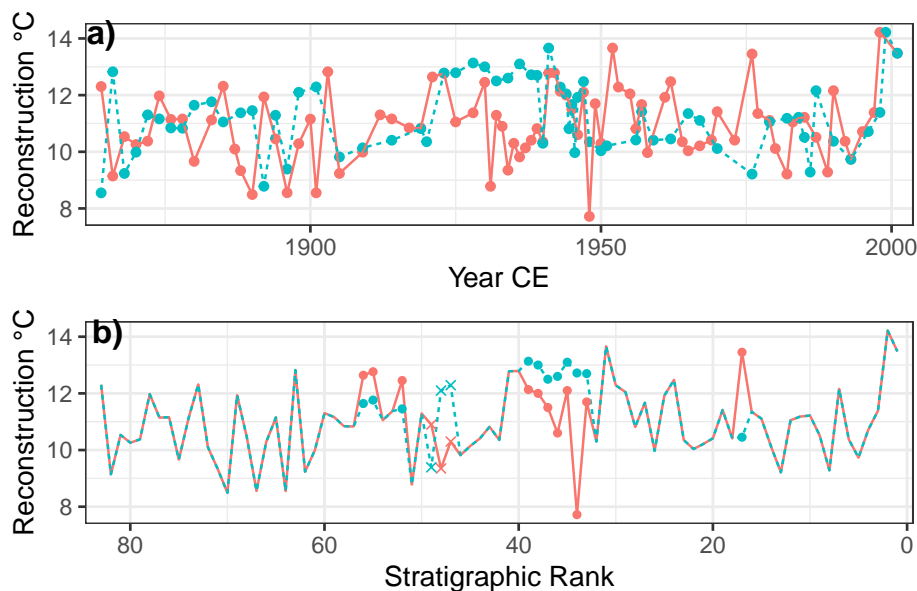


Figure 5: The 150-yr reconstruction (solid, red) and the overlapping portion of the 420-yr reconstruction (dashed, blue) from Silvaplana plotted against a) date, and b) stratigraphic rank. In b) circles indicate integer $^{\circ}\text{C}$ differences, crosses show non-integer differences between reconstructions.

510 Fossil assemblage percent data have been archived for the 420-year reconstruction.
 511 Unfortunately the count sums have not been archived but were reported to
 512 be at least 30. However, the data suggest that many of the samples have much
 513 smaller counts. For example, the sample for 1580 CE has two taxa with relative
 514 abundances of 16.67% and 83.33%. Given that the rarest taxa is probably
 515 represented by a single individual, it seems that this represents a count of 6
 516 individuals. Inspection of Figure 3a from Larocque-Tobler et al. (2009), which
 517 shows the raw counts, supports this interpretation. Forty one of the 134 samples
 518 appear to have a count of fewer than 20 head capsules. In older material from
 519 Lake Silvaplana (Stewart et al., 2011), about a quarter of the fossil samples are
 520 represented by a single taxon with an abundance of 100%: these counts are likely
 521 to be of a single individual. Such extremely low counts cannot be expected to
 522 yield precise reconstructions.

523 The instrumental temperature series used for the calibration-in-time recon-
 524 struction (Larocque-Tobler et al., 2011a) differs substantially from the series
 525 used in the 150-year calibration-in-space paper, and both are different from the
 526 instrumental data in Larocque-Tobler et al. (2010). No version of the instrumental

527 data can be reconciled with the July temperature data from Segl-Maria (Fig S4).

528 3.4. Seebergsee

529 No fossil data have been archived from the anoxic Seebergsee, but the raw
530 count data from Figure 4a in Larocque-Tobler et al. (2011b) can be digitised and
531 samples merged to match their Figure 4b. The paper reports that adjacent fossil
532 samples were merged until there were at least 30 chironomids. However, 69% of
533 the merged samples have count sums lower than 30 with a minimum count sum
534 of twelve. Curiously, all chironomid counts in Figure 4a of the Seebergsee paper
535 are multiples of four. Since the rarest taxa in the fossil data is almost certainly
536 present as a single individual, the probability that all counts are multiples of four
537 is infinitesimally small (under very optimistic assumptions, the probability that
538 a given count is divisible by four is 0.25. The probability that all 186 counts are
539 divisible by four is 10^{-112}). This suggests that the counts have been multiplied
540 by four. If so, all the count sums are below 30 and the minimum count sum is
541 only three.

542 Digitising the July instrumental temperature data from figure 6 in the
543 Seebergsee paper allows the reproducibility of the calibration-in-time result to
544 be tested. The paper does not report which transfer function method was used,
545 but a two component WAPLS model on square-root transformed assemblage
546 data yields a reconstruction that has a correlation of 0.73 with the instrumental
547 data, identical to that reported, and the reconstruction appears to be identical.
548 However, this performance is only achieved if the model is not cross-validated.
549 With leave-one-out cross-validation, which is essential to prevent overfitting
550 and give an unbiased estimate of performance, the correlation drops to 0.13,
551 and the reconstruction has no skill. The importance of cross-validation can be
552 demonstrated by trying to reconstruct random noise: the observed correlation is
553 exceeded by 16.7% of 1000 trials reconstructing Gaussian noise, and so cannot
554 be considered remarkable.

555 There are some other issues with the Seebergsee paper. These include
556 substantial discrepancies between the age-depth model shown in its Figure 3
557 and Figure 4b; difficulties reconciling the reported instrumental data with the
558 series from Château d'Oex (Fig. S6); and the report that the first and second
559 axes of a correspondence analysis (CA) of the fossil data explain 45% and 27%
560 of the inertia, respectively. My attempt to reproduce the CA finds that the first
561 and second axes explain only 13% and 12%, respectively, of the inertia.

562 Larocque-Tobler et al. (2012) use the calibration-in-time model from Larocque-
563 Tobler et al. (2011b) to reconstruct July air temperature from a millennium-long
564 chironomid stratigraphy from Seebergsee. In view of the very poor cross-validated
565 performance of the calibration-in-time model ($r^2 = 0.02$), the millennium-long
566 reconstruction cannot be expected to be skilful.

567 3.5. Baker Lake

568 The main features of the reconstruction of mid-summer water temperature
569 from Baker Lake (Medeiros et al., 2012) can be reproduced from the raw data

(Andrew S. Medeiros, pers comm), and the correlation with the smoothed annual temperature anomaly is high. Baker Lake is a relatively favourable study system: changes in the short ice-free period of this arctic lake due to climate change will have ecological consequences; and the lake experienced a large, year-round warming trend during the studied interval. However, chronological uncertainties, the relatively low resolution, and autocorrelation due to both the relatively smooth chironomid stratigraphy, which may be due to taphonomic processes, and the smoothing of the instrumental data, act together to preclude the detection of a subdecadal signal so the authors never attempted to interpret one.

3.6. *Nurmijärvi*

No data have been archived for Luoto and Ojala (2016), but it is possible to digitise the fossil stratigraphy and test the reproducibility of the calibration-in-time reconstruction of July air temperature. The calibration-in-time model is reported as having a leave-one-out r^2 of 0.64, whereas the reconstruction has a reported correlation with instrumental temperature of only 0.51 (i.e. $r^2 = 0.26$): these two performance statistics should be identical. With the digitised data, I find the leave-one-out r^2 to be only 0.13 (Fig. S7). It is possible that the apparent performance was reported by mistake.

The Nurmijärvi paper reports that there is a 4–8 year lag (perhaps caused by the ecological and taphonomic processes described above) in the chironomid response to temperature, which would seem to preclude the development of a good calibration-in-time model.

3.7. *Tiancai Lake*

Zhang et al. (2017a) report that they correlate their July air temperature reconstruction with a three-year moving average of the instrumental data (Zhang et al., 2017b). However, the instrumental time series has been mis-processed. Rather than using a three-year moving average, the authors take the July air temperature of every third year and interpolate. The significant correlation between the reconstruction and the interpolated triennial data can only be a Type 1 error, unless it is believed that chironomids are only sensitive to July temperature every third year and that they can predict the temperature up to two years ahead. With a three-year moving average, the correlation is weak ($r = 0.12$) and not statistically significant ($p = 0.54$).

3.8. *Speke Hall Lake*

The polluted, eutrophic Speke Hall Lake (Lang et al., 2017) is a curious choice of site on which to apply the Norwegian chironomid calibration set (Brooks and Birks, 2001), conflicting with the usual guidance that sites should be within the environmental space covered by the calibration set to minimise non-analogue problems. Perhaps not surprisingly, none of the fossil samples have good analogues in the calibration set (Fig. S8), and other diagnostics are generally very poor (Fig. S9). The reported lack of correlation between the

612 detrended correspondence analysis scores and the reconstruction (which can
613 be reproduced from the raw data (Andrew S. Medeiros, pers comm)) suggests
614 that temperature is not the most important variable driving the variability in
615 assemblage composition. The reported correlation between the reconstruction
616 and instrumental data, based on sixteen samples, is statistically significant at
617 the $p < 0.05$ level, but given the lack of correction for temporal autocorrelation,
618 the inverse correlation between July temperature and most of the heavy metals
619 measured, and the very poor reconstruction diagnostics, it is more likely that
620 the correlation is due to chance than that the reconstruction has any real skill.

621 4. Conclusions

622 Reliable subdecadal resolution reconstructions depend on the conjunction
623 of a strong direct ecological link to the variable being reconstructed and a
624 simple taphonomy. Accurate and precise chronologies are also needed for a
625 direct comparison with instrumental records. These requirements might be
626 met by diatom-total phosphorus reconstructions, and perhaps by chrysophyte
627 reconstructions of winter climate. It appears unlikely that they will be met by
628 the relatively long-lived, predominately littoral chironomids for reconstructions
629 of a single month's temperature.

630 The apparently extensive evidence for the utility of subdecadal resolution
631 temperature reconstructions is shown to be illusory. The papers reporting
632 chironomid-temperature reconstructions from Lake Żabińskie, the lakes near
633 Abisko, Lake Silvaplana, and Seeburgsee are plagued by pervasive errors and
634 improbabilities and key results cannot be reproduced. The reported correlation
635 between instrumental data and the reconstruction from Tiancai Lake cannot be
636 reproduced and at least some results from Nurmijärvi cannot be reproduced.
637 The Speke Hall Lake reconstruction can be reproduced, but this study is unlikely
638 to replicate in a similar site. The Baker Lake reconstruction is reproducible and
639 it is plausible that the result could be replicated at a similar site, but there is
640 no subdecadal signal.

641 The number and severity of errors in some of the papers reviewed highlights
642 the importance of adopting reproducible practices and archiving all the raw data
643 and computer code needed to reproduce the results.

644 5. Methods

645 All analyses were done in R version 3.4.4 (R Core Team, 2018). Ordinations
646 were fitted with vegan version 2.5.5 (Oksanen et al., 2019) with square-root
647 transformed assemblage data. Transfer functions were fitted with rioja version
648 0.9.21 (Juggins, 2017) using square-root transformed species data. Some diagnos-
649 tics were performed using analogue version 0.17.1 (Simpson and Oksanen, 2018).
650 The reconstruction significance test was run with randomTF from palaeoSig
651 version 2.0.3 (Telford, 2019) with 999 trials for the null distribution.

652 Spatial climate data were extracted from WorldClim2 (Fick and Hijmans,
653 2017); some of the lake positions are approximate as their locations are not

all archived. The regional Polish temperature composite included data from Kaliningrad, Kaunas, Vilnius, and Warsaw GHCN series (Lawrimore et al., 2011) extracted using the KNMI Climate Explorer. The 1951-1980 mean was subtracted from each series. Code to reproduce this paper is archived at <https://doi.org/xxxx.xxxx>.

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References

- Alefs, J., Müller, J., 1999. Differences in the eutrophication dynamics of Ammersee and Starnberger See (southern Germany), reflected by the diatom succession in varve-dated sediments. *Journal of Paleolimnology* 21, 395–407. <https://doi.org/10.1023/A:1008098118867>
- Allison, T.D., Moeller, R.E., Davis, M.B., 1986. Pollen in laminated sediments provides evidence for a mid-Holocene forest pathogen outbreak. *Ecology* 67, 1101–1105. <https://doi.org/10.2307/1939835>
- Amann, B., Lobsiger, S., Fischer, D., Tylmann, W., Bonk, A., Filipiak, J., Grosjean, M., 2014. Spring temperature variability and eutrophication history inferred from sedimentary pigments in the varved sediments of Lake Zabinskie, north-eastern Poland, AD 1907-2008. *Global and Planetary Change* 123, 86–96. <https://doi.org/10.1016/j.gloplacha.2014.10.008>
- Appleby, P.G., Piliposian, G.T., 2006. Radiometric dating of sediment records from mountain lakes in the Tatra Mountains. *Biologia* 61, S51–S64. <https://doi.org/10.2478/s11756-006-0119-4>
- Bajolle, L., Larocque-Tobler, I., Gandouin, E., Lavoie, M., Bergeron, Y., Ali, A.A., 2018. Major postglacial summer temperature changes in the central coniferous boreal forest of Quebec (Canada) inferred using chironomid assemblages. *Journal of Quaternary Science* 33, 409–420. <https://doi.org/10.1002/jqs.3022>
- Bigler, C., von Gunten, L., Lotter, A.F., Hausmann, S., Blass, A., Ohlendorf, C., Sturm, M., 2007. Quantifying human-induced eutrophication in Swiss mountain lakes since AD 1800 using diatoms. *The Holocene* 17, 1141–1154. <https://doi.org/10.1177/0959683607082555>
- Birks, H.J.B., Heiri, O., Seppä, H., Bjune, A.E., 2010. Strengths and weaknesses of quantitative climate reconstructions based on late-Quaternary biological proxies. *The Open Ecology Journal* 3, 68–110. <https://doi.org/10.2174/1874213001003020068>
- Blass, A., Grosjean, M., Troxler, A., Sturm, M., 2007. How stable are twentieth-century calibration models? A high-resolution summer temperature reconstruction for the eastern Swiss Alps back to AD 1580 derived from

696 proglacial varved sediments. *The Holocene* 17, 51–63. [https://doi.org/10.1177/](https://doi.org/10.1177/0959683607073278)
697 [0959683607073278](https://doi.org/10.1177/0959683607073278)

698 Bonk, A., Tylmann, W., Amann, B., Enters, D., Grosjean, M., 2015a. Modern
699 limnology and varve-formation processes in Lake Zabinskie, northeastern Poland:
700 comprehensive process studies as a key to understand the sediment record.
701 *Journal of Limnology* 74, 358–370. <https://doi.org/10.4081/jlimnol.2014.1117>

702 Bonk, A., Tylmann, W., Goslar, T., Wacnik, A., Grosjean, M., 2015b.
703 Comparing varve counting and C-14-AMS chronologies in the sediments of Lake
704 Zabinskie, northeastern Poland: implications for accurate C-14 dating of lake
705 sediments. *Geochronometria* 42, 159–171. [https://doi.org/10.1515/geochr-2015-](https://doi.org/10.1515/geochr-2015-0019)
706 [0019](https://doi.org/10.1515/geochr-2015-0019)

707 Brooks, S.J., Birks, H.J.B., 2001. Chironomid-inferred air temperatures from
708 Lateglacial and Holocene sites in north-west Europe: progress and problems.
709 *Quaternary Science Reviews* 20, 1723–1741. [https://doi.org/10.1016/s0277-](https://doi.org/10.1016/s0277-3791(01)00038-5)
710 [3791\(01\)00038-5](https://doi.org/10.1016/s0277-3791(01)00038-5)

711 Brooks, S.J., Matthews, I.P., Birks, H.H., Birks, H.B., 2012. High resolution
712 Lateglacial and early-Holocene summer air temperature records from Scotland
713 inferred from chironomid assemblages. *Quaternary Science Reviews* 41, 67–82.
714 <https://doi.org/10.1016/j.quascirev.2012.03.007>

715 Charbonneau, P., Hare, L., 1998. Burrowing behavior and biogenic structures
716 of mud-dwelling insects. *Journal of the North American Benthological Society*
717 17, 239–249. <https://doi.org/10.2307/1467965>

718 de Jong, R., Kamenik, C., 2011. Validation of a chrysophyte stomatocyst-
719 based cold-season climate reconstruction from high-alpine Lake Silvaplana,
720 Switzerland. *Journal of Quaternary Science* 26, 268–275. [https://doi.org/](https://doi.org/10.1002/jqs.1451)
721 [10.1002/jqs.1451](https://doi.org/10.1002/jqs.1451)

722 D’Arrigo, R., Wilson, R., Liepert, B., Cherubini, P., 2008. On the “divergence
723 problem” in northern forests: a review of the tree-ring evidence and possible
724 causes. *Global and Planetary Change* 60, 289–305. [https://doi.org/10.1016/j.](https://doi.org/10.1016/j.gloplacha.2007.03.004)
725 [gloplacha.2007.03.004](https://doi.org/10.1016/j.gloplacha.2007.03.004)

726 Eggermont, H., Heiri, O., 2011. The chironomid-temperature relationship:
727 expression in nature and palaeoenvironmental implications. *Biological Reviews*
728 87, 430–456. <https://doi.org/10.1111/j.1469-185x.2011.00206.x>

729 Elbert, J., Grosjean, M., von Gunten, L., Urrutia, R., Fischer, D., Warten-
730 burger, R., Ariztegui, D., Fujak, M., Hamann, Y., 2012. Quantitative high-
731 resolution winter (JJA) precipitation reconstruction from varved sediments of
732 Lago Plomo 47°S, Patagonian Andes, AD 1530–2002. *The Holocene* 22, 465–474.
733 <https://doi.org/10.1177/0959683611425547>

734 Fick, S.E., Hijmans, R.J., 2017. WorldClim 2: new 1-km spatial resolution
735 climate surfaces for global land areas. *International Journal of Climatology* 37,
736 4302–4315. <https://doi.org/10.1002/joc.5086>

737 Fidler, F., Chee, Y.E., Wintle, B.C., Burgman, M.A., McCarthy, M.A.,
738 Gordon, A., 2017. Metaresearch for evaluating reproducibility in ecology and
739 evolution. *BioScience* 67, 282–289. <https://doi.org/10.1093/biosci/biw159>

740 Green, D.G., 1983. The ecological interpretation of fine resolution pollen
741 records. *New Phytologist* 94, 459–477. <https://doi.org/10.1111/j.1469-8137.1983.>

742 tb03459.x

743 Heiri, O., Lotter, A.F., 2001. Effect of low count sums on quantitative
744 environmental reconstructions: an example using subfossil chironomids. *Journal*
745 *of Paleolimnology* 26, 343–350. <https://doi.org/10.1023/a:1017568913302>

746 Hernández-Almeida, I., Grosjean, M., Gómez-Navarro, J.J., Larocque-Tobler,
747 I., Bonk, A., Enters, D., Ustrzycka, A., Piotrowska, N., Przybylak, R., Wacnik,
748 A., Witak, M., Tylmann, W., 2017. Resilience, rapid transitions and regime
749 shifts: fingerprinting the responses of Lake Żabińskie (NE Poland) to climate
750 variability and human disturbance since AD 1000. *The Holocene* 27, 258–270.
751 <https://doi.org/10.1177/0959683616658529>

752 Hernández-Almeida, I., Grosjean, M., Przybylak, R., Tylmann, W., 2015a.
753 A chrysophyte-based quantitative reconstruction of winter severity from varved
754 lake sediments in NE Poland during the past millennium and its relationship
755 to natural climate variability. *Quaternary Science Reviews* 122, 74–88. <https://doi.org/10.1016/j.quascirev.2015.05.029>

756 Hernández-Almeida, I., Grosjean, M., Tylmann, W., Bonk, A., 2015b. Chrys-
757 ophyte cyst-inferred variability of warm season lake water chemistry and climate
758 in northern Poland: training set and downcore reconstruction. *Journal of*
759 *Paleolimnology* 53, 123–138. <https://doi.org/10.1007/s10933-014-9812-4>

760 Joosten, H., de Klerk, P., 2007. In search of finiteness: The limits of
761 fine-resolution palynology of sphagnum peat. *The Holocene* 17, 1023–1031.
762 <https://doi.org/10.1177/0959683607082416>

763 Juggins, S., 2017. Rioja: Analysis of Quaternary science data. R package
764 version 0.9-21. <http://www.staff.ncl.ac.uk/stephen.juggins/>

765 Juggins, S., 2013. Quantitative reconstructions in palaeolimnology: New
766 paradigm or sick science? *Quaternary Science Reviews* 64, 20–32. <https://doi.org/10.1016/j.quascirev.2012.12.014>

767 Juggins, S., Birks, H.J.B., 2012. Quantitative environmental reconstructions
768 from biological data, in: Birks, H.J.B., Lotter, A.F., Juggins, S., Smol, J.P.
769 (Eds.), *Tracking Environmental Change Using Lake Sediments, Vol. 5: Data*
770 *Handling and Numerical Techniques*. Springer, Dordrecht, pp. 431–494. https://doi.org/10.1007/978-94-007-2745-8_14

771 Kamenik, C., van der Knaap, W.O., van Leeuwen, J.F.N., Goslar, T., 2009.
772 Pollen/climate calibration based on a near-annual peat sequence from the Swiss
773 Alps. *Journal of Quaternary Science* 24, 529–546. <https://doi.org/10.1002/jqs.1266>

774 Lang, B., Medeiros, A.S., Worsley, A., Bedford, A., Brooks, S.J., 2017.
775 Influence of industrial activity and pollution on the paleoclimate reconstruction
776 from a eutrophic lake in lowland England, UK. *Journal of Paleolimnology*.
777 <https://doi.org/10.1007/s10933-017-9995-6>

778 Larocque, I., 2001. How many chironomid head capsules are enough? A
779 statistical approach to determine sample size for palaeoclimatic reconstructions.
780 *Palaeogeography, Palaeoclimatology, Palaeoecology* 172, 133–142. [https://doi.org/10.1016/s0031-0182\(01\)00278-4](https://doi.org/10.1016/s0031-0182(01)00278-4)

781 Larocque, I., Grosjean, M., Heiri, O., Bigler, C., Blass, A., 2009a. Comparison
782 between chironomid-inferred July temperatures and meteorological data AD
783

1850–2001 from varved Lake Silvaplana, Switzerland. *Journal of Paleolimnology* 41, 329–342. <https://doi.org/10.1007/s10933-008-9228-0>

Larocque, I., Grosjean, M., Heiri, O., Bigler, C., Blass, A., 2008. Larocque et al. 2008 Lake Silvaplana, Switzerland chironomid temperature [WWW Document]. <https://www.ncdc.noaa.gov/paleo/study/6394>

Larocque, I., Grosjean, M., Heiri, O., Trachsel, M., 2009b. Larocque et al. 2009 Lake Silvaplana 420yr chironomid inferred temperature [WWW Document]. <https://www.ncdc.noaa.gov/paleo/study/8737>

Larocque, I., Hall, R.I., 2003. Chironomids as quantitative indicators of mean July air temperature: Validation by comparison with century-long meteorological records from northern Sweden. *Journal of Paleolimnology* 29, 475–493. <https://doi.org/10.1023/A:1024423813384>

Larocque, I., Hall, R.I., Grahn, E., 2001. Chironomids as indicators of climate change: a 100-lake training set from a subarctic region of northern Sweden (Lapland). *Journal of Paleolimnology* 26, 307–322. <https://doi.org/10.1023/a:1017524101783>

Larocque, I., Pienitz, R., Rolland, N., 2006. Factors influencing the distribution of chironomids in lakes distributed along a latitudinal gradient in northwestern Quebec, Canada. *Canadian Journal of Fisheries and Aquatic Sciences* 63, 1286–1297. <https://doi.org/10.1139/f06-020>

Larocque-Tobler, I., 2010. Reconstructing temperature at Egelsee, Switzerland, using North American and Swedish chironomid transfer functions: Potential and pitfalls. *Journal of Paleolimnology* 44, 243–251. <https://doi.org/10.1007/s10933-009-9400-1>

Larocque-Tobler, I., Filipiak, J., Tylmann, W., Bonk, A., Grosjean, M., 2016a. Corrigendum to “Comparison between chironomid-inferred mean-August temperature from varved Lake Żabińskie (Poland) and instrumental data since 1896 AD” [Quat. Sci. Rev. 111 (2015) 35–50]. *Quaternary Science Reviews* 140, 163–167. <https://doi.org/10.1016/j.quascirev.2016.01.020>

Larocque-Tobler, I., Filipiak, J., Tylmann, W., Bonk, A., Grosjean, M., 2016b. Lake Żabińskie, Poland 115 year August chironomid-inferred temperature [WWW Document]. <https://www.ncdc.noaa.gov/paleo-search/study/19501>

Larocque-Tobler, I., Filipiak, J., Tylmann, W., Bonk, A., Grosjean, M., 2015. Comparison between chironomid-inferred mean-August temperature from varved Lake Żabińskie (Poland) and instrumental data since 1896 AD. *Quaternary Science Reviews* 111, 35–50. <https://doi.org/10.1016/j.quascirev.2015.01.001>

Larocque-Tobler, I., Grosjean, M., Heiri, O., Trachsel, M., 2009. High-resolution chironomid-inferred temperature history since AD 1580 from varved Lake Silvaplana, Switzerland: comparison with local and regional reconstructions. *The Holocene* 19, 1201–1212. <https://doi.org/10.1177/0959683609348253>

Larocque-Tobler, I., Grosjean, M., Heiri, O., Trachsel, M., Kamenik, C., 2010. Thousand years of climate change reconstructed from chironomid subfossils preserved in varved Lake Silvaplana, Engadine, Switzerland. *Quaternary Science Reviews* 29, 1940–1949. <https://doi.org/10.1016/j.quascirev.2010.04.018>

Larocque-Tobler, I., Grosjean, M., Kamenik, C., 2011a. Calibration-in-time versus calibration-in-space (transfer function) to quantitatively infer July

air temperature using biological indicators (chironomids) preserved in lake
sediments. *Palaeogeography, Palaeoclimatology, Palaeoecology* 299, 281–288.
<https://doi.org/10.1016/j.palaeo.2010.11.008>

Larocque-Tobler, I., Quinlan, R., Stewart, M.M., Grosjean, M., 2011b. Chironomid-inferred temperature changes of the last century in anoxic Seebensee, Switzerland: assessment of two calibration methods. *Quaternary Science Reviews* 30, 1770–1779. <https://doi.org/10.1016/j.quascirev.2011.04.008>

Larocque-Tobler, I., Stewart, M.M., Quinlan, R., Trachsel, M., Kamenik, C., Grosjean, M., 2012. A last millennium temperature reconstruction using chironomids preserved in sediments of anoxic Seebensee (Switzerland): consensus at local, regional and Central European scales. *Quaternary Science Reviews* 41, 49–56. <https://doi.org/10.1016/j.quascirev.2012.03.010>

Lawrimore, J.H., Menne, M.J., Gleason, B.E., Williams, C.N., Wuertz, D.B., Vose, R.S., Rennie, J., 2011. An overview of the Global Historical Climatology Network monthly mean temperature data set, version 3. *Journal of Geophysical Research: Atmospheres* 116, D19121. <https://doi.org/10.1029/2011JD016187>

Leemann, A., Niessen, F., 1994. Varve formation and the climatic record in an Alpine proglacial lake: calibrating annually-laminated sediments against hydrological and meteorological data. *The Holocene* 4, 1–8. <https://doi.org/10.1177/095968369400400101>

Livingstone, D.M., Dokulil, M.T., 2001. Eighty years of spatially coherent Austrian lake surface temperatures and their relationship to regional air temperature and the North Atlantic Oscillation. *Limnology and Oceanography* 46, 1220–1227. <https://doi.org/10.4319/lo.2001.46.5.1220>

Livingstone, D.M., Lotter, A.F., 1998. The relationship between air and water temperatures in lakes of the Swiss Plateau: a case study with palaeolimnological implications. *Journal of Paleolimnology* 19, 181–198. <https://doi.org/10.1023/A:1007904817619>

Lotter, A.F., 1998. The recent eutrophication of Baldeggersee (Switzerland) as assessed by fossil diatom assemblages. *The Holocene* 8, 395–405. <https://doi.org/10.1191/095968398674589725>

Lotter, A.F., Anderson, N.J., 2012. Limnological responses to environmental changes at inter-annual to decadal time-scales, in: Birks, H.J.B., Lotter, A.F., Juggins, S., Smol, J.P. (Eds.), *Tracking Environmental Change Using Lake Sediments Vol 5: Data Handling and Numerical Techniques*. Springer, Dordrecht, pp. 557–578. https://doi.org/10.1007/978-94-007-2745-8_18

Luoto, T.P., Ojala, A.E., 2016. Meteorological validation of chironomids as a paleotemperature proxy using varved lake sediments. *The Holocene* 27, 870–878. <https://doi.org/10.1177/0959683616675940>

Masson-Delmotte, V., Schulz, M., Abe-Ouchi, A., Beer, J., Ganopolski, A., Rouco, J.G., Jansen, E., Lambeck, K., Luterbacher, J., Naish, T., Osborn, T., Otto-Bliesner, B., Quinn, T., Ramesh, R., Rojas, M., Shao, X., Timmermann, A., 2013. *Climate change 2013: The physical science basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change*, in: Stocker, T., Qin, D., Plattner, G.-K., Tignor, M., Allen, S., Boschung, J., Nauels, A., Xia, Y., Bex, V., Midgley, P. (Eds.),. Cambridge

University Press, Cambridge, United Kingdom; New York, NY, USA.

Medeiros, A.S., Friel, C.E., Finkelstein, S.A., Quinlan, R., 2012. A high resolution multi-proxy record of pronounced recent environmental change at Baker Lake, Nunavut. *Journal of Paleolimnology* 47, 661–676. <https://doi.org/10.1007/s10933-012-9589-2>

Nagell, B., Landahl, C.-C., 1978. Resistance to anoxia of *Chironomus plumosus* and *Chironomus anthracinus* (Diptera) larvae. *Holarctic Ecology* 1, 333–336. <http://www.jstor.org/stable/3682403>

Ojala, A., Francus, P., Zolitschka, B., Besonen, M., Lamoureux, S., 2012. Characteristics of sedimentary varve chronologies – a review. *Quaternary Science Reviews* 43, 45–60. <https://doi.org/10.1016/j.quascirev.2012.04.006>

Oksanen, J., Blanchet, F.G., Friendly, M., Kindt, R., Legendre, P., McGlinn, D., Minchin, P.R., O'Hara, R.B., Simpson, G.L., Solymos, P., Stevens, M.H.H., Szoecs, E., Wagner, H., 2019. *Vegan: Community ecology package*. R package version 2.5-5. <https://CRAN.R-project.org/package=vegan>

O'Reilly, C.M., Sharma, S., Gray, D.K., Hampton, S.E., Read, J.S., Rowley, R.J., Schneider, P., Lenters, J.D., McIntyre, P.B., Kraemer, B.M., Weyhenmeyer, G.A., Straile, D., Dong, B., Adrian, R., Allan, M.G., Anneville, O., Arvola, L., Austin, J., Bailey, J.L., Baron, J.S., Brookes, J.D., Eyto, E. de, Dokulil, M.T., Hamilton, D.P., Havens, K., Hetherington, A.L., Higgins, S.N., Hook, S., Izmet'eva, L.R., Joehnk, K.D., Kangur, K., Kasprzak, P., Kumagai, M., Kuusisto, E., Leshkevich, G., Livingstone, D.M., MacIntyre, S., May, L., Melack, J.M., Mueller-Navarra, D.C., Naumenko, M., Noges, P., Noges, T., North, R.P., Plisnier, P.-D., Rigosi, A., Rimmer, A., Rogora, M., Rudstam, L.G., Rusak, J.A., Salmaso, N., Samal, N.R., Schindler, D.E., Schladow, S.G., Schmid, M., Schmidt, S.R., Silow, E., Soylu, M.E., Teubner, K., Verburg, P., Voutilainen, A., Watkinson, A., Williamson, C.E., Zhang, G., 2015. Rapid and highly variable warming of lake surface waters around the globe. *Geophysical Research Letters* 42, 10, 773–10, 781. <https://doi.org/10.1002/2015GL066235>

PAGES2k Consortium, 2017. A global multiproxy database for temperature reconstructions of the common era. *Scientific Data* 4, 170088. <https://doi.org/10.1038/sdata.2017.88>

PAGES2k Consortium, 2013. Continental-scale temperature variability during the past two millennia. *Nature Geoscience* 6, 339–346. <https://doi.org/10.1038/ngeo1797>

Parker, D.E., Legg, T.P., Folland, C.K., 1992. A new daily central England temperature series, 1772–1991. *International Journal of Climatology* 12, 317–342. <https://doi.org/10.1002/joc.3370120402>

Peglar, S.M., 1993. The mid-Holocene *Ulmus* decline at Diss Mere, Norfolk, UK: a year-by-year pollen stratigraphy from annual laminations. *The Holocene* 3, 1–13. <https://doi.org/10.1177/095968369300300101>

Perga, M.-E., Frossard, V., Jenny, J.-P., Alric, B., Arnaud, F., Berthon, V., Black, J., Domaizon, I., Giguet-Covex, C., Kirkham, A., Magny, M., Manca, M., Marchetto, A., Millet, L., Pailès, C., Pignol, C., Poulenard, J., Reyss, J.-L., Rimet, F., Savichtcheva, O., Sabatier, P., Sylvestre, F., Verneaux, V., 2015. High-resolution paleolimnology opens new management perspectives for

lakes adaptation to climate warming. *Frontiers in Ecology and Evolution* 3, 72.
<https://doi.org/10.3389/fevo.2015.00072>

Quinlan, R., Smol, J.P., 2001. Setting minimum head capsule abundance and taxa deletion criteria in chironomid-based inference models. *Journal of Paleolimnology* 26, 327–342. <https://doi.org/10.1023/a:1017546821591>

R Core Team, 2018. R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. <https://www.R-project.org/>

Raubitschek, S., Lücke, A., Schleser, G.H., 1999. Sedimentation patterns of diatoms in Lake Holzmaar, Germany - (on the transfer of climate signals to biogenic silica oxygen isotope proxies). *Journal of Paleolimnology* 21, 437–448. <https://doi.org/10.1023/A:1008022532458>

Rösner, R.R., Müller-Navarra, D.C., Zorita, E., 2012. Trend analysis of weekly temperatures and oxygen concentrations during summer stratification in Lake Plußee: a long-term study. *Limnology and Oceanography* 57, 1479–1491. <https://doi.org/10.4319/lo.2012.57.5.1479>

Saunders, K.M., Grosjean, M., Hodgson, D.A., 2013. A 950 yr temperature reconstruction from Duckhole Lake, southern Tasmania, Australia. *The Holocene* 23, 771–783. <https://doi.org/10.1177/0959683612470176>

Schlesinger, M.E., Ramankutty, N., 1994. An oscillation in the global climate system of period 65–70 years. *Nature* 367, 723–726. <https://doi.org/10.1038/367723a0>

Sharma, S., Gray, D.K., Read, J.S., O'Reilly, C.M., Schneider, P., Quadrat, A., Gries, C., Stefanoff, S., Hampton, S.E., Hook, S., Lenters, J.D., Livingstone, D.M., McIntyre, P.B., Adrian, R., Allan, M.G., Anneville, O., Arvola, L., Austin, J., Bailey, J., Baron, J.S., Brookes, J., Chen, Y., Daly, R., Dokulil, M., Dong, B., Ewing, K., de Eyto, E., Hamilton, D., Havens, K., Haydon, S., Hetzenauer, H., Heneberry, J., Hetherington, A.L., Higgins, S.N., Hixson, E., Izmet'seva, L.R., Jones, B.M., Kangur, K., Kasprzak, P., Köster, O., Kraemer, B.M., Kumagai, M., Kuusisto, E., Leshkevich, G., May, L., MacIntyre, S., Müller-Navarra, D., Naumenko, M., Nöges, P., Nöges, T., Niederhauser, P., North, R.P., Paterson, A.M., Plisnier, P.-D., Rigosi, A., Rimmer, A., Rogora, M., Rudstam, L., Rusak, J.A., Salmaso, N., Samal, N.R., Schindler, D.E., Schladow, G., Schmidt, S.R., Schultz, T., Silow, E.A., Straile, D., Teubner, K., Verburg, P., Voutilainen, A., Watkinson, A., Weyhenmeyer, G.A., Williamson, C.E., Woo, K.H., 2015. A global database of lake surface temperatures collected by in situ and satellite methods from 1985–2009. *Scientific Data* 2, 150008. <https://doi.org/10.1038/sdata.2015.8>

Sheppard, P.R., 2010. Dendroclimatology: Extracting climate from trees. *Wiley Interdisciplinary Reviews: Climate Change* 1, 343–352. <https://doi.org/10.1002/wcc.42>

Simola, H., 1977. Diatom succession in the formation of annually laminated sediment in Lovojärvi, a small eutrophicated lake. *Annales Botanici Fennici* 14, 143–148. <http://www.jstor.org/stable/43922137>

Simonsohn, U., 2013. Just post it: The lesson from two cases of fabricated data detected by statistics alone. *Psychological Science* 24, 1875–1888. <https://doi.org/10.1177/0956797613505555>

972 //doi.org/10.1177/0956797613480366

973 Simpson, G.L., Oksanen, J., 2018. Analogue: Analogue and weighted av-
 974 eraging methods for palaeoecology. R package version 0.17-1. [https://cran.r-](https://cran.r-project.org/package=analogue)
 975 [project.org/package=analogue](https://cran.r-project.org/package=analogue)

976 St. George, S., 2014. An overview of tree-ring width records across the
 977 Northern Hemisphere. *Quaternary Science Reviews* 95, 132–150. [https://doi.](https://doi.org/10.1016/j.quascirev.2014.04.029)
 978 [org/10.1016/j.quascirev.2014.04.029](https://doi.org/10.1016/j.quascirev.2014.04.029)

979 Stewart, M.M., Larocque-Tobler, I., Grosjean, M., 2011. Quantitative inter-
 980 annual and decadal June-July-August temperature variability ca. 570 BC to
 981 AD 120 (Iron Age-Roman Period) reconstructed from the varved sediments
 982 of Lake Silvaplana, Switzerland. *Journal of Quaternary Science* 26, 491–501.
 983 <https://doi.org/10.1002/jqs.1480>

984 Telford, R., 2019. PalaeoSig: Significance tests of quantitative palaeoenvi-
 985 ronmental reconstructions. R package version 2.0-3. [https://cran.r-project.org/](https://cran.r-project.org/package=palaeoSig)
 986 [package=palaeoSig](https://cran.r-project.org/package=palaeoSig)

987 Telford, R.J., Birks, H.J.B., 2011. A novel method for assessing the statistical
 988 significance of quantitative reconstructions inferred from biotic assemblages.
 989 *Quaternary Science Reviews* 30, 1272–1278. [https://doi.org/10.1016/j.quascirev.](https://doi.org/10.1016/j.quascirev.2011.03.002)
 990 [2011.03.002](https://doi.org/10.1016/j.quascirev.2011.03.002)

991 Telford, R.J., Birks, H.J.B., 2009. Evaluation of transfer functions in spatially
 992 structured environments. *Quaternary Science Reviews* 28, 1309–1316. [https:](https://doi.org/10.1016/j.quascirev.2008.12.020)
 993 [//doi.org/10.1016/j.quascirev.2008.12.020](https://doi.org/10.1016/j.quascirev.2008.12.020)

994 Telford, R.J., Birks, H.J.B., 2005. The secret assumption of transfer func-
 995 tions: Problems with spatial autocorrelation in evaluating model performance.
 996 *Quaternary Science Reviews* 24, 2173–2179. [https://doi.org/10.1016/j.quascirev.](https://doi.org/10.1016/j.quascirev.2005.05.001)
 997 [2005.05.001](https://doi.org/10.1016/j.quascirev.2005.05.001)

998 Thomas, E.K., Briner, J.P., 2009. Climate of the past millennium inferred
 999 from varved proglacial lake sediments on northeast Baffin Island, Arctic Canada.
 1000 *Journal of Paleolimnology* 41, 209–224. [https://doi.org/10.1007/s10933-008-](https://doi.org/10.1007/s10933-008-9258-7)
 1001 [9258-7](https://doi.org/10.1007/s10933-008-9258-7)

1002 Tilman, D., 1981. Tests of resource competition theory using four species of
 1003 Lake Michigan algae. *Ecology* 62, 802–815. <https://doi.org/10.2307/1937747>

1004 Tokeshi, M., 1995. Life cycles and population dynamics, in: Armitage,
 1005 P.D., Cranston, P.S., Pinder, L.C.V. (Eds.), *The Chironomidae: Biology and*
 1006 *Ecology of Non-Biting Midges*. Springer, Dordrecht, pp. 225–268. [https:](https://doi.org/10.1007/978-94-011-0715-0_10)
 1007 [//doi.org/10.1007/978-94-011-0715-0_10](https://doi.org/10.1007/978-94-011-0715-0_10)

1008 Trachsel, M., Grosjean, M., Schnyder, D., Kamenik, C., Rein, B., 2010.
 1009 Scanning reflectance spectroscopy (380–730 nm): A novel method for quantitative
 1010 high-resolution climate reconstructions from minerogenic lake sediments. *Journal*
 1011 *of Paleolimnology* 44, 979–994. <https://doi.org/10.1007/s10933-010-9468-7>

1012 Trachsel, M., Kamenik, C., Grosjean, M., McCarroll, D., Moberg, A., Brázdil,
 1013 R., Büntgen, U., Dobrovolný, P., Esper, J., Frank, D.C., Friedrich, M., Glaser,
 1014 R., Larocque-Tobler, I., Nicolussi, K., Riemann, D., 2012. Multi-archive summer
 1015 temperature reconstruction for the European Alps, AD 1053–1996. *Quaternary*
 1016 *Science Reviews* 46, 66–79. <https://doi.org/10.1016/j.quascirev.2012.04.021>

1017 Trachsel, M., Telford, R.J., 2017. All age-depth models are wrong, but are get-

ting better. *The Holocene* 27, 860–869. <https://doi.org/10.1177/0959683616675939>

Tylmann, W., Bonk, A., Goslar, T., Wulf, S., Grosjean, M., 2016. Calibrating Pb-210 dating results with varve chronology and independent chronostratigraphic markers: problems and implications. *Quaternary Geochronology* 32, 1–10. <https://doi.org/10.1016/j.quageo.2015.11.004>

van der Knaap, W.O., van Leeuwen, J.F.N., Svitavská-Svobodová, H., Pidek, I.A., Kvavadze, E., Chichinadze, M., Giesecke, T., Kaszewski, B.M., Oberli, F., Kalnina, L., Pardoe, H.S., Tinner, W., Ammann, B., 2010. Annual pollen traps reveal the complexity of climatic control on pollen productivity in Europe and the Caucasus. *Vegetation History and Archaeobotany* 19, 285–307. <https://doi.org/10.1007/s00334-010-0250-6>

van Hardenbroek, M., Heiri, O., Wilhelm, M.F., Lotter, A.F., 2011. How representative are subfossil assemblages of Chironomidae and common benthic invertebrates for the living fauna of Lake De Waay, the Netherlands? *Aquatic Sciences* 73, 247–259. <https://doi.org/10.1007/s00027-010-0173-4>

Velle, G., Brodersen, K.P., Birks, H.J.B., Willassen, E., 2010. Midges as quantitative temperature indicator species: lessons for palaeoecology. *The Holocene* 20, 989–1002. <https://doi.org/10.1177/0959683610365933>

von Gunten, L., Grosjean, M., Kamenik, C., Fujak, M., Urrutia, R., 2012. Calibrating biogeochemical and physical climate proxies from non-varved lake sediments with meteorological data: methods and case studies. *Journal of Paleolimnology* 47, 583–600. <https://doi.org/10.1007/s10933-012-9582-9>

von Gunten, L., Grosjean, M., Rein, B., Urrutia, R., Appleby, P., 2009. A quantitative high-resolution summer temperature reconstruction based on sedimentary pigments from Laguna Aculeo, central Chile, back to AD 850. *The Holocene* 19, 873–881. <https://doi.org/10.1177/0959683609336573>

Wacnik, A., Tylmann, W., Bonk, A., Goslar, T., Enters, D., Meyer-Jacob, C., Grosjean, M., 2016. Determining the responses of vegetation to natural processes and human impacts in north-eastern Poland during the last millennium: combined pollen, geochemical and historical data. *Vegetation History and Archaeobotany* 25, 479–498. <https://doi.org/10.1007/s00334-016-0565-z>

Werner, J.P., Tingley, M.P., 2015. Technical note: Probabilistically constraining proxy age–depth models within a Bayesian hierarchical reconstruction model. *Climate of the Past* 11, 533–545. <https://doi.org/10.5194/cp-11-533-2015>

Witak, M., Hernández-Almeida, I., Grosjean, M., Tylmann, W., 2017. Diatom-based reconstruction of trophic status changes recorded in varved sediments of Lake Żabińskie (northeastern Poland), AD 1888–2010. *Oceanological and Hydrobiological Studies* 46, 1–17. <https://doi.org/10.1515/ohs-2017-0001>

Zhang, E., Chang, J., Cao, Y., Tang, H., Langdon, P., Shulmeister, J., Wang, R., Yang, X., Shen, J., 2017a. A chironomid-based mean July temperature inference model from the south-east margin of the Tibetan Plateau, China. *Climate of the Past* 13, 185–199. <https://doi.org/10.5194/cp-13-185-2017>

Zhang, E., Chang, J., Cao, Y., Tang, H., Langdon, P., Shulmeister, J., Wang, R., Yang, X., Shen, J., 2017b. Supplementary information for: A chironomid-based mean July temperature inference model from the south-east margin of the Tibetan Plateau, China. [WWW Document]. <https://www.dropbox.com/s/>

1064 0z2sl6zwzj69f7h/Zhang%20et%20al%202017__Climate%20of%20the%20Past__dataset.
1065 xlsx?dl=0

Supplementary Data for ‘Review and test of reproducibility of subdecadal resolution palaeoenvironmental reconstructions from microfossil assemblages’

Richard J. Telford

13 August 2019

Lake Żabińskie

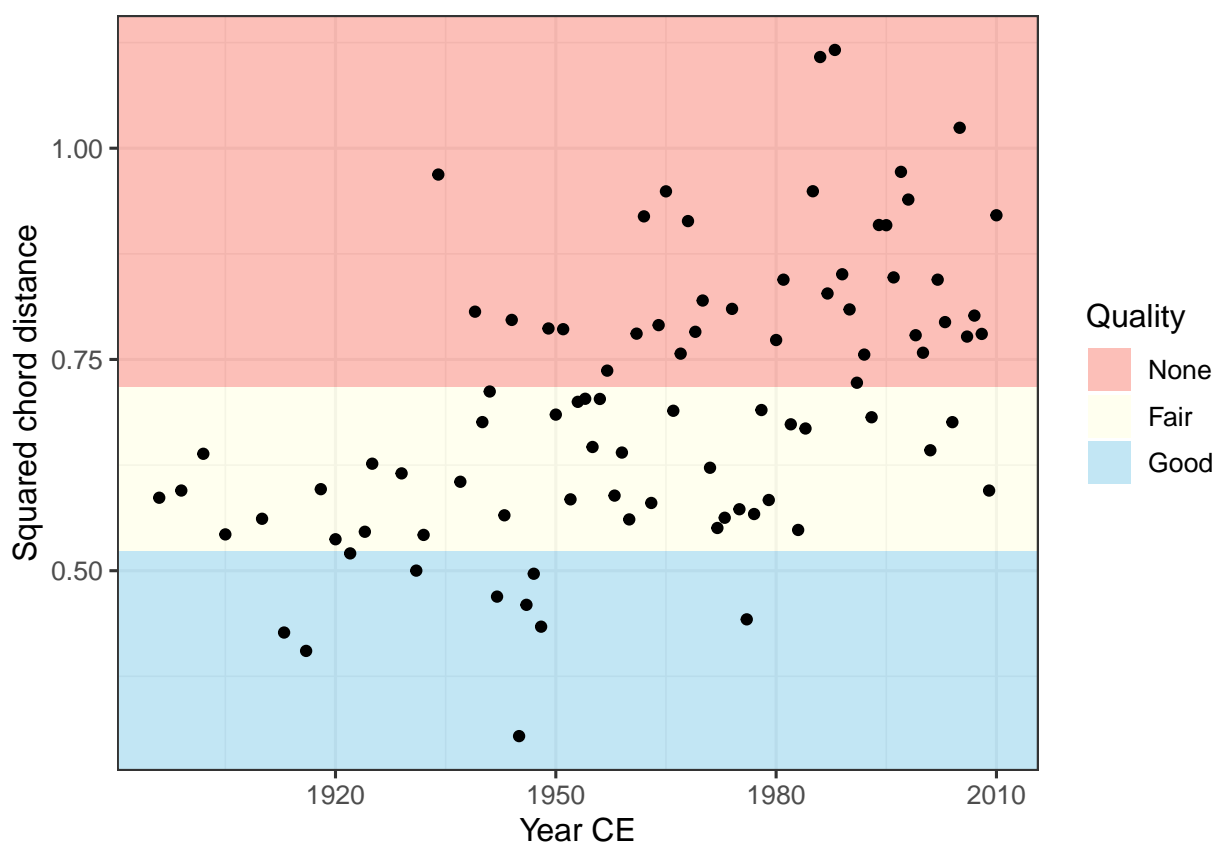


Figure S1: Squared chord distance from Lake Żabińskie fossil samples to nearest analogue in the Canadian-Polish calibration set. Analogue quality indicated by background colour estimated with the methods and thresholds used in the original paper. Calibration set samples with low count sums are omitted, but the results are similar if they are included.

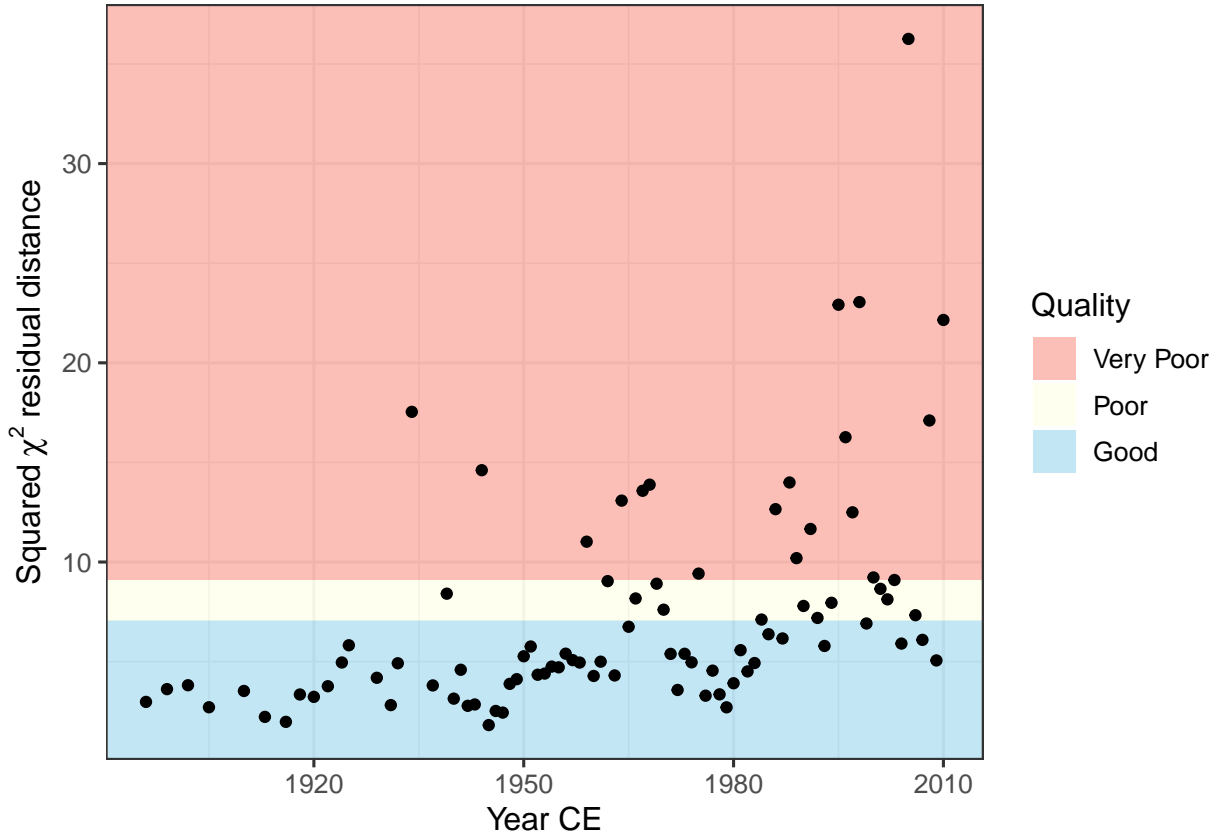


Figure S2: Goodness of fit of the Lake Żabińskie fossil samples added passively to a canonical correspondence analysis of the Canadian-Polish calibration set constrained by temperature.

Abisko lakes

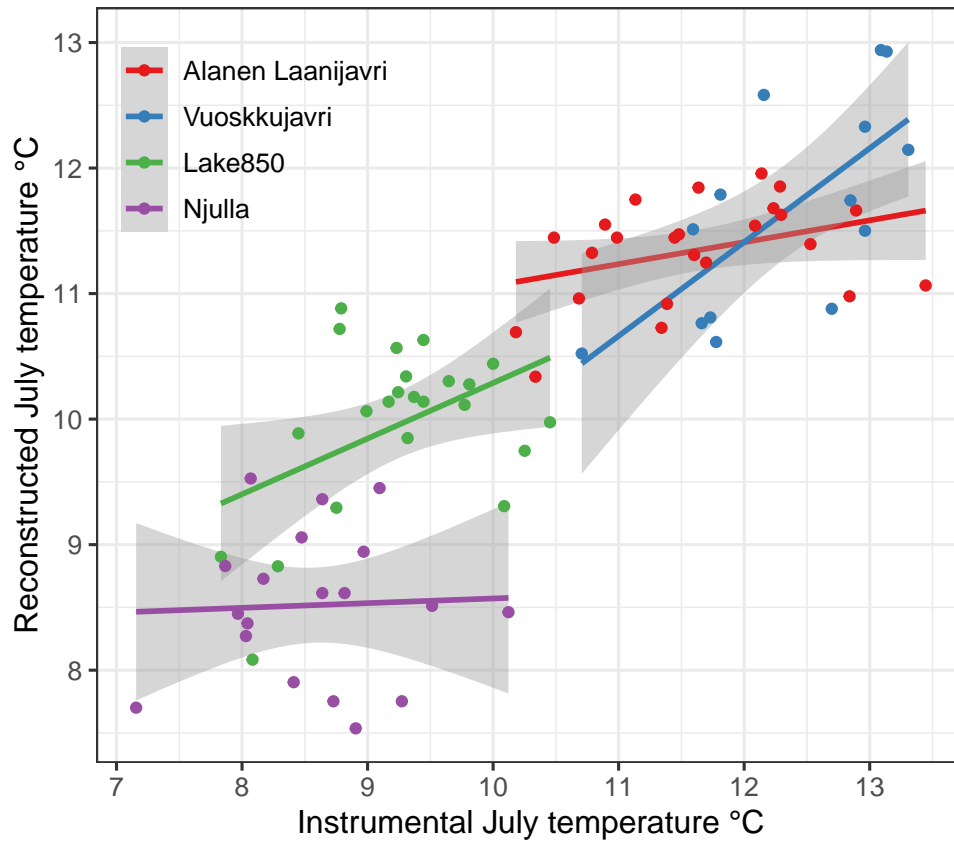


Figure S3: Twentieth century reconstructed and instrumental July air temperatures for four lakes near Abisko. Data digitised from the original paper which reported that all the lakes had a similar correlation between the reconstructed and instrumental record.

Lake Silvaplana

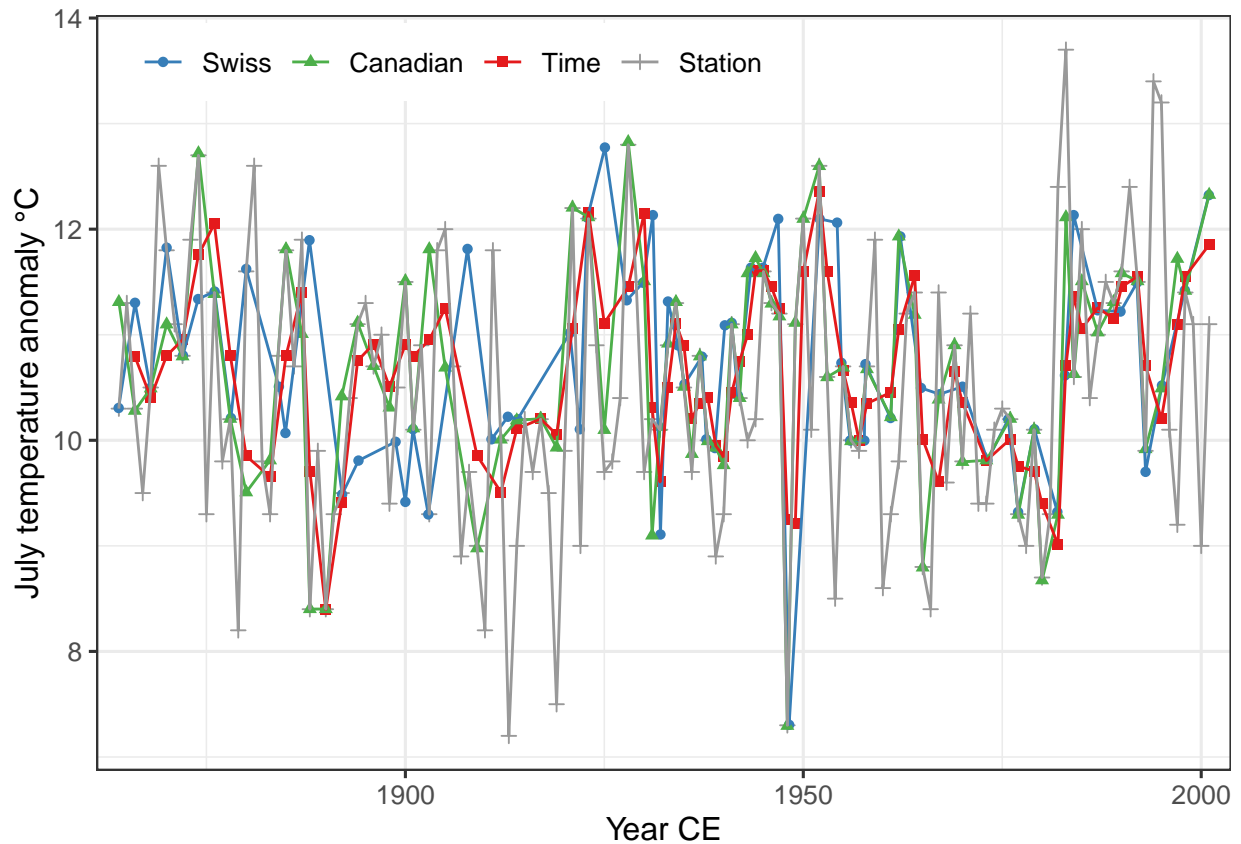


Figure S4: Instrumental July air temperatures from Selg-Maria (Begert et al., 2005) and the instrumental data as digitised from different papers. Digitised data are labelled after the calibration set used to make the reconstruction against which the instrumental data were compared. All the digitised instrumental time series should be identical, unless fossil samples have been merged/unmerged.

Long Swiss climate series were homogenised by Berget et al. (2005). They were reprocessed in 2009 (MeteoSwiss, 2010). The Silvaplana papers probably all used the version of the Segl Maria series by Berget et al. (2005), which is shown here. The differences between these data and the current version of the data are minimal ($r^2 = 0.997$).

Seebergsee

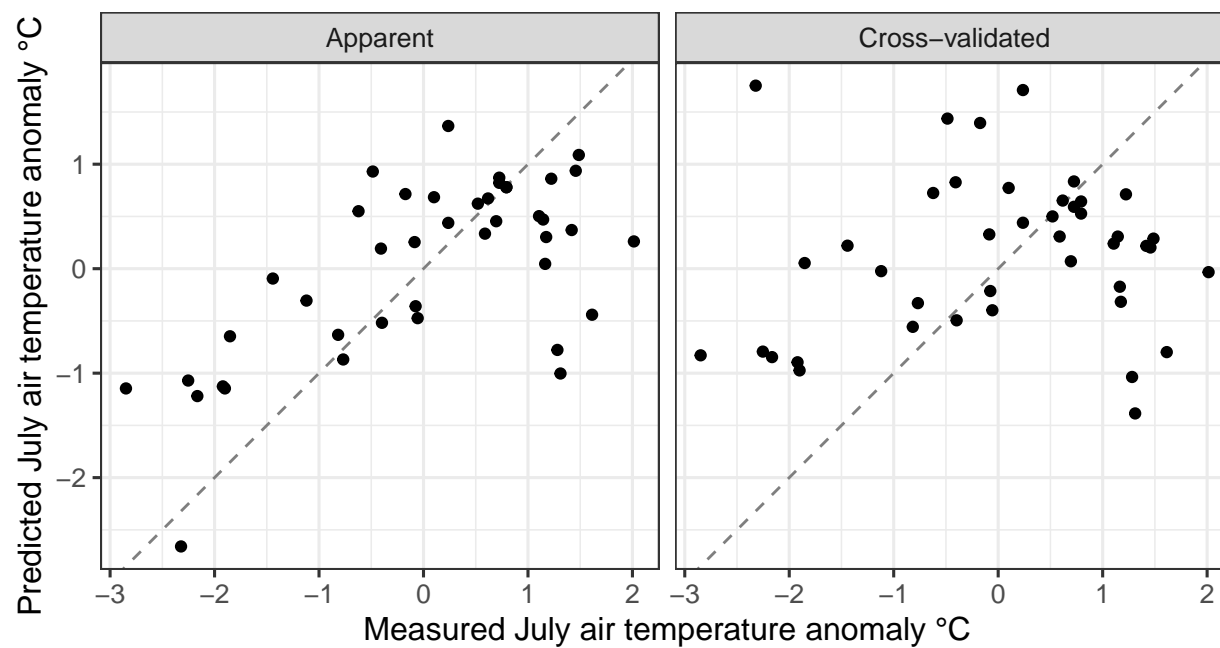


Figure S5: Apparent and cross-validated performance of the calibration-in-time WAPLS-2 model from Seebergsee.

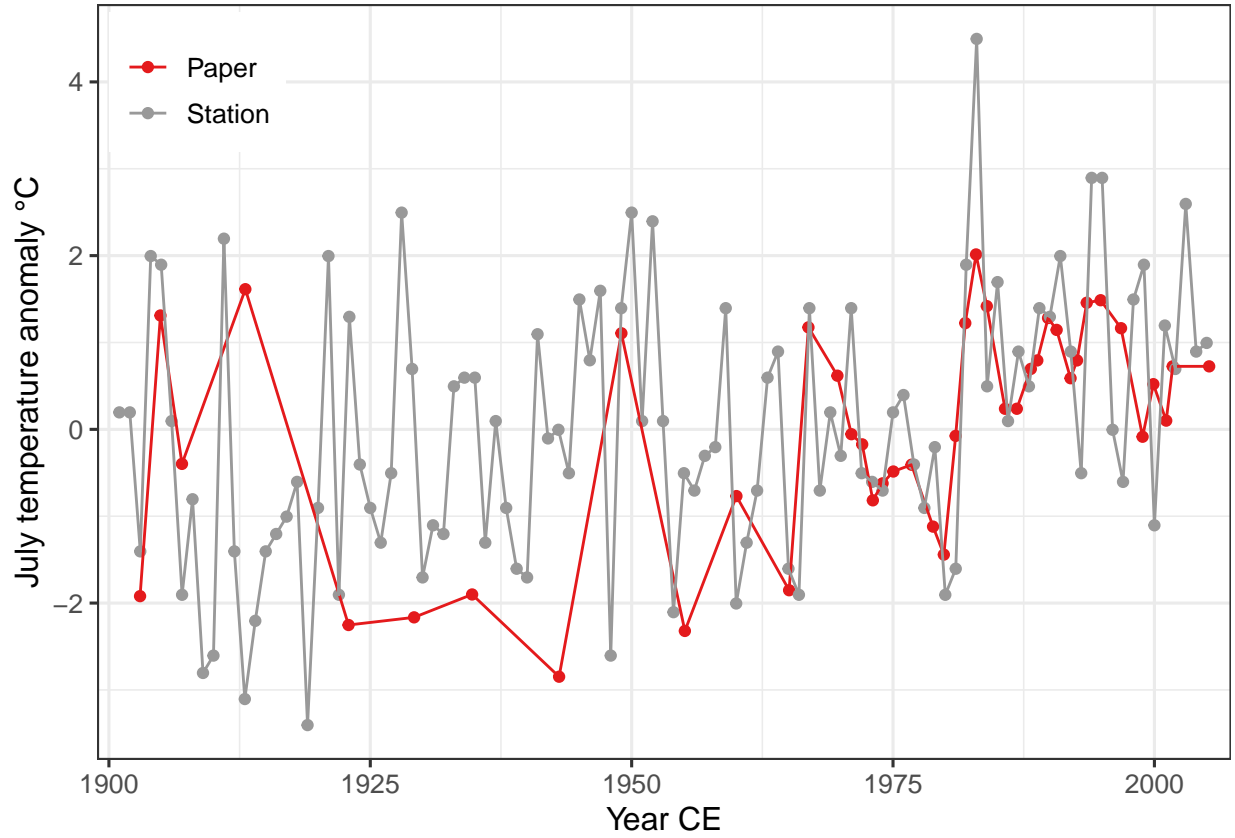


Figure S6: Instrumental July air temperature anomalies from Château-d'Oex (MeteoSwiss, 2010) and the instrumental data as digitised from the Seebergsee paper.

There may be, probably small, differences between the data shown in here and the version of the Château-d'Oex temperature series homogenised by Berget et al. (2005).

Nurmijärvi

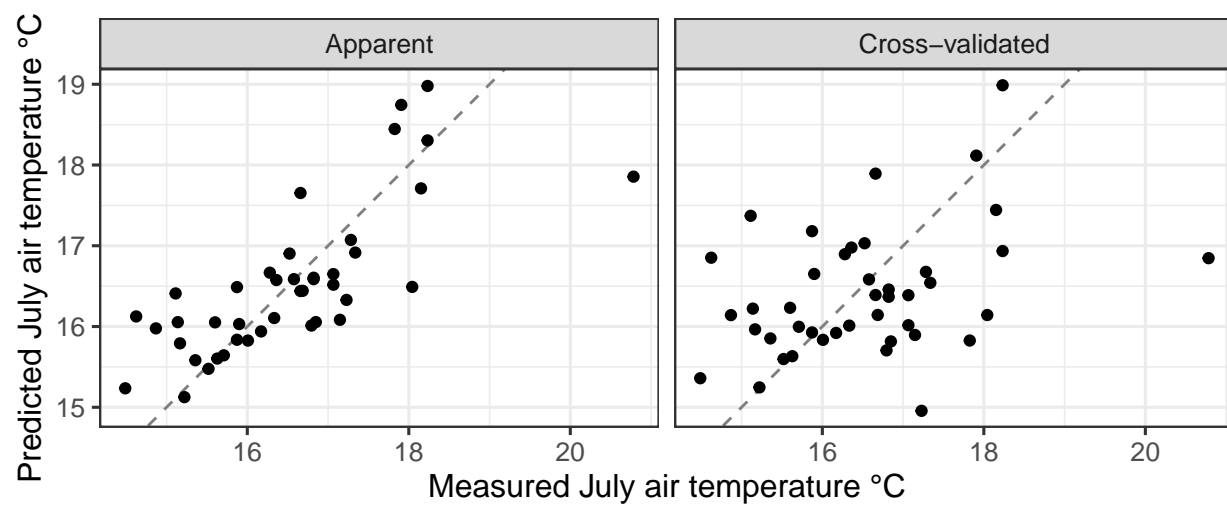


Figure S7: Apparent and cross-validated performance of the calibration-in-time tolerance-weighted weighted averaging model from Nurmijärvi.

Speke Hall Lake

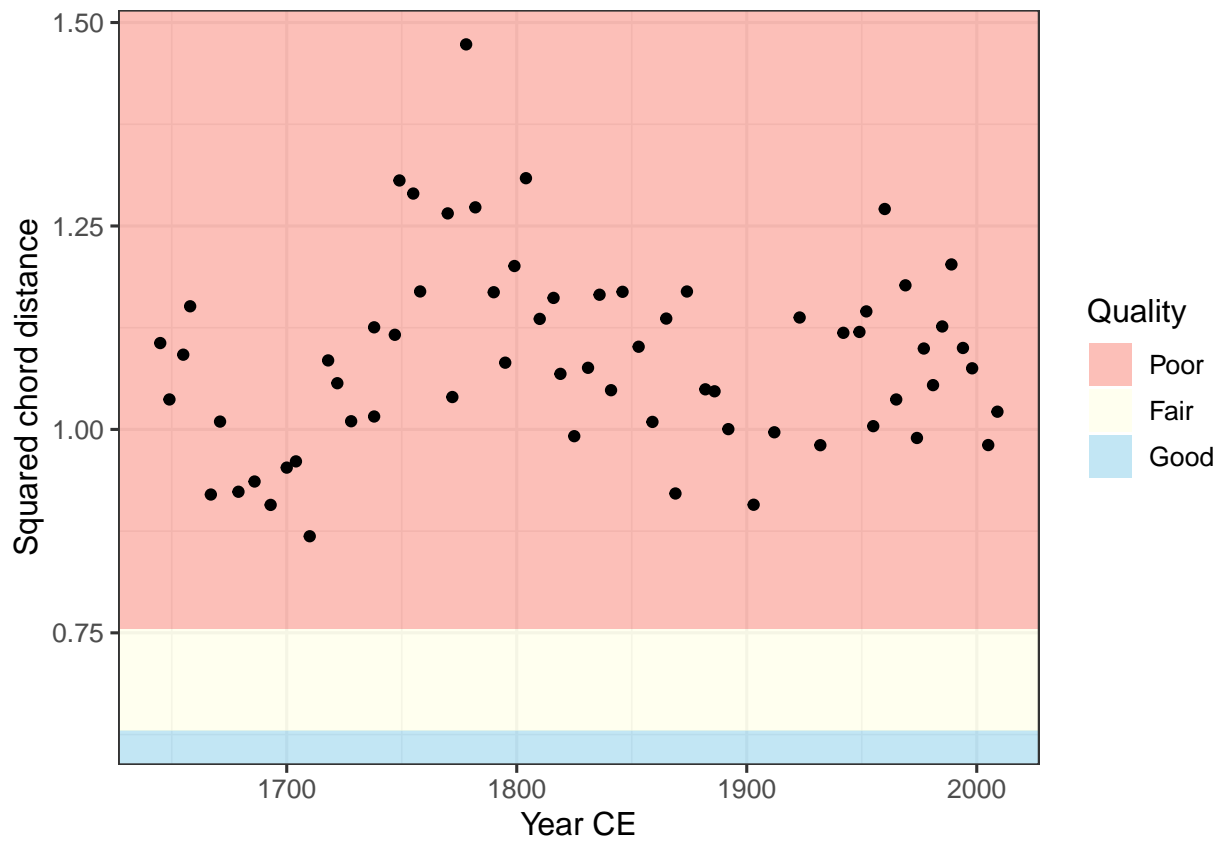


Figure S8: Squared chord distance from Speke Hall Lake fossil samples to nearest analogue in the Norwegian calibration set. Analogue quality indicated by background colour. Thresholds derived from quantiles of all calibration set distances (Good < 0.05, Fair < 0.1).

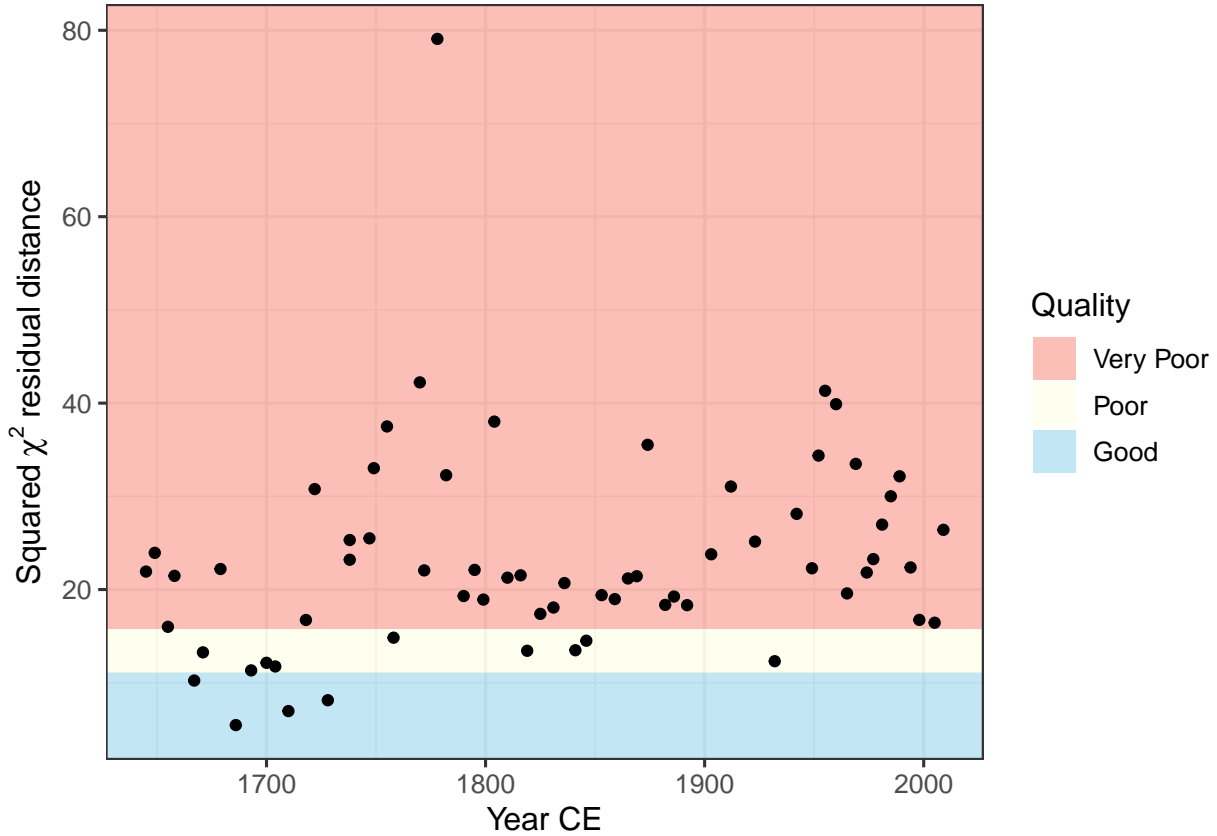


Figure S9: Goodness of fit of the Speke Hall Lake fossil samples added passively to a CCA of the Norwegian calibration set constrained by temperature. Thresholds derived from quantiles of all calibration set squared residual lengths (Good < 0.9, Fair < 0.95). An equivalent figure in the original paper used an inappropriate data tranformation prior to the analysis.

References

- Begert, M., Schlegel, T., Kirchhofer, W., 2005. Homogeneous temperature and precipitation series of Switzerland from 1864 to 2000. *International Journal of Climatology* 25, 65–80. <https://doi.org/10.1002/joc.1118>
- MeteoSwiss, 2010. Originale und homogene reihen im vergleich. <https://www.meteoswiss.admin.ch/content/dam/meteoswiss/de/klima/klima-im-detail/doc/klima-vergleich-original-homogen.pdf>