The challenge of comparing pollen-based quantitative vegetation reconstructions with outputs from vegetation models – a European perspective

Anne Dallmeyer[1], Anneli Poska[2,3], Laurent Marquer[4], Andrea Seim[4,5], Marie-José Gaillard ~~Lemdahl~~[6]

[1] Max Planck Institute for Meteorology, Bundesstrasse 53, 20146 Hamburg, Germany

[2] Department of Geology, Tallinn University of Technology, Ehitajate tee 5, 19086 Tallinn Estonia

[3] Department of Physical Geography and Ecosystem Science, Lund University, Sölvegatan 12, 223 62 Lund, Sweden

[4] Department of Botany, University of Innsbruck, Sternwartestrasse 15, 6020 Innsbruck, Austria

[5] Chair of Forest Growth and Dendroecology, Institute of Forest Sciences, ~~Albert-Ludwig-~~University of Freiburg, 79106 Freiburg, Germany

[6] Department of Biology and Environmental Science, Linnaeus University, Barlastgatan 11, 39182 Kalmar, Sweden

Correspondence to: Anne Dallmeyer (anne.dallmeyer@mpimet.mpg.de)

Abstract

We compare Holocene tree-cover changes in Europe derived from a transient <u>Earth System model</u> ~~MPI-ESM1.2~~ simulation <u>(MPI-ESM1.2, including the land surface and dynamic vegetation model JSBACH)</u> with high spatial resolution ~~LPJ-GUESS~~ time-slice simulations <u>performed in the dynamic vegetation model LPJ-GUESS</u> and pollen-based quantitative reconstructions of tree cover based on the REVEALS model. The dynamic vegetation models and REVEALS agree with respect to the general temporal trends in tree cover for most parts of Europe, with a large tree cover during the mid-Holocene

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and a substantially smaller tree cover closer to the present time. However, the decrease in tree cover in REVEALS starts much earlier than in the models indicating much earlier anthropogenic deforestation than the prescribed land -use in the models. While LPJ-GUESS generally overestimates tree cover compared to the reconstructions, MPI-ESM indicates lower percentages of tree cover than REVEALS, particularly in Central Europe and the British Isles. A comparison of the simulated climate with chironomid-based climate reconstructions reveals that model-data mismatches in tree cover are in most cases not driven by biases in the climate. Instead, sensitivity experiments indicate that the model results strongly depend on the tuning of the models regarding natural disturbance regimes (e.g. fire and wind throw). The frequency and strength of disturbances are – like most of the parameters in the vegetation models – static and calibrated to modern conditions. However, these parameter values may not be valid ~~during~~ for past climate and vegetation states totally different from today's~~s~~. In particular, the mid-Holocene natural forests were probably more stable and less sensitive to disturbances than present day forests that are heavily altered by human interventions. Our analysis highlights the fact that such model settings are inappropriate for palaeo-simulations and complicate model-data comparisons with additional challenges. Moreover, our study suggests that land -use is the main driver of forest decline in Europe during the mid- and late-Holocene.

1 Introduction

Terrestrial land cover is one of the key components of the Earth's ecosystem and a provider of many ecosystem services. It is widely discussed in the context of ongoing climate change, due to its high sensitivity to environmental changes and its role as one of the mitigation agents of the current and projected global warming (e.g. Williamson, 2016; Harper et al., 2018; Smith et al., 2016). Decisions on strategies for the future depend, among others, on our ability to correctly understand the interactions between vegetation and climate over short and millennial timescales. This also requires ~~also~~ that Earth System Models (ESMs) correctly simulate these interactions in the past to ensure reliable model projections (Harrison et al., 2020). In this context, Dynamic Global Vegetation Models (DGVMs) are used, either coupled to ESMs or offline, to simulate past or future climate- and human-induced changes in land-cover composition, biomass production, and carbon storage capacity (e.g. Hickler et al., 2012; Wramneby et al., 2010; Hopcroft et al., 2017; Lu et al., 2018, 2021). However, the DGVM parametrization (bioclimatic limits, disturbance intervals, fire regimes, etc.) are commonly static and based on the current state of land cover although it is characterized by unstable vegetation composition due to rapidly changing natural and anthropogenic stressors (Hengl et al., 2018). This is one of several caveats of DGVMs that may lead to erroneous projections for the future.

Comparison of DGVM simulations for the past with proxy records -of- ~~Holocene~~ vegetation composition, is a way to evaluate the performance of DGVMs. Among existing empirical proxies of past vegetation, pollen records from lake sediments or peat deposits have the best potential for quantitative reconstructions of plant abundance or spatial cover.

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However, pollen records are subject to several shortcomings such as unknown size of source area of pollen and differences in pollen productivity and dispersal properties between plant taxa (e.g. Prentice, 1985). These issues imply that pollen percentages (and pollen accumulation rates) from fossil pollen assemblages can only provide qualitative or semi-quantitative information on past vegetation changes, i.e. sporadic or regular presence, occurrence in more or less large quantities, and increases and decreases in abundance of plant taxa. Different methods have been developed to overcome these problems and reconstruct plant cover, e.g. biomization, the modern analogue technique (MAT), pseudo-biomization, and the landscape reconstruction algorithm (REVEALS and LOVE models). These methods are described and evaluated in e.g. Hellman et al. (2008a) and Roberts et al. (2018).

The model REVEALS (Regional Estimates of Vegetation Abundance from Large Sites (Sugita, 2007) is the only method so far that accounts for inter-taxonomic differences in pollen productivity and dispersal and deposition properties and provides estimates of plant cover (in % cover of a defined area) for individual taxa. In recent years, datasets of pollen-based REVEALS plant cover were produced at a 1° grid-cell spatial scale for large regions of the world, i.e. Europe, China and N America-Canada (Trondman et al., 2015; Marquer et al., 2017; Dawson et al., 2018; Cao et al., 2019; Githumbi et al. 2022a, Li et al, 2023; Serge et al., 2023). These datasets are appropriate for use in paleoclimate modelling (e.g. Strandberg et al., 2021, 2023) and evaluation of dynamic vegetation models (e.g. Marquer et al., 2017; 2018) and scenarios of anthropogenic land-cover change (e.g. Kaplan et al., 2017).

Europe is a good test bed for model evaluation because it offers a dense network of pollen records and thus, many one of the most intensively studied areas of the world in terms of number and density of pollen records and number of pollen-based reconstructions of regional plant cover at semi-continental and continental scales (Trondman et al., 2016 5; Marquer et al., 2014; 2017; Githumbi et al., 2022a; Serge et al., 2023 Dawson et al., 2018). Moreover, these reconstructions were successfully combined with auxiliary datasets (four covariates: latitude, longitude, elevation, independent scenarios of past deforestation) to create spatially continuous maps for 6 ka (Pirzamanbein et al., 2014; 2020) and continuous time windows from 11.7 ka BP to present (Githumbi et al., 2022b) using spatial statistical models.

Pollen-based vegetation reconstructions analyses indicate large changes in plant-species composition and distribution over the Holocene. They represent both natural and human (land use)-induced changes, the latter increasing gradually from 6 ka BP until today. These land-use relatedSuch human-induced land-cover changes could potentially have a large impact on the complex climate-vegetation interactions and represent a significant climate forcing in the past (Ruddiman et al., 2015; Ruddiman, 2007; Boy et al., 2022; Huang et al., 2020). Regional climate-model simulations have shown that the anthropogenic deforestation of Europe at 6ka BP according to the KK10 scenarios (Kaplan et al., 2009) and the pollen-based land-cover reconstructions of Githumbi et al. (2022a) result in regional cooling or warming of 1 °C depending on the region and season (Strandberg et al., 2014 and 2022; Strandberg et al., 2014)

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An eEarlier evaluations of the performance of the DGVM LPJ-GUESSs by comparing model-simulated land cover with pollen-based plant-cover reconstructions in Europe hasve shown clear differences between the two for the ~~Early~~ first one to two millennia of the Holocene and over the last 7000-6000 years (e.g. Marquer et al., 2017; 2018). The largest discrepancies are found in the abundance/cover of open land, with LPJ-GUESSmodels generally underestimating the extent of unforested land. While ~~these~~ model-data mismatches are commonly associated with biases in climate inputs (Strandberg et al., 2022), an increasing amount of evidence shows good conformity between climate--model outputs and climate reconstructions inferred from other proxies than pollen. It implies that mismatches are rather related to the lagged reaction of trees to climate change (Dallmeyer et al., 2022) and the increasing effect of anthropogenic land-cover change in Europe from 6000 years ago (Kleinen et al., 2011; Braconnot et al., 2019).

The aim of this study is to ~~analyse the Holocene vegetation change in Europe with the focus on~~ exploreing the mechanisms behind discrepancies between DGVM-simulated and empirically reconstructed (pollen based) plant abundance andvegetation distributions. We compare tree--cover changes simulated by two commonly used DGVMs with different inherent vegetation representation and parametrization, the MPI-ESM1.2 land-model component JSBACH and the DGVM LPJ-GUESS, with empirical pollen-based REVEALS plant-cover reconstructions for six time windows of the Holocene and five areas along S-N and W-E transects through central and northern Europe. Both model-simulated and pollen-reconstructed plant cover is expressed in percentage cover of a known area. Henceforth, all ages are given in calibrated ^{14}C kilo years BP, abbreviated "ka".

2 Methods

In this study, we use DGVMs with very different modelling approaches. JSBACH is the land-surface component of the comprehensive Earth System model MPI-ESM, and thus calculates interactively the grid-cell cover fractions of different plant functional types (PFTs) in line with the simulated climate (Reick et al., 2021). In contrast, LPJ-GUESS is a so-called "gap" model that calculates the growth of individuals or cohorts in patches, according to a prescribed climatic forcing (Smith et al., 2001). While we explore one of the few worldwide existing transient Holocene Earth System model simulations (MPI-ESM), the high spatial resolution time-slice experiments in LPJ-GUESS have been conducted in this study. LPJ-GUESS is used here because it is well-tested for the European domain (e.g. Hickler et al., 2012). The results of both models are compared to the pollen-based REVEALS reconstructions of plant cover for Europe from Marquer et al. (2017). These reconstructions were chosen before others based on other methods such as MAT (e.g. Davis et al., 2015) or pseudobiomization (e.g. Fyfe et al., 2010) because the REVEALS model has been shown to be the best approach to produce reliable quantitative reconstructions of forest cover and the only method available to date to reconstruct the cover of individual plant taxa (e.g. Hellman et al., 2008a; Roberts et al., 2018; see also Introduction above). The methodological

strategy used in this study is summarised in the flow chart of Figure 1. Figure 1 summarises the strategy for the comparison between the Dynamic Global Vegetation Models (DGVMs) JSBACH (as interactive component of the Earth System Model MPI-ESM1.2) and the offline DGVM LPJ-GUESS, and the comparison between the DGVMs and the pollen-based plant-cover reconstructions using the REVEALS model. The different models, simulations and methods involved are described in detail below.

2.1 The Earth System Model MPI-ESM1.2 transient simulation

2.1.1 Model description

We use the existing transient simulation for the period 7.95 – 0.1 ka conducted by Bader et al. (2020) in the Max-Planck-Institute Earth System Model The model MPI-ESM (version 1.2) (Mauritsen et al., 2019). This simulation has been forced by changes in insolation (Berger, 1978), greenhouse-gas concentrations (F. Joos, personal communication; see Köhler, 2019 and Brovkin et al., 2019), stratospheric sulphate aerosol injections imitating volcanic eruptions (Toohey and Sigl, 2017), spectral solar irradiance (Krivova et al., 2011) and human-induced land cover changes (preliminary version of Hurtt et al., 2020). The atmosphere and land model were applied with a spectral resolution of T63 (approx. 200km on a Gaussian grid). While the simulation and forcing mechanisms were described in detail in earlier studies (Bader et al., 2020; Brovkin et al., 2019; Dallmeyer et al., 2020), we focus here only on the dynamic vegetation calculation and the land-use forcing.

consists of the general circulation model of the ocean MPIOM (Jungclaus et al., 2013) coupled to the atmospheric general circulation model ECHAM6.3 (Stevens et al., 2013). MPIOM includes the global ocean biogeochemistry model HAMOCC (Ilyina et al., 2013). Vegetation and terrestrial carbon-cycle dynamics are calculated by the land-surface scheme JSBACH3 (Reick et al., 2021 and 2013), incorporated in ECHAM6.3. JSBACH3 includes the soil carbon model YASSO (Goll et al., 2015), a 5-layer hydrology scheme (Hagemann and Stacke, 2015) and MPI-ESM includes the land-surface model JSBACH with the dynamic vegetation module developed by Brovkin et al. (2009). In this module, natural vegetation is represented by eight different plant functional types (PFTs). Trees (four PFTs) can either be tropical or extratropical (i.e. boreal + temperate) temperate, and evergreen or deciduous. The open land cover is represented by two herbaceous PFTs (C3 and C4 grass) and two shrub PFTs (Grassy types are distinguished in C3 and C4 grass and the last two types represent raingreen shrubs and cold resistant shrubs). Furthermore, three anthropogenic Lland -use (anthropogenic vegetation) types is are included as three PFTs (i.e. C3 pasture, and C4 pasture, and crops). Therefore, total plant cover is represented by eleven PFTs (see details on the implementation of anthropogenic PFTs below). Different PFTs can coexist in each grid cell as the model uses a tiling approach, i.e. the grid cell is tiled in mosaics of fractional PFT coverages. The establishment of each natural PFT is constrained by temperature thresholds representing their respective bioclimatic tolerance. The fractional cover

of each PFT is, by and large, determined by the relative differences in annual net primary productivity (NPP) between the PFTs. Natural mortality and disturbances, such as wind throw and fire, reduce the cover fraction of PFTs. Woody PFTs are generally favoured at the expense of grass, but in regions with frequent disturbances or bioclimatic conditions near the bioclimatic thresholds, shrubs or even grass may win the competition as they can recover more quickly than trees. The relative presence of grasses and woody PFTs is thus implicitly determined by the strength of the disturbances. While fire has a different effect on grass than on woody PFTs, wind throw only reduces the woody PFT types. The disturbance rate of wind throw is proportional to the simulated wind power, but it is weighted by the averaged wind speed in each grid cell to account for the adaptation of woody plants to local wind conditions. In addition, wind throw is set to zero if wind speeds are below a threshold that is determined by the long-term average maximum wind speed.

For each grid cell, JSBACH calculates the fraction of the grid-cell that is not covered by vegetation (bare soil fraction) that represents both seasonal and permanently unvegetated ground. Since the area of bare soil cannot be estimated by the REVEALS model (see. 2.3), the PFT fractions in this study are scaled based on the total area covered by vegetation, i.e. the bare soil fraction is not considered and the cover fractions of the PFTs are adjusted to sum up to 1one in each grid cell.

Human-induced land-cover changes affect the simulation only for the last ~2000 years. In the simulation, the anthropogenic land-cover changes have been prescribed from a preliminary version of the LUH2 dataset (Hurtt et al., 2020). This dataset provides the land-use changes for the 850CE to 2100CE period and could therefore only be used as forcing of the model for the last 1000 years of the transient simulation (from 1.1 ka on). To slowly build up the land use in the course of the simulation, from zero to the anthropogenic influenced land cover at 1.1 ka, a transition period of 1000 years has been implemented starting at 2.1 ka. Land use is read annually but calculated daily by linear interpolation. Land use is prescribed in the form of transition maps that define the fraction of area that is converted from natural vegetation to crops and pasture or vice versa. Pasture is first distributed in the area covered by grass before it replaces forested area (Reick et al., 2021). After this rule has been applied, the remaining anthropogenic land-cover change is equally distributed to all PFTs, relative to their individual cover fractions, in order to have the same gain or loss of cover fraction.

More details on the dynamic vegetation module can be found in Brovkin et al. (2009) and; Reick et al. (2021 and 2013). An evaluation of the simulated pre-industrial (PI) climate is provided in Appendix A.

2.1.2 Transient simulation in MPI-ESM1.2

170 After having been run in quasi-equilibrium for mid-Holocene (7950 BP) boundary conditions, the model MPI-ESM1.2 was used to produce a transient simulation for the period 7950 - 100 BP (Bader et al., 2020; Dallmeyer et al., 2020). The atmosphere and land model was applied with a spectral resolution of T63 (approx. 200km on a Gaussian grid) with 47 levels in the vertical. The ocean model has been configured in the horizontal resolution GR15 (i.e. 256x220 on a bipolar grid, 12 to 180km) with 64 vertical levels. The transient simulation was performed using the following forcings:

175   a)   orbital-induced insolation changes (Berger, 1978), updated every decade

       b)   Methane, nitrous oxide, and carbon dioxide concentrations inferred from ice core records (F. Joos, personal communication; see Köhler, 2019 and Brovkin et al., 2019), updated every decade

       c)   stratospheric sulphate aerosol injections imitating volcanic eruptions, prescribed from the Easy Volcanic Aerosol (EVA) forcing generator (Toohey and Sigl, 2017), read annually, but calculated daily by linear interpolation

180 Spectral Solar Irradiance forcing, includes extrapolated 11-years solar cycle based on sun-spot observations-sets of far infrared, near infrared and visible radiation (Krivova et al., 2011), read annually, but calculated daily by linear interpolation

A preliminary version of the LUH2 dataset (Hurtt et al., 2020): This forcing begins 1100 BP, but to slowly build up the land-use from zero to the first land-use state in the LUH2 dataset, a transition period of 1000 years is implemented starting at 2100 BP. Land-use is read annually, but calculated daily by linear interpolation. Land use is prescribed in the form of
185 transition maps that define the fraction of area that is converted from natural vegetation to crops and pasture or vice versa. Pasture is first distributed in the area covered by grass before it replaces forested area (Reick et al., 2021). After this rule has been applied, the remaining anthropogenic land-cover change is equally distributed to all PFTs, relative to their individual cover fractions, so that they experience the same gain or loss of cover fraction.

A detailed description of the transient simulation and the forcing mechanisms can be found in Bader et al., 2020, Brovkin et
190 al., 2019 and Dallmeyer et al., 2020. An evaluation of the simulated pre-industrial climate is provided in Appendix A.

2.2 The dynamic vegetation model LPJ-GUESS simulated plant cover

2.2.1 Model description

LPJ-GUESS (Lund-Potsdam-Jena General Ecosystem Simulator) is an individual-based dynamic ecosystem model optimised for global to regional studies (Smith et al., 2001 and 2014; Sitch et al., 2003; Smith et al., 2001). The model
195 employs a representation of vegetation dynamics (successional processes: establishment, growth, mortality) of a forest "gap" model allowing explicit representation of competition for resources (light, water, nutrients etc.). Growth of individuals or cohorts is simulated in several a number of replicate patches of 0.1 ha representing the grid cell. The climatic conditions and soil type are assumed to be identical between the patches. The probability of stochastic patch-destroying disturbance (fire, wind etc.) occurrence is controlled by pre-set generic disturbance intervals. When disturbances occur, the vegetation cover of
200 the one patch is destroyed. The proportion of vegetation affected is dependent on the total number of prescribed patches. The

vegetation is simulated as ~~Plant Functional Types (~~PFTs~~)~~ discriminated in terms of bioclimatic limits and physiological characteristics. The standard global PFT set comprises 10 woody PFTs representing major higher plant types of boreal, temperate, and tropical biomes and two grass PFTs distinguished by C3 and C4 photosynthetic pathways. Land use is implemented using external inputs determining the proportional distribution of up to seven land-cover types (natural vegetation, urban areas, cropland, managed forest, pastureland, peatland and barren land) and associated specialised PFT set to simulate the land-cover dynamics and biogeochemical fluxes (Lindeskog et al., 2013).

The model has been applied and benchmarked in a number of studies, both for global and European conditions (Smith et al., 2008; Hickler et al., 2012) and agricultural landscapes (Müller et al., 2021; Lindeskog et al., 2013). It is among the best available C cycle models (Piao et al., 2013) and can account for C-N interactions (Smith et al., 2014). Model performance in terms of reproducing vegetation and hydrological and biogeochemical cycles for past, present and future applications has been tested in numerous studies (e.g. Garreta et al., 2010; Miller et al., 2008; Olofsson and Hickler, 2008).

2.2.2 High-resolution time-slice simulations with LPJ-GUESS

LPJ-GUESS was forced with ca. 120 year monthly resolved climate (total cloud cover, precipitation and 2m air temperature) from the transient MPI-ESM1.2 simulation. Model runs were performed for six time slices of which four are distributed at nearly equal time intervals between 8 ka and 2 ka (representing mid and late Holocene) and two fall in the period with prescribed land use in the MPI-ESM1.2 simulation. The periods of the input climate data have been selected based on the two criteria of being close to the period of interest (e.g. 8 ka) and showing a relatively stable climate, characterised by little effects of the prescribed volcanic activity in the MPI-ESM1.2 simulation on the regional climate. The age intervals and acronyms of the six time slices are provided in Table 1.

To minimise the effect of systematic biases in the simulated climate on LPJ-GUESS~~model~~-simulated vegetation, an anomaly approach was used (~~cf.~~ Wohlfahrt et al., 2008). First, the anomaly between each month of each year of a simulated period (e.g. 8 ka) and the respective climatological monthly mean at the end of the simulation (i.e. 0~~1~~99.2 kaBP – 0.1 ka~~00 BP~~) was calculated based on the original T63 gridded MPI-ESM1.2 output. These anomalies were then interpolated bilinearly to a regular grid with a spatial resolution of 0.5° x 0.5° and added to a reference climate dataset, i.e. the climatological monthly mean of the years 1901-1930 taken from the CRU TS 4.0 dataset (Harris et al., 2020; University oOf East Anglia Climatic Research Unit (CRU) et al., 2017). The anomaly approach has the advantage of preserving regional climatic gradients that are an imprint of e.g. a complex orography (Harrison et al., 1998) despite the relatively coarse spatial resolution used in the MPI-ESM1.2 simulation. ~~In order~~ Tto avoid negative values in precipitation and cloud cover (resulting from the anomaly~~is~~ calculation) in the LPJ-GUESS climate forcing data, all negative values were set to 0 for these variables.

230   The spin-up to reach vegetation and biogeochemical equilibrium was set to 300 years using the first 30 years of the detrended climate-data of the time window. Inputs of monthly climate variables (temperature, precipitation and cloud cover), soil texture data described in (Sitch et al., 2003) and land-use proportions derived from the JSBACH output (1_ka and PI time-slice), along with a set of PFT specific parameters determining the bioclimatic niche and physiological parameters (growth form, leaf phenology, photosynthetic pathway, life history etc.) werase used to set up the simulation. The number of

235   simulated patches was set to 25 and the disturbance interval to standard 100 years for all model runs except for the disturbance sensitivity tests. These tests were also run with 25 patches while changing the disturbance interval in each run (25, 75, and 200 years). Each time a disturbance occurred, one patch (i.e. ca. 4%) of the simulated vegetation was destroyed.

      Yearly outputs of the computed PFT-specific leaf-area indexes (LAI) per grid cell were recorded. The spatial resolution of the LPJ-GUESSuess runs was matching that of the climate inputs. Simulations ran uninterrupted for the whole time series.

240   The model-produced annual record of PFT-specific LAI was averaged over the last 30 years of the modelled time slice and converted to fractional plant cover (FPC) by applying a simplified version of Lambert-Beer law (Sitch et al., 2003; Monsi and Saeki, T., 1953; Monsi, 2004; Prentice et al., 1993):

$$FPC(PFT) = \left(1.0 - exp\left(-0.5 * LAI(PFT)\right)\right) (1)$$

      All woody FPC(PFT) fractions were summed to represent total tree cover (TC) of the grid cell, and the grass FPC(PFT) and

245   land- use-related FPC(PFT) (for 1_ka and PI only) were summed to represent the total open land cover (OC) fraction. To ensure comparability with pollen-based vegetation-cover estimates (i.e., assuming 100% vegetation cover), the model-based TC and OC were recalculated to sum up to 100% cover in each grid cell.

2.3 Quantitative pollen-based vegetation reconstructions using the REVEALS model

2.3.1 The REVEALS model

Pollen records consist of pollen counts from samples taken at many levels in a sediment or peat core. Pollen counts include counts for all identified pollen-morphological types (or taxa, named "pollen types" or "pollen taxa") corresponding to plant families, genera, groups of species or species. Pollen counts are then used to calculate pollen percentages for all pollen types and, if possible, pollen accumulation rates (PAR). The latter requires chronologies based on many 14C dates with high time resolution as well as sediment or peat sequences with relatively regular accumulation rates through time, i.e., no major events in sediment or peat deposition. These requirements imply that high-quality Holocene PAR records are few in comparison to pollen % records. Pollen productivity varies between plant taxa and pollen dispersal properties vary between pollen types (depending on their size and shape). These issues imply that pollen percentages (and PARs) from fossil pollen assemblages can only provide qualitative or semi-quantitative information on past vegetation changes, i.e., sporadic, or regular presence of plant taxa, presence in more or less large quantities, increases and decreases of plant taxa. The REVEALS model (Regional Estimates of Vegetation Abundance for Large Sites; Sugita, 2007) was developed to estimate regional plant abundance using pollen % records from large lakes and corrects for the biases due to the inter-taxonomic differences in pollen productivity, dispersal, and deposition. However, REVEALS can also be applied with pollen records from multiple small sites, although it generally results in larger standard errors (SEs) on the estimates of plant cover, as was demonstrated with model simulations (Sugita, 2007) and empirical data (Trondman et al., 2016). REVEALS has explicit assumptions (Sugita, 2007) of which the most critical ones in the context of this study are described in the Discussion section. REVEALS requires several parameters of which relative pollen productivity (RPP) and fall speed of pollen (FSP) are the most crucial. The model was first validated in southern Sweden (Hellman et al., 2008a) and later in several regions of Europe (e.g. Soepboer et al., 2010), in northern America (Sugita et al., 2010), and in China (Wan et al., 2022). The spatial scale of a REVEALS reconstruction was estimated to ≥ 100 km x 100 km for modern vegetation in southern Sweden (Hellman et al., 2008b) and assumed to be in the same order of magnitude for European vegetation in general and through most of the Holocene. REVEALS is therefore well suited to produce pollen-based gridded reconstructions of plant cover at a spatial scale appropriate for comparison with DGVM simulations.

2.3.2 Pollen-based REVEALS plant-cover reconstructions of Holocene plant cover in Europe

2.3.1 The REVEALS model

The REVEALS model was developed to estimate regional plant abundance using pollen count records from large lakes (> 50 ha) and corrects for the biases due to the inter-taxonomic differences in pollen productivity, dispersal, and deposition (Sugita, 2007). REVEALS can also be applied with pollen records from multiple small sites (lakes and bogs < 50 ha), although it generally results in larger standard errors (SEs) on the estimates of plant cover, as was demonstrated with model simulations (Sugita, 2007) and empirical data (Trondman et al., 2016). REVEALS has explicit assumptions (listed in Sugita, 2007), e.g. there is no vegetation growing on the basin (i.e. REVEALS is not appropriate for pollen records from large bogs, e.g. Li et al. 2020), wind comes from all directions and wind speed is constant through time and space (i.e. 3m/s for Europe), pollen productivity of plant taxa and fall speed of pollen are taxon-specific constants through time, and plants are distributed on a flat topography (i.e. topography is not taken into account in the model, therefore REVEALS is not suited for mountain regions unless the distribution of pollen records is adequate, Marquer et al., 2020). Validation of the model and evaluation of

the effect of assumption violation on reconstruction of plant cover in Europe can be found in e.g. Hellman et al. (2008a), Trondman et al. (2015, 2016), and Marquer et al. (2020).

The spatial scale of a REVEALS plant-cover reconstruction was estimated to ≥ 100 km x 100 km for modern vegetation in southern Sweden (Hellman et al., 2008b) and is therefore well suited to produce gridded pollen-based reconstructions of plant cover at a spatial scale appropriate for comparison with e.g. DGVM simulations. The existing gridded plant-cover reconstructions for Europe (see references in the Introduction) assume that the spatial scale of REVEALS-reconstructed plant cover in Europe and over the Holocene is in the same order of magnitude as that estimated for southern Sweden.

~~There are two~~ 2.3.2 Dataset of Holocene pollen-based REVEALS plant cover in Europe used in this study

In this study, we used the REVEALS plant-cover dataset from the reconstructions in Marquer et al. (2017) and archived in PANGAEA data archive (Marquer et al., 2019) because it was the only dataset available for the entire Holocene at the time of our analysis. The reconstructions cover large areas of northern and central Europe, i.e. Ireland, the British Isles, and several regions on a latitudinal transect from the Alps in the south to northernmost Norway in the north (Fig. 2). The grid system, pollen-data handling, and REVEALS application (parameter setting, etc.) used for these reconstructions follow the protocol described in Trondman et al. (2015) and used globally by the PAGES LandCover6k working group (e.g. Gaillard et al., 2015; Dawson et al., 2018). ~~They were both produced at a 1º spatial scale for studies on land use as a climate forcing using climate models and DGVMs (Strandberg et al., 2014 and 2022). Trondman et al. (2015) performed reconstructions for five key time windows of the Holocene, while Githumbi et al. (2022a and 2022b) produced reconstructions for 25 consecutive time windows between 11.7 ka and the present. Moreover, Holocene REVEALS reconstructions for 19 of the best pollen records from large lakes in Europe (Marquer et al., 2014) and for 36 1ºx 1º grid cells including these 19 pollen records and all other available pollen records around them (Marquer et al., 2017) were performed to study questions related to Holocene vegetation dynamics and plant diversity along N-S and W-E transects through Europe. All gridded REVEALS reconstructions for Europe follow the same protocol as described in Trondman et al. (2015) and used globally by the PAGES LandCover6k working group (e.g. Li et al., 2023; Dawson et al., 2018). For the present study, we chose the REVEALS reconstructions from Marquer et al., (2017) as they were the only ones available for the entire Holocene at the time of our analysis and represented the best ones in terms of quality of the pollen records used. The major difference between Marquer et al. (2017) reconstructions and the more recent ones by Githumbi et al. (2022a and 2022b) is the dataset of RPPs and FSPs used (see Discussion section).~~

~~The REVEALS dataset of Marquer et al. (2017) covers large parts of northern and central Europe, i.e., Ireland, the British Isles, and several regions on a latitudinal transect from the Alps in the south to northernmost Norway in the north (Fig. 2).~~ The dataset contains REVEALS estimates for 25 plant taxa and 25 consecutive time windows, 22 between 11.7 ka and 0.7

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ka with a 500 years-resolution, and three with a shorter time resolution, 0-0.1 ka, 0.1-0.35 ka, and 0.35-0.7 ka. They are based on 28 grid cells within a total of 36 1°x 1° grid cells which include large lakes (≥ 50 ha) and all other available pollen records around them (from small lakes and bogs (< 50 ha)). The 8 remaining grid cells correspond to a mix of small sites. The number of pollen records per grid cell varies between one and 32. Only seven grid cells include a single site, in each case a large lake (i.e. reliable reconstructions). Five grid cells with less reliable reconstructions (< 3 small sites, only small bogs or one large bog with small sites) are indicated in Fig. 2. The reconstructions for these grid-cells can be biased towards local plant cover. For criteria of reconstruction reliability, see Trondman et al. (2015, 2016). The REVEALS estimates are based on 151 pollen records and available for 25 consecutive time windows over the last 11.7 ka as follows: 0-0.1 ka, 0.1-0.35 ka, 0.35-0.7 ka, and 500-year time-windows between 0.7 and 11.7 ka.

The pollen records were selected from the European Pollen Database (Giesecke et al., 2014), the Alpine Palynological Database (University of Bern, Switzerland), or provided by individual authors. The pollen records are from large lakes (≥ 50 ha) and small sites (lakes and bogs < 50 ha). The grid system, pollen-data handling, and REVEALS application (parameter setting etc.) follow the protocol described in Trondman et al. (2015), and the RPP and FSP dataset of Mazier et al. (2012) was used.

The larger the number of pollen records (sites), the better the REVEALS reconstruction (Sugita, 2007; Trondman et al., 2016). However, given that the REVEALS model was developed for pollen records from large lakes, a single pollen record from a large lake provides a reliable reconstruction of regional plant cover (Hellman et al., 2008; Trondman et al., 2016). The number of pollen records per grid cell varies between one and 32 in the dataset of Marquer et al. (2017). Only seven of the 36 grid cells include a single site, and in each case it is a large lake, therefore the REVEALS estimates are reliable. The REVEALS estimates for grid cells including only two small sites or one large bog (with few additional small sites) are considered as less reliable given that large bogs violate one of the assumptions of the REVEALS model and estimates based on a few small sites will be biased towards local plant cover (see e.g. Li et al., 2020; Githumbi et al., 2022a for details). Such grid cells are few in Marquer et al. (2017) and indicated in Fig. 2. They are: one grid cell with one large bog and three small sites and one grid cell with two small lakes (both in southern Finland), and one grid cell with one small lake and one small bog (Scandinavian mountains, Sweden-Norway boundary). Moreover, there is one grid cell in Great Britain with four pollen records from small bogs only, which may bias the REVEALS reconstruction towards local plant cover on the bogs.

For each of the 36 grid cells the REVEALS model has been is run for each pollen record individually and the mean REVEALS estimates of plant cover (and their SEs) for the grid cell are thenhave been calculated for the 25 plant taxa (Tab. 2). The total cover of plant taxa within a grid cell is 100%. This is due to the fact, that estimatingREVEALS cannot estimate the cover of bare ground from pollen is a challenge. So far, only one attempt of estimating bare ground from pollen in northern China using the modern analogue approach has been published (Sun et al., 2022).

Plant-functional types (PFTs) were defined following Wolf et al. (2008). However, modifications had to be made as pollen-based plant cover provides the total cover of each plant taxon irrespective of whether it belongs to one or several PFTs. Therefore, each plant taxon can be included in only one PFT (Table 2). The method used to calculate mean SEs for grid cells

350 and the PFTs Ses and the delta method (Stuart and Ord, 1994), is described in Li et al. (2020).

2.4 Methods used to compare DGVM-simulated and REVEALS-estimated tree cover Comparison between vegetation model-simulations and pollen-based REVEALS reconstructions

The differences between REVEALS and JSBACH estimates, REVEALS and LPJ-GUESS estimates and JSBACH and LPJ-GUESS estimates have been assessed for tree cover, deciduous tree cover and conifer tree cover in each grid cell. We

355 calculated the absolute value of the differences between two estimates for each grid cell, and we defined a scale of agreement based on this absolute value and the data distribution over the entire study region (i.e. absolute values for all grid cells) for each time window. The first quartile and the median have been calculated. A good agreement corresponds to an absolute value of the differences between two estimates lower than the first quartile. An agreement corresponds to an absolute value of the differences between two estimates situated between the first quartile and the median. A disagreement

360 corresponds to an absolute value of the differences higher than the median. TheAll results have been plotted using ArcGIS 10.6 to observe the spatial distribution of the differences between the different past vegetation reconstructions.

The squared chord distance (Prentice, 1980) was calculated for each time window to evaluate the spatial dissimilarities over time between REVEALS, JSBACH and LPJ-GUESS in terms of total, deciduous, and evergreen tree cover.To evaluate the overall spatial dissimilarities between REVEALS and JSBACH and LPJ-GUESS regarding the total tree cover, deciduous

365 tree cover and evergreen tree cover, the squared chord distance (Prentice, 1980) is calculated for each time window. The squared chord distance is commonly used to calculate the dissimilarities between two sets of data that represent assemblages, i.e. plant composition. In this study we apply it to the ensemble of grid cells across our study region, i.e. we are studying how dissimilar the spatial grid compositions between REVEALS and the DGVMsynamic Global Vegetation Models are for each time window.

370 **3 Results**

We compare the vegetation change simulated by the two models JSBACH (coupled in MPI-ESM1.2) and LPJ-GUESS (forced with MPI-ESM1.2 climate) with pollen-based REVEALS reconstructions. Please note, that land-cover changes (decreases or increases) are expressed in absolute fractions of the grid cells, e.g. an increase in cover by 20% at x ka from a cover of 50% of the grid cell at y ka implies that the cover at x ka is 70% of the grid cell.

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3.1 European tree cover change since 8 ka

The two models simulate a similar European potential natural vegetation history but with very different total tree-cover fractions over time (Fig. 3). LPJ-GUESS shows a dense tree coverage between 80-100% in large parts of Europe during mid- and Late Holocene. More open landscapes are simulated for the Mediterranean area (particularly southwestern Spain and southern Italy) and the mountainous regions of Scandinavia and the Alps. Vegetation is nearly constant until the prescribed land-use forcing sets in, substantially reducing the tree cover in western, central and eastern Europe at the 1 ka and PI time-slices.

JSBACH generally simulates a similar spatial gradient as LPJ-GUESS, with fewer trees in the Mediterranean region and the Scandinavian mountains and the largest tree cover in central Europe. However, tree coverage is much lower in most parts of Europe during mid- and Late Holocene than simulated by LPJ-GUESS, mostly reaching tree cover fractions of 40-70%. Particularly obvious are the very low tree cover fractions in the coastal areas of the Atlantic Ocean and Mediterranean Sea as well as in southern Europe, leading to remarkable differences between JSBACH and LPJ-GUESS during mid- to Late Holocene. In contrast, JSBACH simulates a slightly higher tree cover fraction than LPJ-GUESS in parts of northern Scandinavia and the Alpine Region.

As in LPJ-GUESS, JSBACH indicates relatively constant tree cover distributions until the prescribed land -use is applied. Due to the higher potential natural tree cover fractions in western, central and eastern Europe, the land use has a stronger effect in LPJ-GUESS than in JSBACH, reducing the differences between the models with respect to the total tree cover fraction for the time-slices influenced by land use (1 ka and PI).

The REVEALS-based estimates of European tree coverage for 8 ka (Fig. 3) indicate dense forests in most of the represented regions with highest tree - cover fractions along the Baltic Sea (Central and Boreal Europe) and in the Alpine region and a more open landscape on the British Isles. The reconstructed tree cover in Northern Scandinavia is also quite high and therefore strongly deviates from the model results.

Fig. 4 displays the mean trend in the different subregions simulated by the models and REVEALS, based on the different time-slice experiments. Since the models do not reveal much variability in the tree cover, the few simulated data points were linearly interpolated. The presented time series therefore only shows the long-term trend but cannot reflect the centennial variability.

For the Alpine region, the dynamic vegetation models and the REVEALS-based reconstructions compare well with total tree-cover estimates around 70-85% in the mean. REVEALS indicates a tree cover maximum at 6 ka and a decreasing trend already starting at 4 ka from about 86% at 4 ka to approximately 56% at PI. In contrast, the models simulate relatively constant tree coverage for the period 8 ka to 2 ka followed by a sharp drop in the tree cover fraction after 2 ka as a response

405 to the prescribed land use. Tree cover is considerably reduced in all datasets, and nearly halved between 2_ka and PI (from about 75% to 35%) in the models. The several millennia long mismatch in timing and intensity between the models and the reconstructions indicate not only a later onset of but also a too extensive deforestation in the models compared to REVEALS estimates.

For Central Europe, LPJ-GUESS simulates constant high tree coverage of more than 90% between 8_ka and 2_ka, and a
410 sharp reduction (to 27% at PI) thereafter. JSBACH reveals a similar, albeit weaker, vegetation dynamic with mid-Holocene tree cover fractions of approx. 50% and a substantially decreased tree cover at PI (~22%). Also, for this region, the REVEALS-based reconstructions show a different temporal trend. The decrease in tree cover begins as early as 6_ka and accelerates towards PI. At 8_ka, tree cover is estimated at 85%, agreeing well with the LPJ-GUESS estimates. Tree cover is nearly halved to 41% at PI according to REVEALS. The reduction in tree coverage after 2_ka parallels the trend in the
415 JSBACH simulation, but on higher tree cover fractions.

LPJ-GUESS and REVEALS indicate similarly high (~85-90%) mid-Holocene tree-cover fractions in Boreal Europe, whereas JSBACH simulates much less trees (~60%). Both, REVEALS and LPJ-GUESS, show relatively constant tree cover between 8_ka and 4_ka. In REVEALS, the tree cover fraction decreases slightly from 4_ka but is substantially reduced (by 36%) only in the last ~2000 years. In LPJ-GUESS, tree coverage declines by only 15% between 2_ka and PI. It is difficult to
420 figure out the effect of land use in this model-setup, but the differences in the magnitude of the tree-cover decrease could indicate an underestimation of the prescribed land use intensity in this region in the models. In JSBACH, land use has hardly any effect at all in the regional mean.

For Northern Scandinavia, the models and the REVEALS reconstructions show a similar trend of steadily decreasing tree cover during the Holocene but differ significantly in absolute tree cover. REVEALS estimates a tree coverage of approx.
425 80% at 8_ka and declining to 60% at PI. LPJ-GUESS simulates a mean tree cover of 58% at 8_ka and 26% at PI. In JSBACH, the simulated tree cover fraction is reduced from 40% at 8_ka to 21% at PI. The relative decrease in tree coverage (tree cover is halved in both models) is thus much stronger in the models than estimated by the REVEALS-based reconstructions.

The models and reconstructions show the largest deviations on the British Isles.-_While JSBACH indicates very low tree cover fractions of approx. 20% at 8_ka and a slight decrease in tree cover after 2_ka to 13% at PI, LPJ-GUESS simulates high
430 tree-cover fractions during the mid-Holocene (~92%) and a sharp drop to 58% between 2_ka and 1_ka. REVEALS estimates a tree cover of 60% at 8_ka and a constant decrease between 6_ka and 2_ka to 27%, followed by a stronger drop in tree cover towards 1_ka (to 10%) and a slight recovery towards PI (to 14%).

In summary, the inter-model spread is largest in the British Isles and Central Europe as well as in southern Europe (no REVEALS region) with much lower tree cover fractions in JSBACH than simulated by LPJ-GUESS. Compared to the

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REVEALS estimates, JSBACH underestimates the tree cover in practically all regions and during most of the time slices except for the Alpine region, whereas the LPJ-GUESS estimated tree cover is well comparable to the REVEALS estimates in Boreal and Central Europe. However, LPJ-GUESS significantly underestimates the tree cover in Northern Scandinavia and overestimates it in the British Isles. The models suggest rather constant European tree coverage between 8 ka and 2 ka, whereas the REVEALS-based reconstructions show a stronger dynamic with the tendency of a mid-Holocene (6 ka) maximum tree cover followed by a steady decline. The prescribed land use has a larger effect in LPJ-GUESS than in JSBACH, indicating differences in its implementation and a higher tree cover level at the onset of land use in LPJ-GUESS. Compared to the REVEALS reconstructions, the prescribed land use in the models appears to be too large in parts of Central and western Europe (particularly in the Alpine region) and too small in Boreal Europe.

3.2 REVEALS *versus* vegetation models

3.2.1 Temporal distribution of the differences

The calculated chord distances between the datasets indicate that the simulated tree-cover pattern in LPJ-GUESS is overall in better agreement with REVEALS during the mid-Holocene than the one inferred by JSBACH (Fig. 5). Particularly for the 8 ka time slice, the total tree cover simulated by LPJ-GUESS compares very well with the reconstructions, yielding a chord distance below one1. The chord distance for the evergreen and deciduous tree cover is substantially higher, indicating that the ratio of deciduous to evergreen forest is not quite as well represented by LPJ-GUESS.

The results for JSBACH suffer from the severe underestimation of deciduous trees compared to REVEALS. This is reflected in the very high chord distance at 8 ka. However, the spatial distribution of evergreen trees is better represented by JSBACH at 8 ka. The agreement of the two different models to REVEALS converges towards 1 ka and PI. The representation of the deciduous trees gets continuously better in JSBACH, while it gets worse in LPJ from 6 ka to 2 ka. This affects the model-data agreement with respect to the total tree cover. The mismatch in tree cover between LPJ-GUESS and REVEALS increases towards 1 ka. For JSBACH, the chord distance to the REVEALS estimated tree cover distribution is relatively constant until 4 ka and getting better afterwards.

Since the mismatch to the reconstructions in the distribution of evergreen trees is relatively constant in both models, the improvement of the agreement in total tree cover between the DGVMs and REVEALS is clearly driven by the increasingly better representation of the deciduous trees in the DGVMs with time. Even if generally, the cover fractions of the evergreen trees in both models agree better with the REVEALS estimates than the deciduous tree cover fractions. Since 4 ka, the total tree cover distribution simulated by JSBACH agrees better with REVEALS than the LPJ-GUESS simulated distribution, but the misrepresentation of the ratio of evergreen to deciduous trees remains.

Overall LPJ-GUESS shows the best agreement with REVEALS for the 6_ka time-slice, followed by the PI time-slice.
JSBACH also indicates an improvement of the agreement with respect to the cover fraction of evergreen trees for 6_ka.
However, it agrees best with REVEALS for PI conditions.

3.2.2 Spatial distribution of the differences

We calculate a three-scale agreement index to quantify and evaluate the spatial difference between the models and the
REVEALS-based reconstructions in each grid-_cell (cf. Methods). The pattern indicates model-specific regions with
systematic agreement or disagreement (Fig. 6). For instance, JSBACH fails to reproduce the REVEALS mid-Holocene tree
cover fraction in large parts of central Europe, the Baltic States and southern Sweden, while LPJ-GUESSuess shows
reasonably good overall agreement in these regions. Both models produce tree cover estimates not comparable with
REVEALS reconstructions for the British Isles. For JSBACH this mismatch gets slightly better towards PI, mainly due to an
improved representation in some grid-_cells in northern Germany, Scotland and Ireland. In most grid cells on the British Isles
and in western Germany, the fractional coverage of evergreen trees simulated by both models has better agreement with the
REVEALS estimates than the deciduous tree cover fraction, underlining the previous finding of the deciduous trees as a
driver of the model-data mismatch. However, in eastern Europe (Poland, Baltic States) the deciduous tree coverage
simulated by both models is more in line with the REVEALS estimates than the evergreen tree coverage.

The tree cover fraction in the Alpine region simulated by JSBACH corresponds better to the REVEALS estimates than in
other regions until the land use in the model sets in. All grid-cells in these regions receive agreement or even good agreement
with respect to the JSBACH-derived total tree cover. In contrast, LPJ-GUESS indicates only some agreement in the central
Alpine region. The fraction of deciduous trees is mostly in agreement for both models during the mid-Holocene, but the
agreement decreases towards the late-Holocene. The simulated fractions of evergreen trees disagree with the REVEALS-
based estimates at all time-slices.

In most grid cells of Northern Scandinavia, the deciduous and total tree cover do not match the reconstructions during all
time-slices. The evergreen trees, however, show agreement in many grid cells, which becomes even better towards 1_ka.

The total tree coverage in REVEALS and JSBACH compares well in the domain southern/central Norway to central Sweden
and southern Finland, reaching values of agreement to good agreement in most grid cells during all periods. LPJ-GUESS
does not agree as well as JSBACH with REVEALS, particularly in southern Norway. Interestingly, in these regions
JSBACH is not able to capture the REVEALS-estimated fractions of evergreen and deciduous trees, indicating a
misrepresentation of the ratio of the different tree types. In contrast, LPJ-GUESS captures the REVEALS deciduous tree-
cover fraction for the Late Holocene.

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Based on the results presented in Fig. 5 and Fig-. 6, it is not possible to determine which model is overall more consistent with REVEALS. The agreement strongly depends on the region. It should be noted, however, that in a statistical sense the thresholds for the three scale-agreement indices are equal for both models, but the absolute values of the thresholds may differ, depending on the general spatial variability in the models that is higher in JSBACH than LPJ-GUESS. This statistical definition of the threshold allows JSBACH to be still rated in the category "agreement" with larger absolute differences to REVEALS.

4 Discussion

Comparison of the tree- cover fractionabsolute values s and their spatial and temporal distributions as simulated by the DGVMs Dynamic Global Vegetation Models and estimated from pollen data by the REVEALS model is challenging. When compared with REVEALS, JSBACH rather underestimates European tree cover during mid- and Late Holocene, while LPJ-GUESS mainly overestimates the tree coverage, at least in those time-slices without human impact prescribed to the models (8 ka to 2 ka). In addition, the model agreement withto the REVEALS results shows a spatially varying pattern that is different for each model. Whereas JSBACH reveals the strongest mismatch to the reconstructions in Central Europe during all periods, LPJ-GUESS shows relatively good agreement in this region. Both models fail to reproduce the tree cover history on the British Isles and the in Northern Scandinavia. For the Alpine region, the simulated JSBACH- simulated tree- cover fractions correspond well to the REVEALS estimates, while the LPJ-GUESS values more often disagree.

In most regions, the models are not able to simulate the correct ratio of deciduous to evergreen trees, but the evergreen tree cover distribution is generally more in line with the REVEALS data than the deciduous tree coverage. While LPJ-GUESS shows a relatively good overall agreement with REVEALS at 8 ka and particularly at 6 ka, the mismatch increases from 4 ka to 1 ka. In contrast, the distributions simulated by JSBACH continuously improve (in the mean) towards PI, indicating a convergence of the model results through time.

We assume the following possible reasons behind the model-data differences and discuss them thoroughly in the next sections (here, model refers to the two DGVMs and data to the pollen-based REVEALS plant cover):

a) climate and spatial resolution biases in the models

b) oversimplified vegetation dynamics in the models

c) modern parametrizations and tuning of the models to modern conditions

d) differences between the pollen-based REVEALS estimate of deforestation due to land use and the prescribed land use as well as differences in the land-use implementations in the models

e) shortcomings of the data, i.e. caveats of the REVEALS model itself and the pollen-based reconstructions of plant cover used in this study.

4.1 Effect of biases in the simulated climate and climate input fields on the tree cover distribution

Based on a previous study (Dallmeyer et al., 2021) analysing a slightly different Holocene MPI-ESM1.2 simulation, we assume summer temperature to be the main climatic driver of the simulated vegetation in the European regions considered here. To infer possible biases in the simulated climate trend, we compare the LPJ-GUESS temperature forcing data with chironomid-based reconstructions of the temperature of the warmest month (Twarm) (Fig. 7), extracted from the synthesis of Kaufman et al. (2020) and few other sources (references are given in Table B1 in Appendix B). To infer possible biases in the simulated climate trend, the anomalies of the MPI-ESM1.2-simulated temperature of the warmest month (Twarm) to pre-industrial climate are added to the CRU TS4.0 reference state and then compared to chironomid-based Twarm reconstructions (Fig. 7), extracted from the synthesis of Kaufman et al. (2020) and few other sources (references are given in Table B1 in the Appendix B). Thus, we compare basically the climate that is prescribed to LPJ-GUESS with proxy-based reconstructions. We assume that summer temperature is the main climatic driver of the vegetation in the regions considered here, as previous analysis based on a slightly different Holocene MPI-ESM1.2 simulation has revealed for most parts of the regions considered in this study (Dallmeyer et al., 2021). It should be stressed, however, that temperature reconstructions based on chironomids may be subject to various caveats, whose detailed description is beyond the scope of this paper. Other environmental changes in e.g., nutrient, anoxia, and salinity, can indeed also lead to changes in chironomid assemblages, which may affect the chironomid-inferred temperatures (e.g. Velle et al., 2010). Nevertheless, the distribution of chironomid assemblages generally strongly correlates with the warm-season air and lake temperature in the temperate and subarctic regions, although the causal relationships are still not fully understood (e.g. Eggermont and Heiri, 2012).

The model-based mean temperature dynamics are well in line with the reconstructions. Simulated Twarm levels in the Alpine region are different for all sites and cover a wider range than the reconstructions indicate. While Twarm decreases relatively uniformly at all sites after 8 ka in the model, two of the four reconstructions show a climatic optimum between 7 ka and 4 ka. This is consistent with the maximum tree-cover fraction in the REVEALS data at the 6 ka and 4 ka time slices in this region. The good agreement of the simulated and reconstructed mean temperature of the warmest month reflects the similarly high tree cover fractions estimated by REVEALS and simulated by both models.

For Central Europe, only one chironomid-based reconstruction exists in Kaufman et al. (2020), i.e. for Lake Zabieniec (Płóciennik et al., 2011). For this record, new reconstructions have been published recently (Luoto et al., 2019; Kotrys et al., 2020). With a mean Twarm of around 16 °C, these reconstructions indicate a substantially, roughly 5°C, cooler summer climate than prescribed to LPJ-GUESS at 8 ka. Furthermore, the reconstructions reveal stable or slightly increasing Twarm

for the mid-Holocene in contrast to the decreasing trend in the model. For the Late Holocene, the reconstructions of the Zabieniec record diverge strongly, revealing contradictory changes in Twarm~~.~~, but ~~Also, for the Late Holocene, all reconstructions from Zabieniec~~ indicate all a much colder climate in this region than revealed by the simulated climate that has been used to force LPJ-GUESS. However, a chironomid-based reconstruction from northern Poland (Lake Spore, Pleskot et al., 2022) is well in line with the model. Both show declining Twarm from about 19°C to 17.5°C during the Late Holocene and only reveal substantial differences for the last 500 years ~~BP~~. The difference in Twarm level at Lake Zabieniec may act as an indicator for a slightly too warm model derived climate in the southern part of Central Europe. This warm bias may also extend to the southeastern part of the simulation region, displayed exemplarily for the 4 ka time slice (Fig. 7). This warm bias may contribute to the slightly overestimated tree cover fraction in LPJ-GUESS compared to the REVEALS estimates in Central Europe.

In Boreal Europe, model-based Twarm is higher by approximately 2-3°C during the entire period compared to the chironomid-based reconstructions. While the model indicates an almost linear decrease in Twarm since 8 ka, some reconstructions reveal a warming towards 6 ka and decreasing temperatures afterwards. This dynamic is also visible in the mean over the region and fits nicely to the REVEALS-based reconstruction of tree cover change. Regardless of these discrepancies, the modelled temperature range is still located well within the tolerance limits of boreal tree taxa. Since the tree cover fractions are similar in LPJ-GUESS and REVEALS, this bias in temperature does not affect the total tree cover distribution.

Simulated temperatures of the warmest month are on a similar level as the reconstructions in Northern Scandinavia during mid-Holocene, not explaining the 20% lower tree cover fraction in LPJ-GUESS compared to the REVEALS estimates. The reconstructions reveal only a slight decrease in Twarm since 8 ka, in contrast to the stronger decrease in the model. These differences in temperature trend are in line with the slight deviations in the rate of tree cover decline between LPJ-GUESS and REVEALS in Northern Scandinavia. However, both modelled and reconstructed Twarm averages fluctuate close to the 10 °C, with the modelled temperature falling below it during the Late Holocene. The 10°C limit is - according to the known Köppen's Rule (Köppen~~, W.,~~ 1936) - accepted as delimiter of boreal forest distribution, and Twarm falling under this limit could be a possible cause for low model-based tree cover estimates.

Tree-cover fractions simulated by LPJ-GUESS show the highest discrepancy with the REVEALS estimates on the British Isles, overestimating tree coverage by more than 30%. While the mean trend in Twarm during the period 6 ka to 3 ka is relatively similar (slightly decreasing) between the prescribed climate and the chironomid-based reconstructions, LPJ-GUESS shows a constant, large tree cover until land use sets in and REVEALS indicates a strong decrease in tree cover

starting at 6_ka. These deviations in total tree-cover changes through time cannot be explained by the simulated summer temperature.

Comparing the chironomid-based reconstructions with the climate simulated by MPI-ESM1.2 directly is not very meaningful, since many of the chironomid sites are located in mountainous regions. Due to the coarse spatial resolution of the model, one can expect the climate to be rather too warm and to show much stronger differences in annual extremes than in the mean over the seasons. The evaluation of the pre-industrial climate (Appendix A and Fig. A1 in the Appendix A) reveals only minor differences in the seasonal temperature mean between the CRU TS4.0 data and the simulated MPI-ESM1.2 climate for PI.

The total tree coverage simulated by JSBACH agrees well with the LPJ-GUESS results in the Alpine region and Northern Scandinavia with Twarm limited tree growth. JSBACH strongly underestimates tree coverage in Central Europe and for the British Isles.– The former region does not experience any substantial differences in seasonal temperature, and the wetter climate in MPI-ESM1.2 probably does not induce changes in total tree coverage as the land cover of the region is not moisture-limited. The latter region is affected by strong deviations in the precipitation pattern with a much drier and rather warmer climate at most grid-cells for which REVEALS reconstructions are available. Since the strong differences in tree coverage compared to LPJ-GUESS cover the entire British Isles, and thus also the regions with overestimated precipitation, these climate biases cannot be responsible for the strong deviations in vegetation composition between the models.

In Boreal Europe, winter and summer climate in MPI-ESM1.2 is slightly cooler than observed. Since the climate is nevertheless well within the climatic tolerance range for extratropical tree PFTs in JSBACH, this does not limit the establishment of trees in JSBACH. However, while JSBACH has only one deciduous and one evergreen PFT for temperate and boreal conditions, the PFT list of LPJ-GUESS includes several PFTs developed specifically considering boreal conditions (Table 2). This may contribute to the substantially lower tree coverage in JSBACH compared to LPJ-GUESS.

We conclude that climate (temperature) biases are not the main driver of differences in the total tree-cover fraction between the REVEALS estimates and the vegetation models and between LPJ-GUESS and JSBACH in most of the regions. The simulated climate trends are mostly in line with the chironomid-based reconstructions and in case of differences, these cannot cause the discrepancies in the tree-cover trend. Therefore, it is highly likely that land use is the reason for the much earlier tree-cover decline in the reconstructions than in the models and that land use is the main driver of the long-term Holocene tree-cover change in Europe.

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4.2 Effect of oversimplified vegetation and soil dynamics in the models

Even though we have adjusted the PFT distributions calculated with the different methods to be more compatible with each other, ~~there are~~ various technical reasons ~~that~~ can lead to differences between the ~~Dynamic Global Vegetation Models~~ DGVMs and between these and REVEALS. Many processes can only be represented in a simplified form or have not been implemented in vegetation models yet. For instance, vegetation is only aggregated in a few PFTs, which neither match in number or definition to each other nor to REVEALS (Table 2) and not all PFTs considered in REVEALS have a counterpart in the models. Whereas REVEALS reconstructions reflect actual land cover including understorey vegetation, wetlands and other specialised communities, the models estimate terrestrial high ground vegetation only. In contrast, the models calculate a fraction of uncovered soils that cannot be determined in pollen-based reconstructions and is therefore not included in the REVEALS data. LPJ-GUESS considers early and late successional deciduous and evergreen tree types separately, whereas JSBACH distinguishes only one ~~single~~ deciduous and evergreen tree type, respectively. This differentiation allows for quick re-establishment of the tree cover after disturbance or climate induced mortality, making LPJ-GUESS simulated forest cover considerably more stable and less affected by random deforestation events.

Seed dispersal is not included in either of the models. Seeds are assumed to be available everywhere all of the time, whereas in the real world they need to be transported before a tree can grow in a new spot under favourable environmental and site conditions for tree growth. Moreover, depending on existing vegetation and tree species, the success of the establishment of a tree species might differ and thus their regional abundance. Although the dispersal- and migration-related delays in the establishment can mostly be expected in connection with the reforestation of Europe during the post-glacial and early Holocene period (Giesecke et al., 2017), they could be the reason for the differences in the mid-Holocene tree-cover changes between the models and REVEALS. The models show rather stable, in some regions slightly decreasing, mean tree cover with time, while REVEALS estimates tree-cover maxima between 6 ka and 4 ka for grid cells in Boreal and Central Europe and in the Alpine regions.

In addition, the models have simplified soil dynamics and do not consider changes in soil type or soil build-up. Permafrost soils, peat- or wetlands and blanket bogs are not represented in these simulations. Therefore, the models lack representation of important habitats such as mires and bog, whose (mainly treeless) vegetation increases the openness in reconstructed vegetation. Most of Europe had substantially more wetlands in the past than at present with its highly drained landscapes (Čížková et al., 2013). Furthermore, several European countries have ca. 15 (Sweden, Ireland and Scotland) - 30% (Finland) of territory occupied by blanket bogs and wetlands even today. Fyfe et al. (2013) highlight the considerable bias in landscape openness between the British Isles and continental Europe throughout the Holocene in the REVEALS-based study and suggest that this could at least partly be due to the considerably higher proportion of wetlands and uplands in the land cover of the British Isles compared to continental Europe. The formation of blanket bogs, typical for cool and hyper oceanic

climates, was accelerated by climate cooling starting ca. 6 ka (Gallego-Sala et al., 2016), which may, together with
640 anthropogenic deforestation, explain the small woodland cover. As a result of the disregard for of wetlands, models tend to
overestimate the tree-cover fraction in wetland-rich regions. This potential overestimation may explain the deviations
between REVEALS and LPJ-GUESS revealing much higher tree cover fractions in the British Isles and slightly higher tree
cover in Boreal and Central Europe than REVEALS.

4.3 Effect of modern parametrizations and of tuning of the model to modern conditions

645 Each vegetation model uses a different way of representing vegetation or incorporating processes such as PFT establishment,
plant competition, natural mortality, or reductions of plants by disturbances such as fire, wind throw and insects. However,
they share the implemented equations of these processes and the thresholds used (e.g. for the bioclimatic tolerance) are
validated and calibrated using modern observations. Parameters are tuned to meet modern vegetation distributions and do not
change in time. This means that all basic settings such as bioclimatic limits, allocation and mortality timescales or sensitivity
650 to disturbances are assumed to be constant over the simulation time. While it is true that species-specific bioclimatic limits
change slowly and most of the simulations do not reach the timescales necessary for considerable changes, the validity of the
assumption that present-day plant distribution is in equilibrium with climate is questionable. Modern species-distribution
patterns are heavily influenced by centuries of agriculture and forest management. The recent, abrupt global climate change
combined with other human impacts on plant species has enhanced the difference between potential and realised niche,
655 making it highly doubtful that the current species-distribution limits would represent the actual bioclimatic envelope of plant
species in Europe. Furthermore, these values may not be valid for climate states totally different from today, such as the mid-
Holocene; this also refers to the concept of no-analog communities (e.g. Williams and Jackson, 2007). The simulated tree
cover in areas occupied by forest biomes is largely dependent on the implemented disturbance extent, severity and interval.
However, the causes and consequences of disturbances are largely different in natural- and human-influenced landscapes.
660 The mid-Holocene natural forests were probably much more stable and less sensitive to disturbances than present-day forests
that are heavily altered by human interventions. We hypothesize that the large underestimation of the tree fraction on the
British Isles and Central Europe in JSBACH compared to REVEALS is a consequence of too much wind throw in the
model. In contrast, the overestimated tree-cover fraction in LPJ-GUESS may at least partly be related to a too small spatial
scale and frequency of disturbance occurrences in LPJ-GUESS. We tested this hypothesis in a sensitivity study in which we
665 extended the MPI-ESM1.2 spin-up run for 8 ka with halved sensitivity of the trees to the wind throw (i.e. halving the wind
damage scaling parameter in JSBACH). For LPJ-GUESS, we performed a set of simulations changing the interval between
disturbance occurrences from 25 years to 50 years, 75 years and 200 years using the 8 ka climate forcing. The standard setup
used in the comparison with REVEALS and JSBACH above is 100 years.

Wind throw reduction substantially increases the tree cover simulated by JSBACH in a broad area in mid-Europe (48-58°N), including the regions of the British Isles and Central Europe, and along the Norwegian Atlantic coast (Fig. 8). These are also the regions in which JSBACH substantially deviates from REVEALS and strongly underestimates tree cover in the transient simulation. However, in the other regions, only few REVEALS grid cells are affected, thus in regional means over the grid cells, the effect of wind throw is smaller.

The disturbance frequency tests with LPJ-GUESS show that each reduction of the occurrence interval by 25 years leads to approximately 4% less tree cover. This reduction is mostly related to a decrease in the cover of the evergreen PFTs (Fig. 8). The deciduous PFTs are not so heavily affected, as the disturbance gives some advantage to the early-successional deciduous PFT (Table 2), parametrized keeping in mind quick establishment and growth. Therefore, the shortened disturbance interval leads to an increased representation of deciduous trees at the expense of evergreen ones in northern and eastern Europe (Fig. 8). While LPJ-GUESS-simulated total tree cover is in general already in rather good accordance with REVEALS reconstructions, especially during mid-Holocene, the conformity could be improved by increasing the disturbance frequency.

4.4 Effect of land use

European land cover has been affected by humans during most of the Holocene. Pollen-based studies show the first traces of crop cultivation in southern Europe (Eastern part of the Mediterranean basin) more than 10000 years ago and its spread to the southern fringe of northern Europe during the following six millennia (Githumbi et al., 2022a). This transition to agrarian subsistence led to substantial anthropogenic deforestation of large parts of Europe. Particularly early, the coastal areas of western Europe were strongly affected and had already lost half of the natural forest cover 3500-5500 years ago (Roberts et al., 2018). While pollen-based studies can give detailed insights into of land-cover development in an area, these records are often not quantitative, spatially discrete and do not have a high (preferably annual) resolution, all prerequisites of input datasets for vegetation models. New, pollen-based reconstructions by Githumbi et al., 2022b address most of the above-listed obstacles, providing excellent quality, quantitative and spatially discrete proxy-based continental scale land cover reconstructions for Europe during the Holocene. However, while these reconstructions are suitable for usage as land-cover representation in climate models or for validation of vegetation model performance, reconstructions do not distinguish between the natural and anthropogenic land-cover types and therefore cannot be directly used as an anthropogenic land cover change (ALCC) input to vegetation models. There are no attempts at using land use inferred from pollen-based REVEALS reconstructions of plant cover in DGVMs published so far. Most DGVMs use prescribed land use derived from various ALCC scenarios such as KK10 (Kaplan et al., 2009) and HYDE 3.2 (Klein Goldewijk et al., 2017). However, KK10, HYDE and other ALCCs exhibit large discrepancies in their estimates of the starting time, spatial pattern and intensity of anthropogenic land-cover change, making it a challenge to simulate the effect of human-induced vegetation changes with DGVMs (Kaplan et al., 2017; Gaillard et al., 2010). Here we used a preliminary version of the LUH2 dataset by (Hurtt et al.,

700 2020), that assumes no human interference with land cover prior to 2 ka. The increased disagreement between LPJ-GUESS and REVEALS for the British Isles and some Central and Boreal European sites for 4 ka and 2 ka is probably due to considerable anthropogenic deforestation of these areas already prior to 2_ka. During the last two millennia the overall agreement between the models (especially JSBACH) and REVEALS increases, showing the significance of accounting for anthropogenic deforestation.

705 In the models, substantially different approaches are used to handle land use. While in LPJ-GUESS, a certain grid-cell fraction is reserved for land use related land-cover types, JSBACH calculates natural vegetation first and afterwards applies land transitions with land-use types preferentially replacing grasslands. These differences in implementation of land use explain the larger impact of land use on LPJ-GUESS-simulated than JSBACH-simulated tree cover. LPJ-GUESS-simulated natural vegetation dynamics agree well with REVEALS-estimated plant cover at 8_ka and 6_ka, while JSBACH
710 underestimates tree cover. The more the landscape is affected by humans, the greater the differences in tree cover between LPJ-GUESS and REVEALS, since the full scope of actual land use in the past cannot be reproduced by the models. JSBACH simulates a rather open landscape and prescribed ALCC has - due to specifics of its implementation - relatively little impact on the tree cover fraction. The combination of these two characteristics leads to a convergence of REVEALS and JSBACH tree-cover over time and space through the last two millennia. LPJ-GUESS, differently from JSBACH, applies
715 the prescribed ALCC proportions directly, making the accuracy of the used ALCC dataset especially important. The increased disagreement of the models compared to REVEALS in the Alpine region suggests that a too strong anthropogenic deforestation has been prescribed in this area.

4.5 Caveats of the REVEALS model and pollen-based reconstructions of plant cover

Many of the assumptions of the REVEALS model are violated in the "real world" and/or violated in the past, which has been
720 described and discussed in detail earlier (e.g., Hellman et al., 2008a; Sugita et al., 2010; Mazier et al., 2012; Trondman et al. 2015; Li et al., 2020). Major assumptions are: a) there is no vegetation growing on the basin (i.e., REVEALS was developed for pollen records from lakes), b) wind comes from all directions and wind speed is constant through time, and c) pollen productivity of plant taxa and fall speed of pollen are constant through time. Moreover, the REVEALS model is not well suited for pollen-based reconstructions in mountainous areas, which is discussed in Marquer et al. (2020). Given that the
725 model does not account for topography, a flat topography can be considered as a model assumption, although not mentioned as such in Sugita (2007). These assumptions are violated to various degrees. Bogs are covered by vegetation, and therefore violate a major assumption of the REVEALS model. Pollen records from bogs are used in the continental REVEALS reconstructions (e.g., Githumbi et al., 2022a) and in the dataset of Marquer et al. (2017) that have been used in this study. The decision to use these pollen records is motivated by a) using only pollen records from lakes would decrease significantly
730 the number of available pollen data, and b) Trondman et al., (2016) showed that pollen records from small bogs can be used

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The effects of the violation of assumptions cannot be quantified and accounted for, i.e., REVEALS estimates cannot be corrected for these effects. However, in the context of this study, they canTherefore, possible effects of the violation of assumptions must be considered as possible causes behind discrepancies between REVEALS estimates of plant/PFT cover and DGVMs simulated PFT cover. The fact thatBecause topography is not accounted for considered in the REVEALS model and reliable reconstructions in mountainous regions require a large number of pollen records from large lakes representing the major altitudinal vegetation zones (Marquer et al., 2020), the implies that REVEALS estimates used here need to be considered as uncertain in the Scandinavian mountains and the Alps, are uncertain which and may explain discrepancies with DGVMs at the grid-cell scale level in these areas (Fig. 3). At the regional scale, REVEALS, LPJ-GUESS and JSBACH agree when standard deviations are considered (Fig. 4). REVEALS tree cover for the grid cell with four pollen records from small bogs (Britain; Fig. 2) may be biased towards local plant cover on the bogs (i.e., overestimated cover of open land) because there are no additional pollen records from large lakes (or several small lakes) in the grid cell that would "correct" the mean REVEALS estimate towards less open land. This bias could contribute to the significant discrepancy between REVEALS and the two DGVMs for this grid cell.

Beside the violation of model assumptions, there are other features of REVEALS reconstructions that may play a role in the REVEALS-DGVM comparison. REVEALS reconstructs the cover of individual plant taxa rather than the cover of PFTsplant functional types or land-cover types (such as "open land"). When the REVEALS taxon-based estimates are summed into PFTs or land-cover types, the cover of individual plant taxa cannot be distributed between several PFTs or land-cover types. This shortcoming should not affect the comparison with the PFT cover simulated by both DGVMsmodels in this study, given that the simulated woody PFTs include only woody plants, and open land only herbs. Furthermore,

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~~A~~acceleration of the development of *Calluna* heathland at the expense of woodland from 6 ka <u>in coastal areas of westernmost Europe</u>~~on~~ <u>is</u> due to land use (grazing and burning)~~in coastal areas of westernmost Europe~~ (Nielsen et al., 2012)

765 <u>and</u> could explain some discrepancies in the cover of open land between REVEALS and the DGVMs, however at the grid-cell scale level ~~only~~ (Fig. 3). The REVEALS cover of *Calluna* is ~~indeed~~ between 10 and 60 % from 6 ka to 1 ka in several grid cells of the British Isles, Denmark and southern Norway (Trondman et al., 2015; Marquer et al., 2019).

~~Another feature of t~~<u>T</u>he REVEALS model ~~is the~~ use<u>s</u> of pollen data <u>to reconstruct plant cover and, therefore, only plants producing a sufficient amount of pollen to be recorded by pollen analysis will be reconstructed</u>~~implying that the model~~

770 ~~reconstructs the cover of plants that produce pollen~~. This <u>limitation</u> is mainly critical in the case of trees and shrubs that may start to produce pollen after many years. Therefore, the cover of young trees is not included in REVEALS tree cover that may therefore underestimate <u>actual</u> tree cover. However, REVEALS tree cover is never significantly lower than the tree cover simulated by the DGVMs, except for the British Isles where tree cover is larger for LPJ-GUESS than for both JSBACH and REVEALS in all time windows except 1 ka and PI. ~~Finally,~~ <u>Another limitation of</u> the REVEALS

775 reconstructions <u>is the availability of values of relative pollen productivity (RPP) for the plants documented in the pollen records. At the time of the REVEALS reconstruction used in this study (Marquer et al., 2017), RPP values were available for 25 plant taxa (excluding most entomophilous plants to avoid violation of the model assumptions "pollen is transported by wind" (Sugita, 2007)). Nevertheless</u>~~used in this study are based on 25 plant taxa only. However~~, the pollen types ascribed to the~~se~~ <u>25</u> plant taxa represent > 90% of the pollen counts and most missing taxa belong to <u>less abundant</u> herbs that would not

780 decrease the REVEALS tree cover very significantly.

In summary, ~~there are few~~ <u>violation of model assumptions and other</u> caveats of the REVEALS model itself <u>on the one hand</u> and the REVEALS dataset used in this study <u>on the other hand</u> ~~that~~ <u>need to be considered as possible</u> ~~can~~ contribut<u>ion</u>~~e~~ to the discrepancies between REVEALS-estimated and DGVM-simulated tree cover. <u>However, T</u>~~t~~he discrepancy between REVEALS and LPJ-GUESS in the British Isles is obviously due to the significant land <u>-</u>use in this region from 6 ka on and

785 the lack of wetlands in <u>the DGVMs</u>~~the model~~. The mismatch of JSBACH in Central Europe, Boreal Europe, and the British Isles, i.e. the underestimation of forest cover in comparison to LPJ-GUESS and REVEALS, indicate<u>s</u> that the implementation scheme ~~(and the tuning) of JSBACH is problematic for these regions in particular.

5 Summary and Conclusions

We compare pollen-based quantitative reconstructions of Holocene tree cover in Europe estimated by REVEALS with a

790 transient simulation of the last 8000 years undertaken with the Earth System Model MPI-ESM1.2 (including the dynamic vegetation model JSBACH) and time-slice simulations conducted with the ~~dynamic vegetation model~~ <u>DGVM</u> LPJ-GUESS. Both models and the reconstructions indicate larger tree cover in most parts of Europe at 8 ka compared to PI but differ

substantially with respect to the total area covered by trees and the age of the start of deforestation. While LPJ-GUESS generally overestimates tree cover fractions compared to REVEALS, JSBACH indicates much lower percentages of forested area in most parts of the region, albeit with a similar spatial pattern as LPJ-GUESS.

The total area covered by trees is relatively constant in the models until the prescribed land-use sets in, i.e. after the 2 ka time slice. In contrast, REVEALS indicates a 6 ka maximum in tree cover in some grid-cells in Central and Boreal Europe and particularly in the Alpine region. A comparison of the simulated climate with chironomid-based climate reconstructions reveals that climate biases only marginally cause these disagreements between the simulated and reconstructed trend in tree cover. Instead, the reconstructed 6 ka maximum in some areas may be related to dispersal and migration-induced delays in the establishment of some tree taxa. These processes are not included in the models.

According to REVEALS, anthropogenic deforestation starts much earlier (~4 ka in the Alpine region and 6 ka in Central Europe and the British Isles) than in the model forcing. While the decline in the tree cover fraction in REVEALS is relatively steady, the prescribed land use induces a sharp drop in tree cover in most regions in the models, indicating a too intensive land use in central and western Europe, particularly in the Alpine region. Prescribed land use in Boreal Europe seems to be too weak in the DGVMs~~models~~ compared to the REVEALS estimates. The prescribed land use has a larger effect in LPJ-GUESS than in JSBACH, pointing to differences in the implementation of land -use and the higher tree cover level at the onset of deforestation in LPJ-GUESS compared to the generally more open landscape in JSBACH. Thus, the differences between the DGVMs~~models~~ and REVEALS in the Late Holocene trend can clearly be attributed to the incorrect appearance of anthropogenic deforestation in the models, contributing to the overestimation of tree cover in LPJ-GUESS.

The strongest differences between the models with respect to the total tree cover occur for the British Isles, Central Europe and southern Europe as well as the Atlantic coastal regions, in which JSBACH simulates small tree-cover fractions at all time slices. Both DGVMs~~models~~ show spatial differences in the agreement with the REVEALS results. Whereas LPJ-GUESS indicates relatively good agreement with REVEALS in Central Europe, JSBACH exhibits the strongest mismatch with the REVEALS reconstructions for all time slices. This is partly caused by a too strong wind throw in JSBACH that substantially reduces the simulated cover fraction of trees in large parts of Europe as shown by additional sensitivity experiments. The strength of the effect of disturbances on the vegetation is static and calibrated to modern conditions, like most of the parameters in vegetation models. However, the mid-Holocene natural forests were probably much more stable and less sensitive to disturbances than the heavily human-altered present-day forests. This would justify and require the use of a lower storm vulnerability of forest in JSBACH for this period. Thus, whether ~~these~~ the modern climate-derived model parameter values may be valid for the entire transient simulation, is questionable.

Both vegetation models fail to reproduce the tree-cover changes in (mountainous) Northern Scandinavia and on the British Isles. For the other regions, the degree of agreement varies with time. LPJ-GUESS exhibits the overall best agreement with the REVEALS reconstructions at 6 ka, while JSBACH agrees best with REVEALS at PI. In most regions, the models are not able to simulate the correct ratio of deciduous to evergreen trees. In the mean, the distribution of evergreen trees agrees better between the models and REVEALS than the distribution of deciduous trees, except for eastern Europe. A steady improvement of the agreement of the deciduous tree cover in JSBACH with REVEALS leads to a reduction of the model-data mismatch towards PI. In contrast, the misrepresentation of land-use history in the models, i.e., a substantial anthropogenic deforestation prior to 2 ka, leads to a worsening of the agreement between LPJ-GUESS and REVEALS with time. Consequently, the model results converge towards PI.

Our study highlights the fact that model settings that are tuned for present-day conditions may be inappropriate for palaeo-simulations and complicate model-data comparisons with additional challenges. Moreover, our analysis identifies land-use as the main driver of the decrease in forest cover in Europe during the mid- and Late Holocene, as has been suggested by pollen studies and the more recent efforts to quantify pollen-inferred changes in plant cover (e.g. Robert et al., 2018) and the various scenarios of anthropogenic land-cover change developed over the last ca. 20 years (e.g. HYDE and KK10, see synthesis in Gaillard et al., 2010). Mid- and Late Holocene changes in climate have only a minor effect on forest cover, although changes in cover and distribution of individual plant taxa depend on both land use and climate (Marquer et al., 2017).

Tables

Table 1: Overview of selected time-slices and their acronyms in this study. The MPI-ESM1.2 periods used as climate forcing for the LPJ-GUESS model are provided in both, model years (1001 = 7949 BP) and corresponding calibrated ^{14}C years BP. The time windows for the pollen-based REVEALS -estimates of regional plant cover follow the standard protocol used in PAGES LandCover6k (Marquer et al., 2019).

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Acronym	MPI-ESM 1.2 model years	MPI-ESM period [BP]	REVEALS time window	REVEALS period [BP]
8 ka	1090-1220	7860-7730	18	8200-7700
6 ka	2881-3000	6069-5950	14	6200-5700
4 ka	4870-5000	4080-3950	10	4200-3700
2 ka	6960-7078	1990-1872	6	2200-1700
1 ka	8000-8130	950-820	4	1200-700
PI	8660-8780	290-170	2	350-100

Table 2: Assignment of the plant taxa used in the pollen-based REVEALS reconstructions to the plant-functional types (PFTs) in the Dynamic Global Vegetation Models LPJ-GUESS and JSBACH.

REVEALS taxa	LPJ-GUESS PFT set	JSBACH PFT set	Land-cover type
Abies	Temperate needleleaved evergreen tree (TeNE)	Extra-tropical evergreen trees	
Picea	Boreal needleleaved evergreen tree		
Pinus	Boreal shade-intolerant evergreen tree		
Carpinus	Shade-tolerant temperate broadleaved summergreen tree		
Fagus			
Tilia			Forest
Ulmus			
Alnus	Shade-intolerant broadleaved summergreen tree	Extra-tropical deciduous trees	
Betula			
Corylus			
Fraxinus			
Quercus			
Salix			
Juniperus	Boreal evergreen shrub	cold shrubs	
Calluna vulgaris			
Artemisia	Cool (C3) grass (grasslands and pastures)	C3 grass (grasslands and pastures)	Open
Cyperaceae			
Filipendula			
Gramineae			
Plantago lanceolata			
Plantago media			
Plantago montana			
Rumex acetosa-t			
Cerealia-t	Crops (Triticum spp., Hordeum vulgare, Secale cereale , Avena	Crops	

Secale-t	sativa, "TeWW")
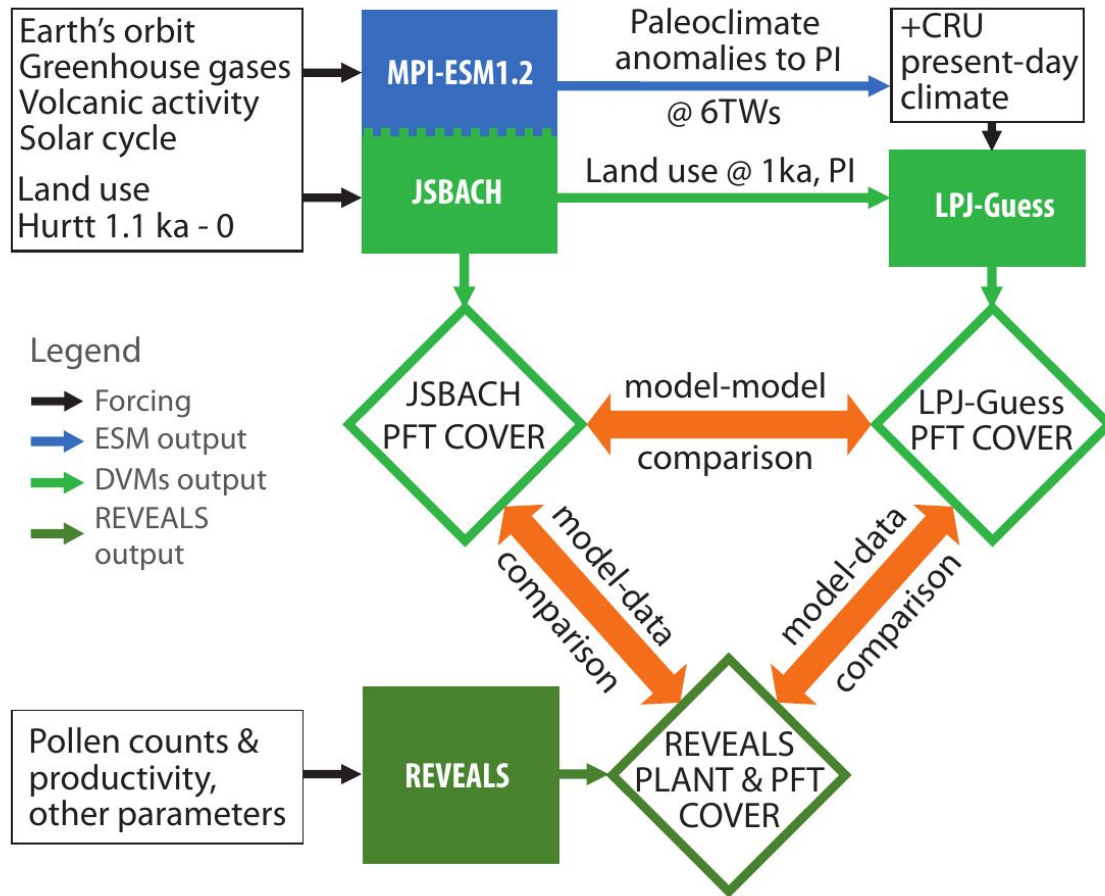
Figures

Figure 1: Flow chart of the strategy for the comparison between the plant functional type (PFT) cover simulated by the Dynamic Global Vegetation Models JSBACH (interactively coupled in the Earth System Model MPI-ESM1.2) and LPJ-GUESS (standalone model) and the pollen-based REVEALS plant-cover reconstructions. The MPI-ESM1.2 simulation and the REVEALS-based reconstructions have been published earlier in Bader et al., (2020) and Marquer et al. (2019), respectively. Within this study, LPJ-GUESS simulations for six different time-windows (TWs) were performed and compared with the MPI-ESM1.2 results and REVEALS reconstructions. As land-use forcing for JSBACH, a preliminary version of the LUH2 dataset by Hurtt et al. (2020) was used. The climate and land-use forcings for LPJ-GUESS were extracted from the output of the MPI-ESM1.2 model, but to overcome temperature biases due to the coarse spatial resolution, the MPI-ESM simulated climate anomalies to PI were interpolated bilinearly to a 0.5°x0.5° grid and added to the observational CRU-dataset (Harris et al., 2020) before prescribing them to LPJ-GUESS.
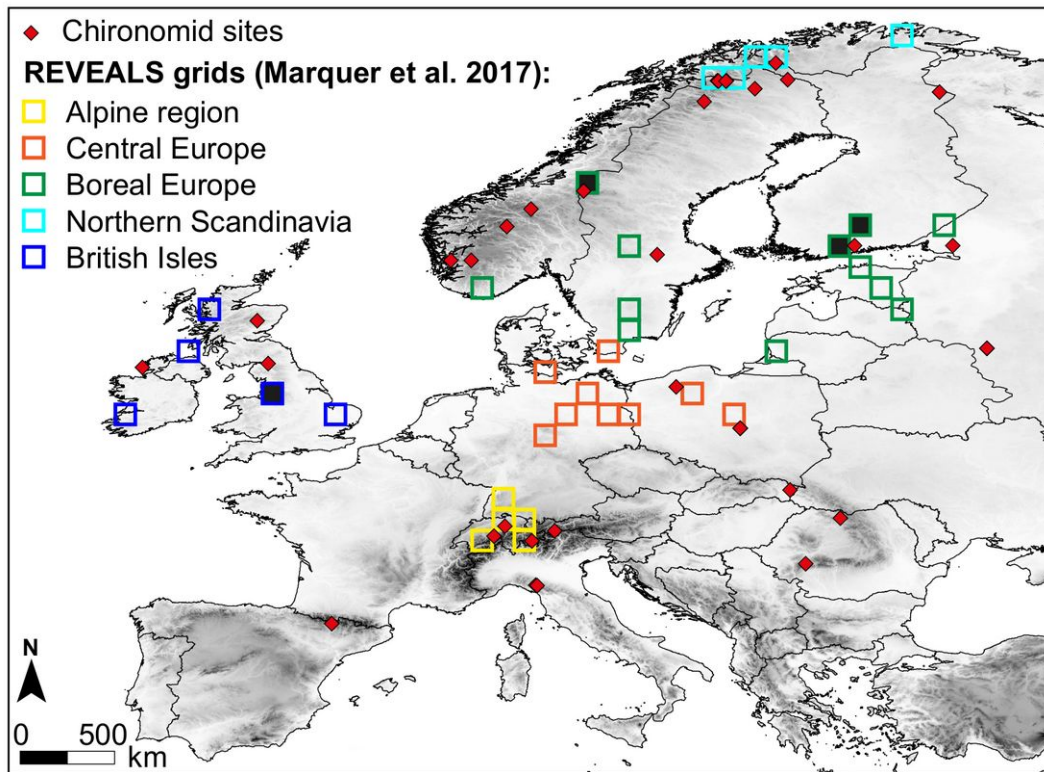
Figure 2: Grid cells with REVEALS estimates of plant abundances, grouped into five biogeographical regions (for detailed definition see Marquer et al., (2017), see figure legend for the names of the regions) and sites with chironomid data (red diamonds). Note that the four REVEALS grid cells that are colour-filled include only two small sites or one large bog (with few additional small sites) and are, thus, less reliable (cf. Sec. 2.3.2 for explanation).
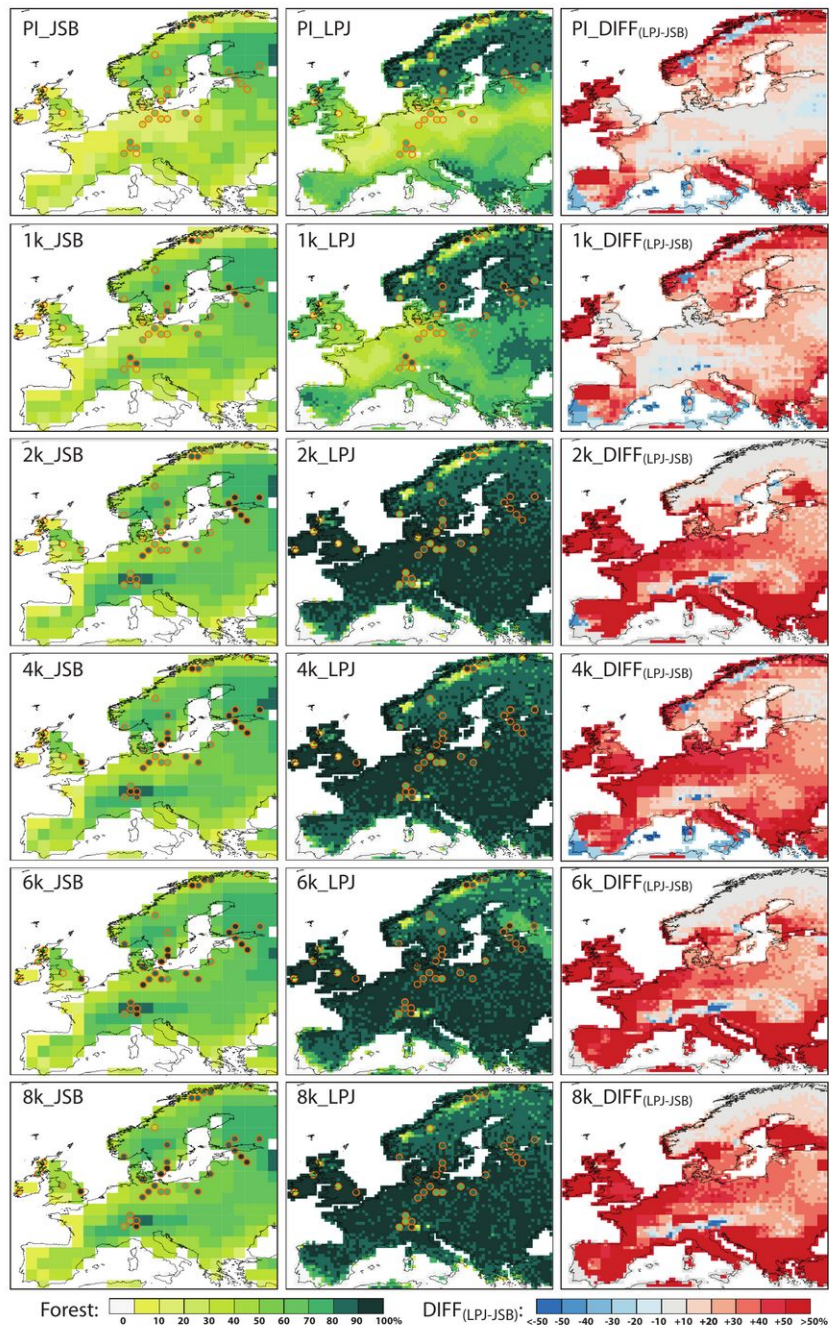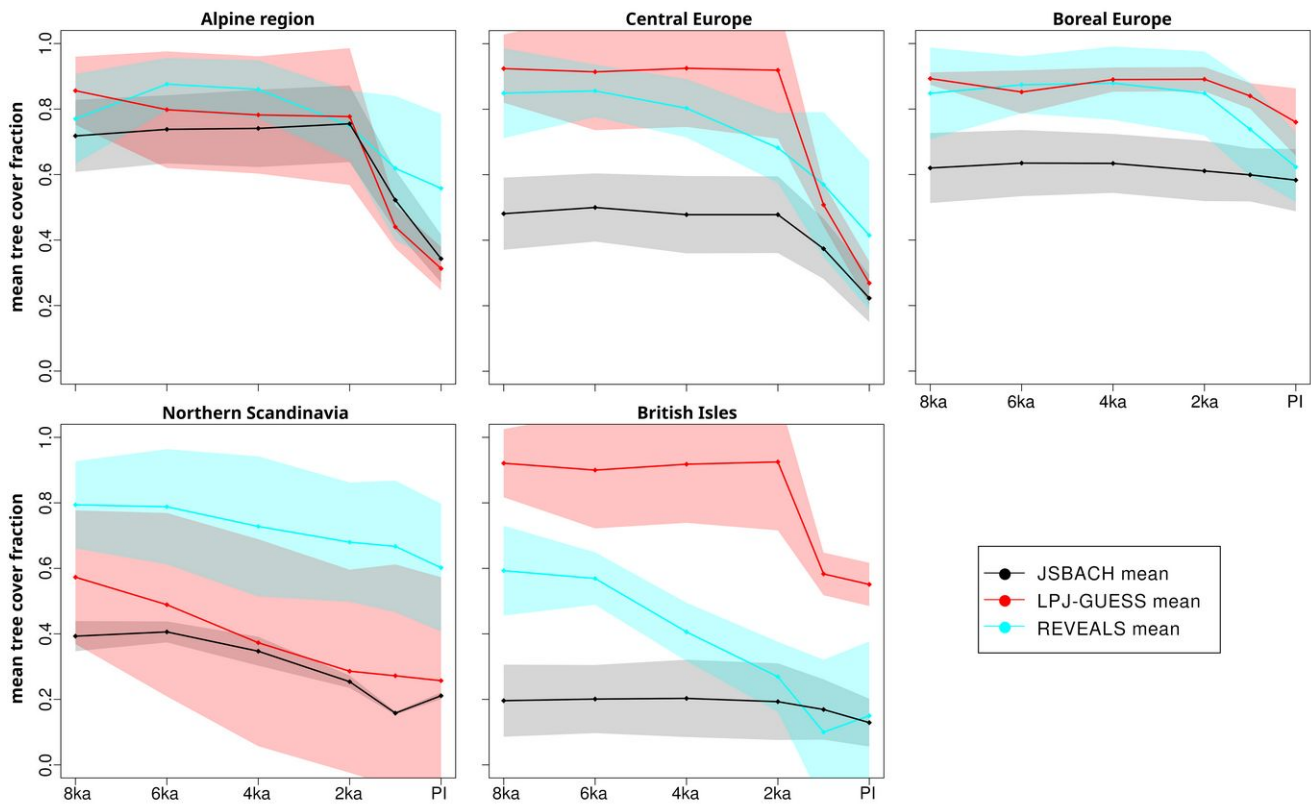
Figure 3: Total tree-cover fraction (in absolute fraction of the grid-cells) for six time-slices simulated by JSBACH (JSB) (left) and by LPJ-GUESS (LPJ) (centre) and the model difference (right). The pollen-based REVEALS tree cover (dots) is superimposed on the maps with the same colour scheme as the model results.

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Figure 4: Model-simulated and pollen-based REVEALS mean tree-cover fraction averaged over the five regions as displayed in Fig. 2, JSBACH (black), LPJ-GUESS (red) and REVEALS (cyan). The inter grid-cell +/- standard deviation is shown with shading. Note that the standard errors on the REVEALS estimates are not considered. Time-series are plotted based on six time-slices (dots) and linearly interpolated.
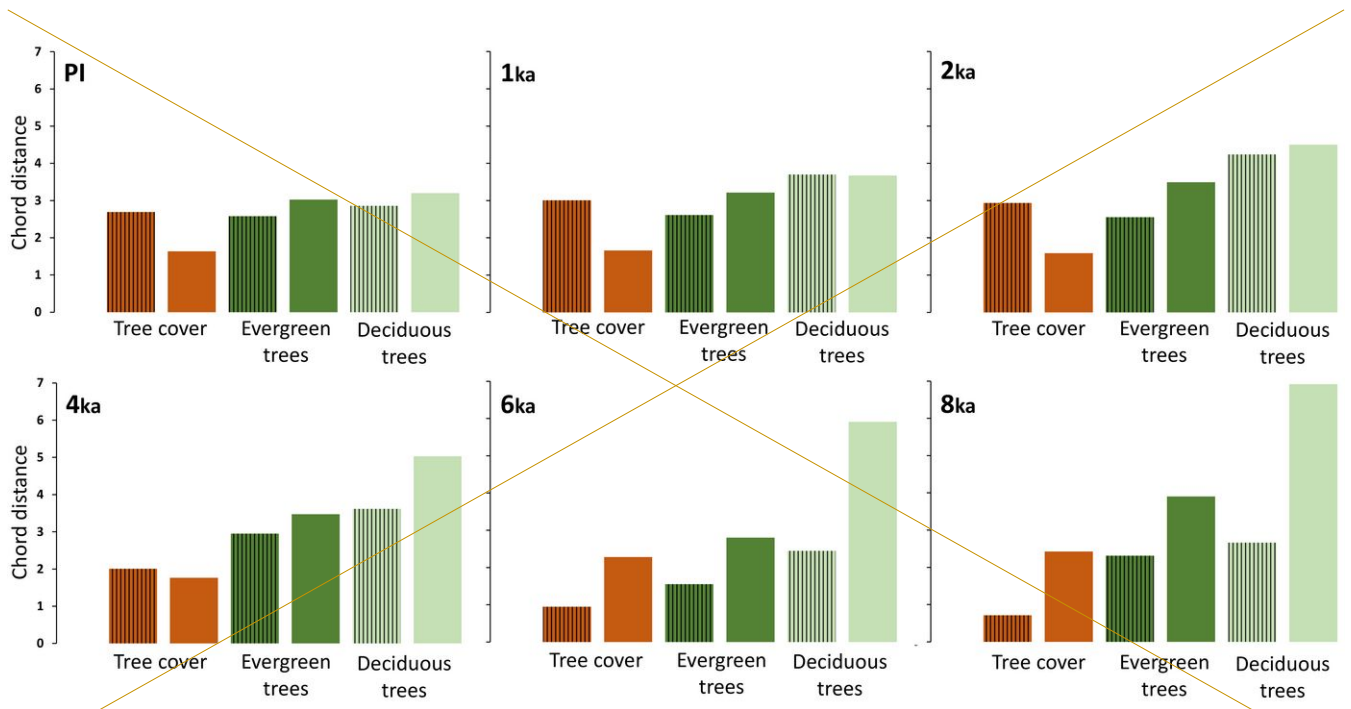
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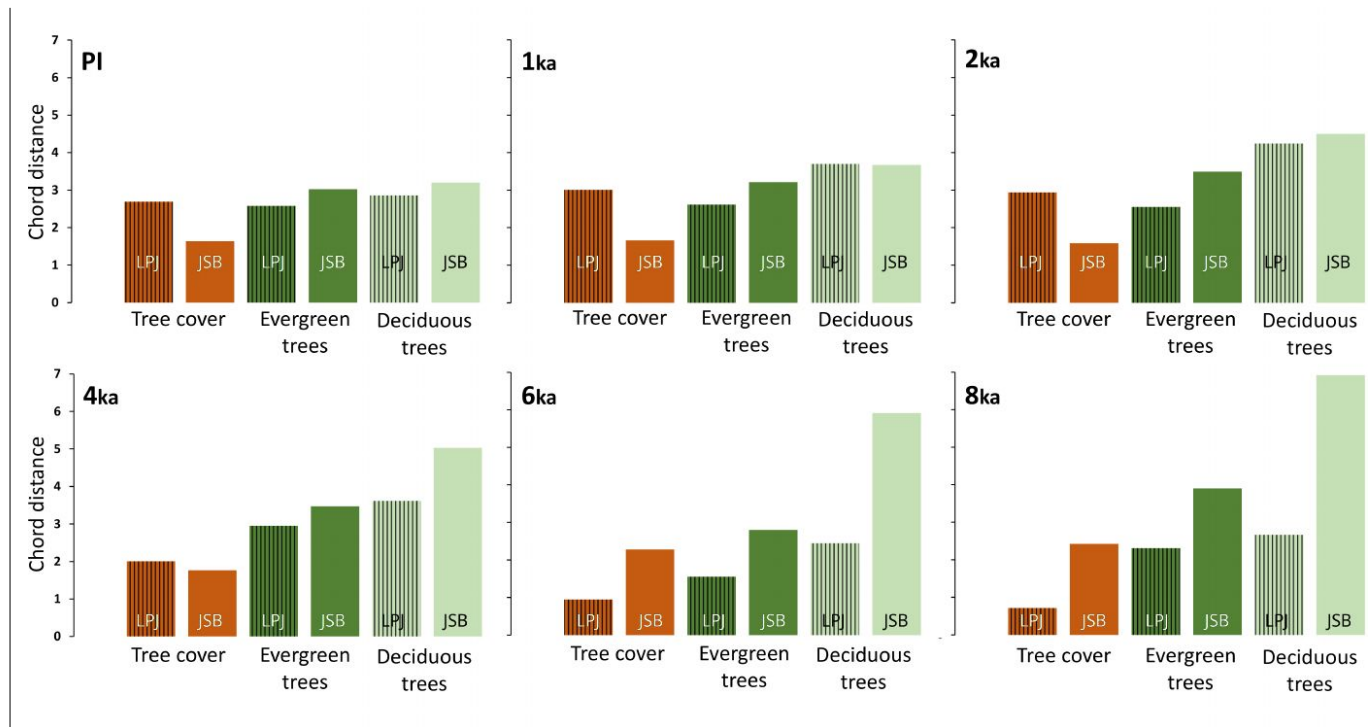
Figure 5: Chord distance between JSBACH and REVEALS (no pattern) and between LPJ-GUESS and REVEALS (vertical stripes) for total tree cover (brownish), evergreen trees (dark green) and deciduous trees (light green) for six time-slices.

885

REVEALS$_{JSBACH}$Forest

- agreement
- no agreement
- good agreement
- NO DATA

REVEALS$_{LPJ-Guess}$Forest

REVEALS$_{JSBACH}$TET

REVEALS$_{LPJ-Guess}$TET

REVEALS$_{JSBACH}$TDT
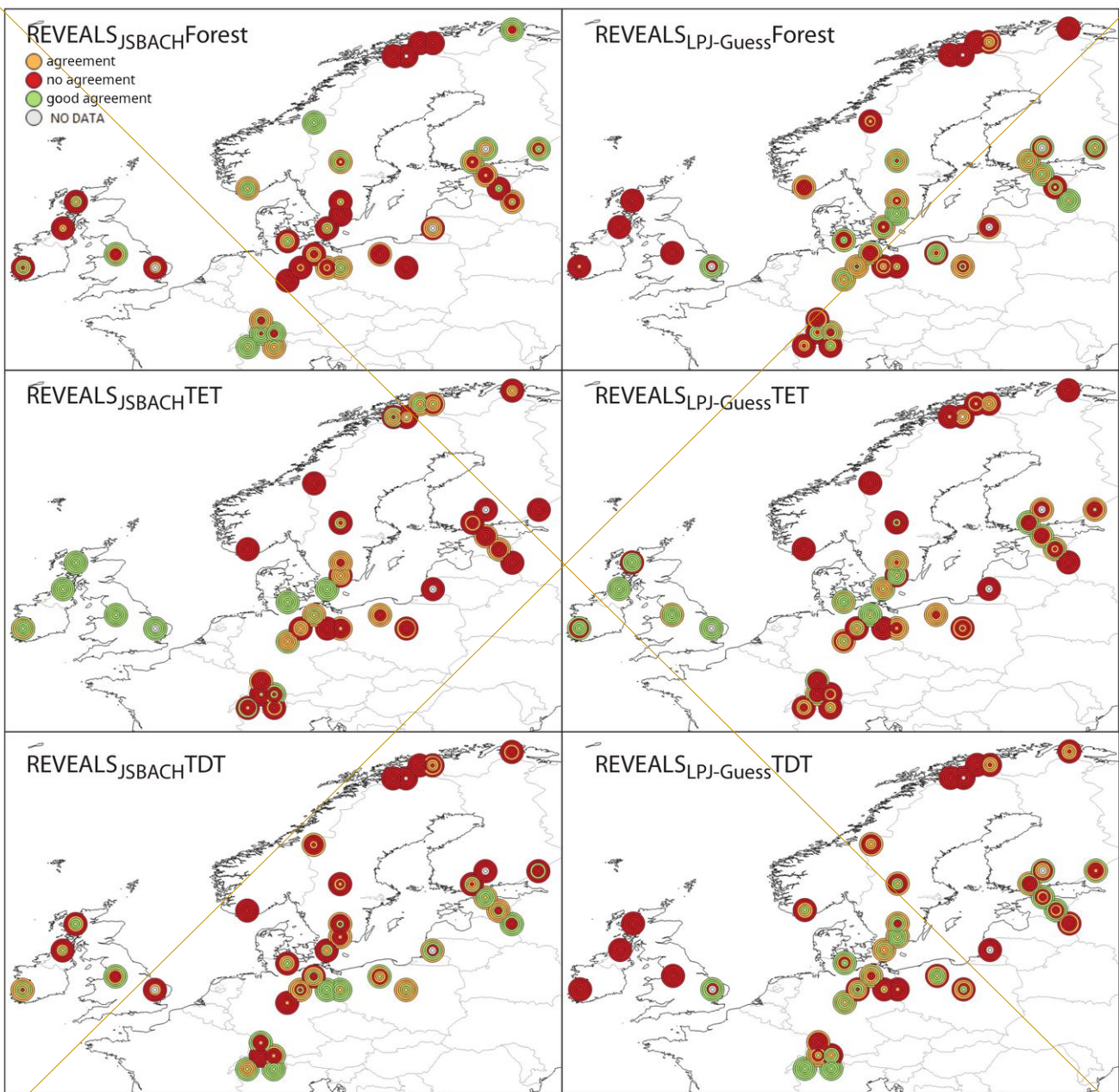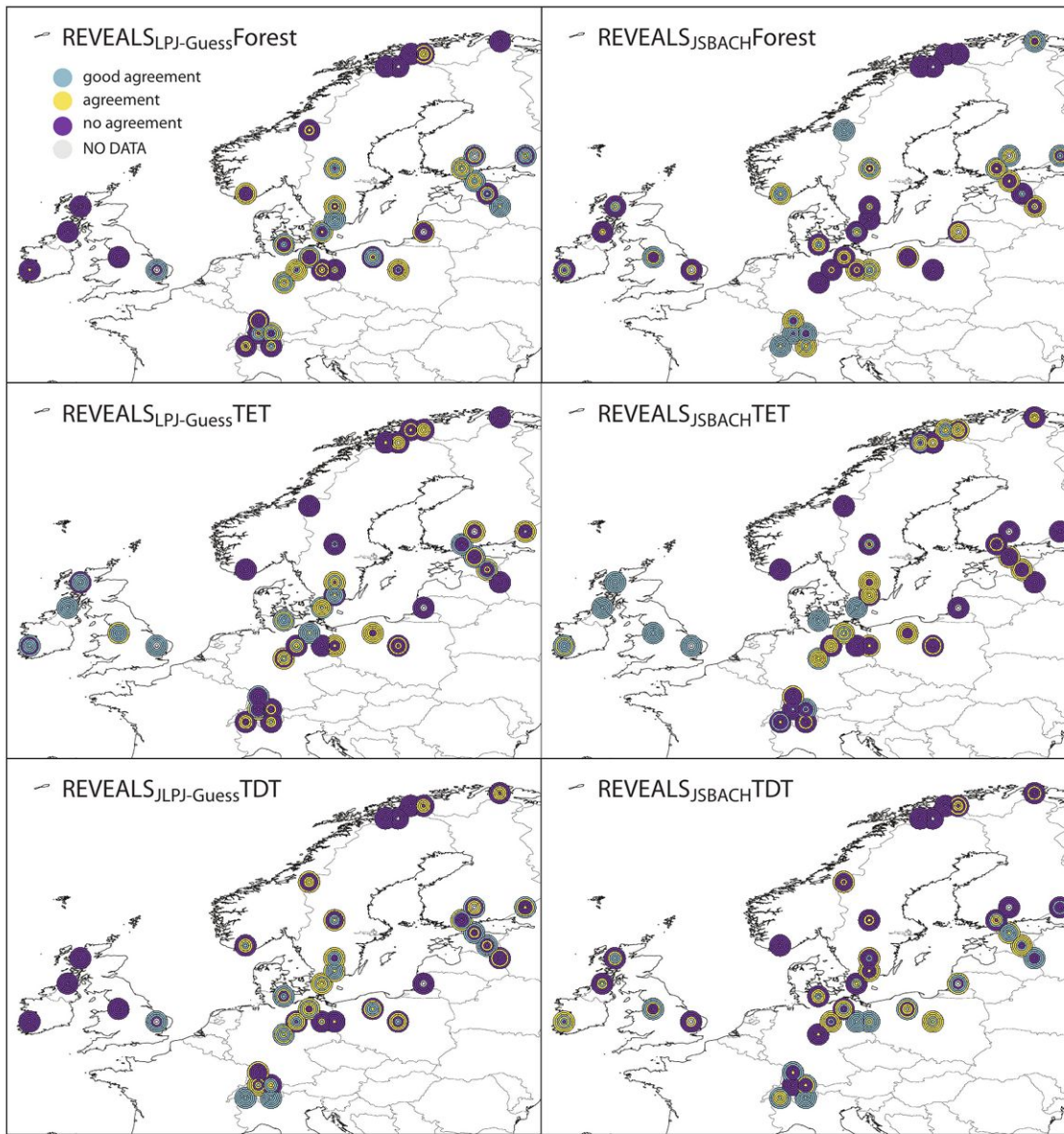
REVEALS$_{LPJ-Guess}$TDT

Figure 6: Agreement between REVEALS and LPJ-GUESSJSBACH (left) and REVEALS and JSBACHLPJ-GUESS (right) for total tree cover (Forest, upper panel), evergreen trees (TET, centre) and deciduous trees (TDT, lower panel) (see legend for colours). The plot is based on a three-scale agreement index to quantify and evaluate the spatial differences between the model-simulated vegetation and the pollen-based REVEALS plant cover in each of the REVEALS grid cells (see Methods for details). The six circles display the results for the six time slices, from the oldest (8 ka, outermost circle) to the younger one (PI, innermost circle).
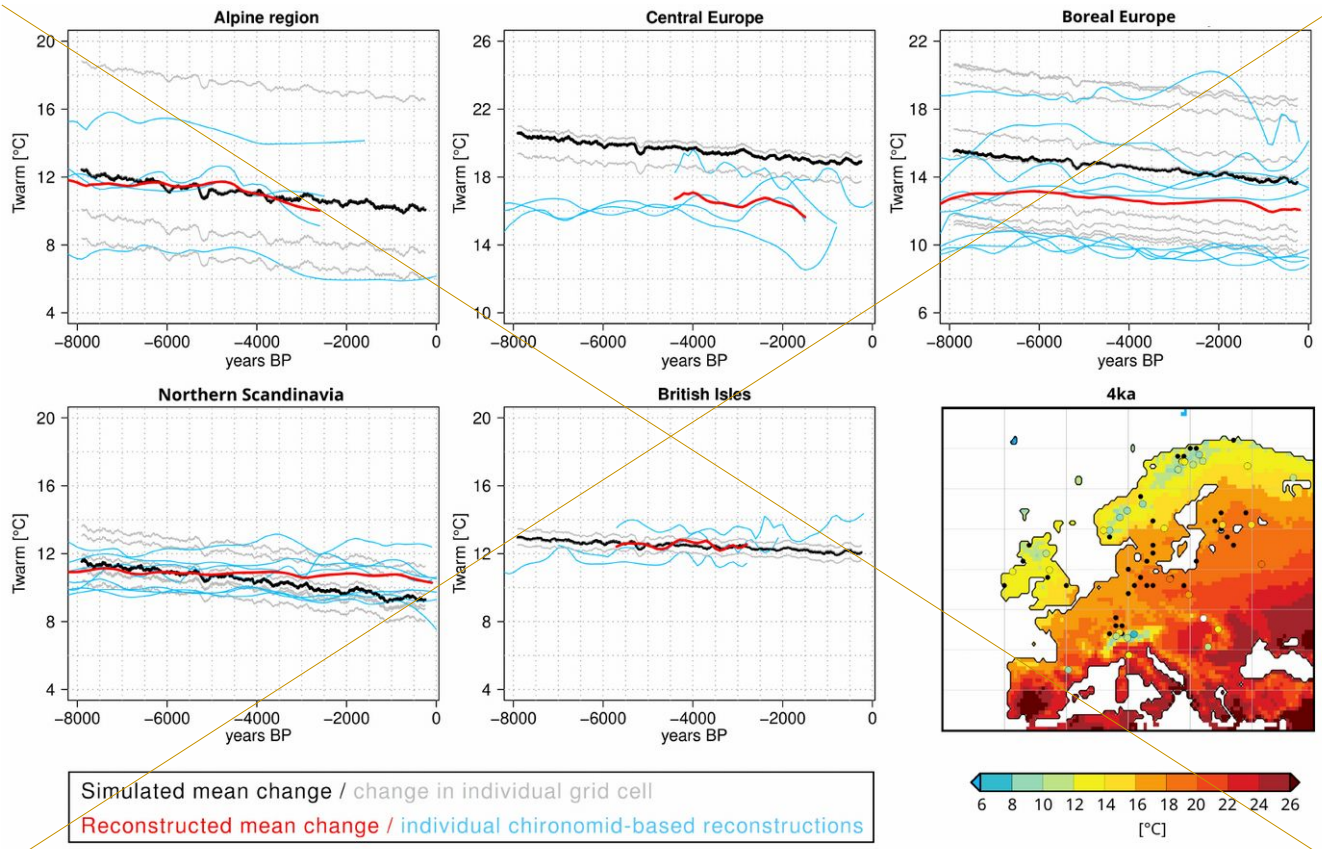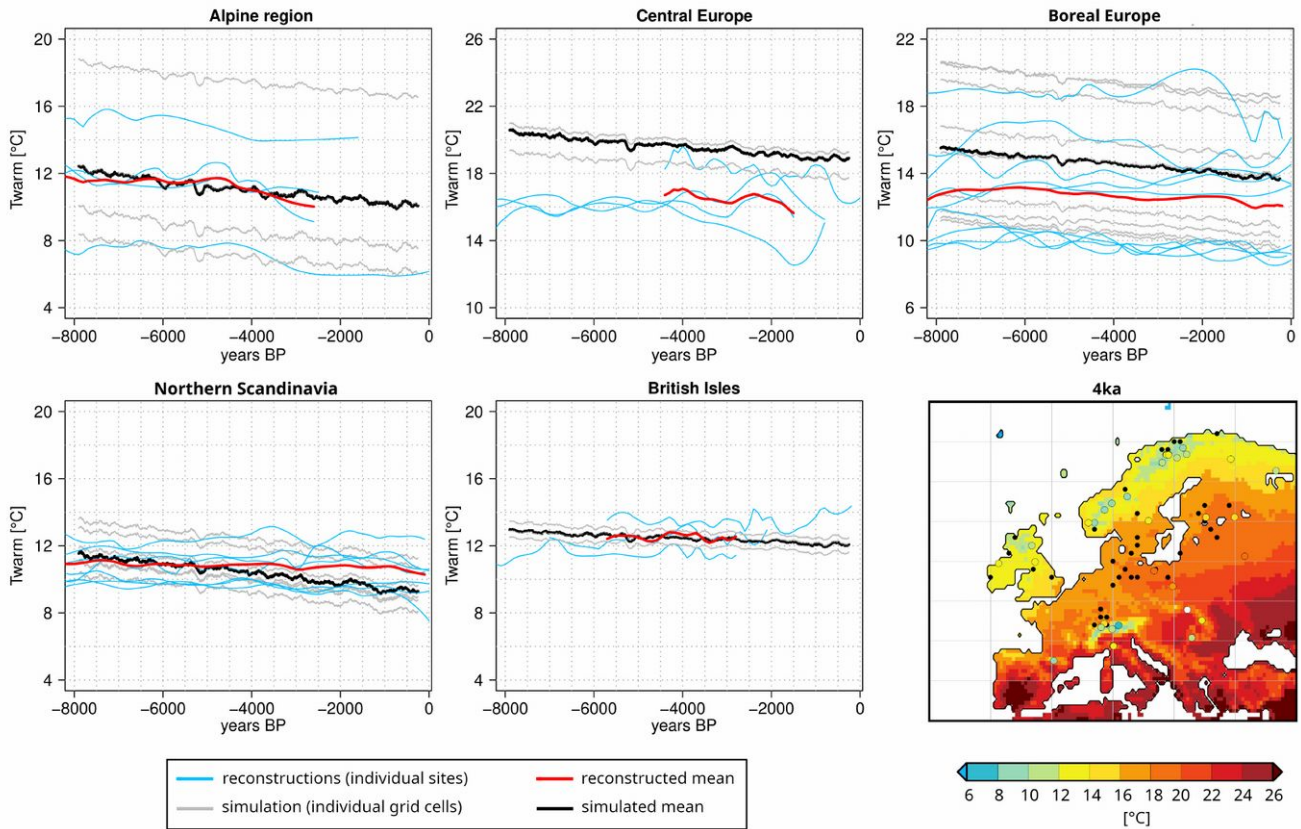
Figure 7: Comparison of the LPJ-GUESS climate forcing data (i.e. MPI-ESM1.2-simulated ~~Holocene change~~ in temperature of the warmest month (Twarm, [°C]) added as anomaly (from PI) to the CRU TS 4.0 datase t (Harris et al., 2020) and smoothed by a 200-year running mean), and chironomid-based temperature reconstructions (smoothed and interpolated on an equally distant time -axis) for the study regions (Fig. 2) (see Tab. B1 in ~~the~~ Appendix B for further details on the reconstructions). For each region, we have plotted the simulated ~~changes in~~ Twarm in the individual grid cells with chironomid-inferred Twarm available (grey) and their mean (black), as well as the individual chironomid-inferred Twarm (blue) and their mean (red). In the bottom right panel, we show the model-simulated (gridded) and chironomid-inferred (colour dots) Twarm [°C] at 4 ka. The black dots represent the location of the grid cells with pollen-based REVEALS plant-cover reconstructions.
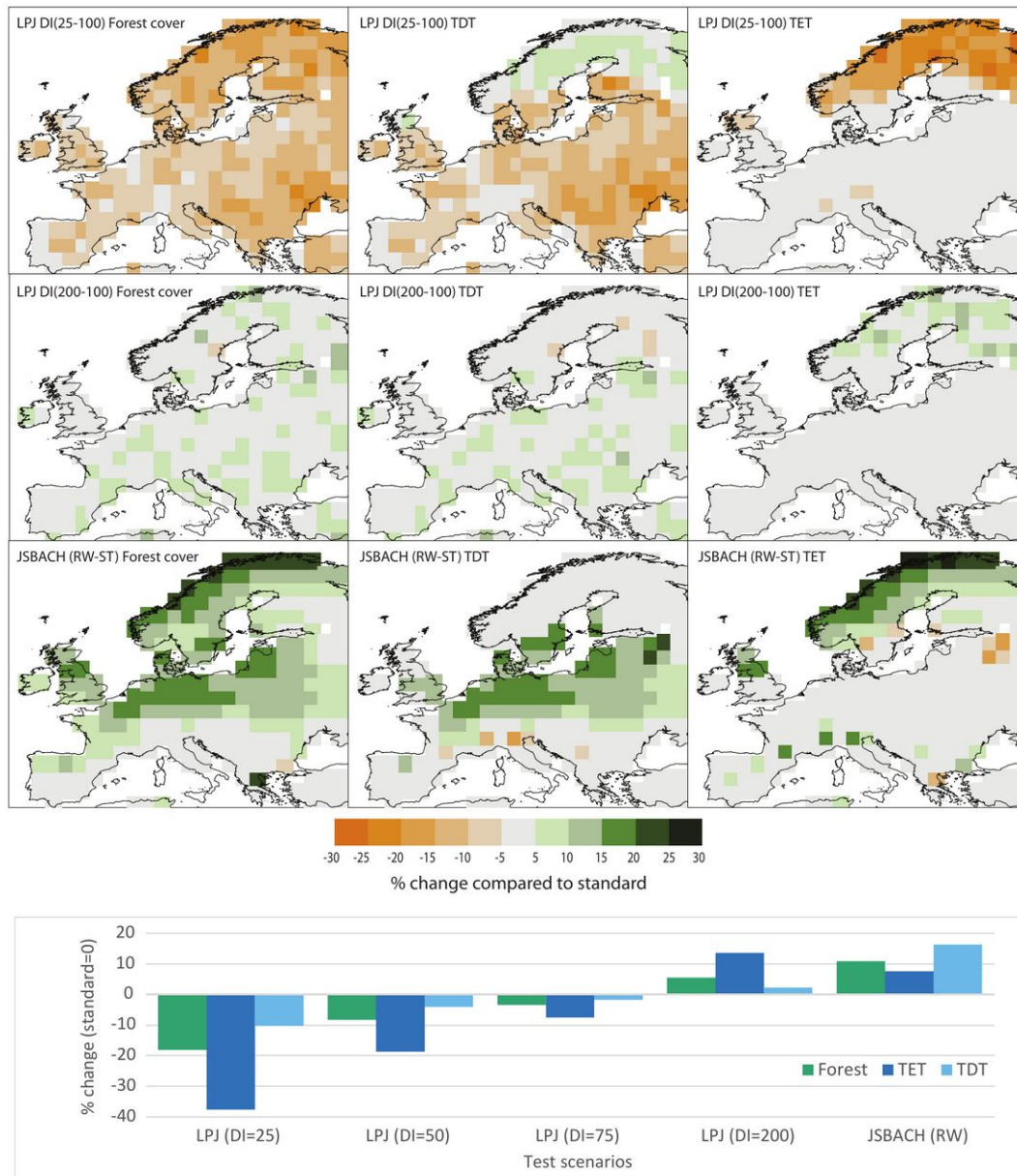
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Figure 8: Sensitivity test with different disturbance settings. Upper nine panels: tree cover difference [in absolute % of the grid-cell area] between LPJ-GUESS simulations with an occurrence interval of 100 years (standard setup) and 25 years (upper row) or 200 years (second row) and between the JSBACH simulations in the standard setup (ST, used in this study) and with halved wind damage scaling parameter (RW) (bottom row). For each setup, the differences in cover of total tree (Forest) (left column), deciduous tree (TDT) (mid column) and evergreen tree (TET) (right column) are shown. For LPJ-GUESS, occurrence intervals of e.g. 25 years means that stochastic patch-destroying disturbance occur once per 25 simulation years. Lower panel (bar plot): difference in simulated tree-cover fractions between the simulations with standard setup and the four different disturbance-occurrence intervals (DI) in LPJ-GUESS (LPJ) and between the standard and reduced wind-throw (RW) simulations in JSBACH, averaged over the entire region.

Appendix A Evaluation of the PI climate simulated by MPI-ESM1.2

To get at least partly rid of systematic model biases such as induced by the smoothed orography in the relatively coarse model grid of MPI-ESM1.2, we have used an anomaly approach to design the climate forcing fields for LPJ-GUESS. We have added the anomaly between a certain time -slice and the pre-industrial (PI) climatological mean simulated by MPI-ESM1.2 and have added this anomaly to observations (CRU TS 4.0, period 1901-1930, Harris et al., 2020). These modified climate states have then been used as forcing for LPJ-GUESS. Thus, the vegetation models experienced the same climate dynamics during the Holocene, but have a different reference state, i.e. CRU TS4.0 observations in LPJ-GUESS but PI climate in JSBACH. To infer the differences between these basic states, we evaluate the MPI-ESM1.2 PI climate with the CRU TS 4.0 dataset (Fig. A1). The differences in temperature correlate with the orographic pattern revealing the strongest mismatch in mountainous regions. Here, MPI-ESM1.2 generally calculates higher temperatures than observed due to much lower mountain heights in the coarse resolution used in the simulation. Furthermore, simulated summer and winter temperatures are underestimated by 1-2 K in large parts of central, eastern and northern Europe compared to the observations. South of 50°N temperatures are much too high in MPI-ESM1.2 during PI, particularly during summer in the Mediterranean domain. However, in the regions for which REVEALS estimates exist in (Marquer et al., 2017), temperatures differ only slightly between the CRU observations and MPI-ESM1.2.

The annual mean precipitation is strongly overestimated by MPI-ESM1.2 in most regions of the European continent, ranging up to 950 mm/year in southern Norway and 700 mm/year in central France and Spain. Precipitation levels are way too low around the Mediterranean Sea (up to 1250 mm/year) and along the West Atlantic Coast of the British Isles (up to 2000 mm/year) and Scandinavia (up to 1000 mm/year). However, we assume that in the regions analysed in this study (i.e. the regions with REVEALS reconstructions), the vegetation dynamic is driven by the temperature signal as inferred in another study based on a slightly different MPI-ESM1.2 simulation (cf. Dallmeyer et al., 2021). In the Mediterranean area, the deficit in precipitation and too warm climate in the model probably contribute to the underestimated tree coverage, but this region is not a core part of our study here.
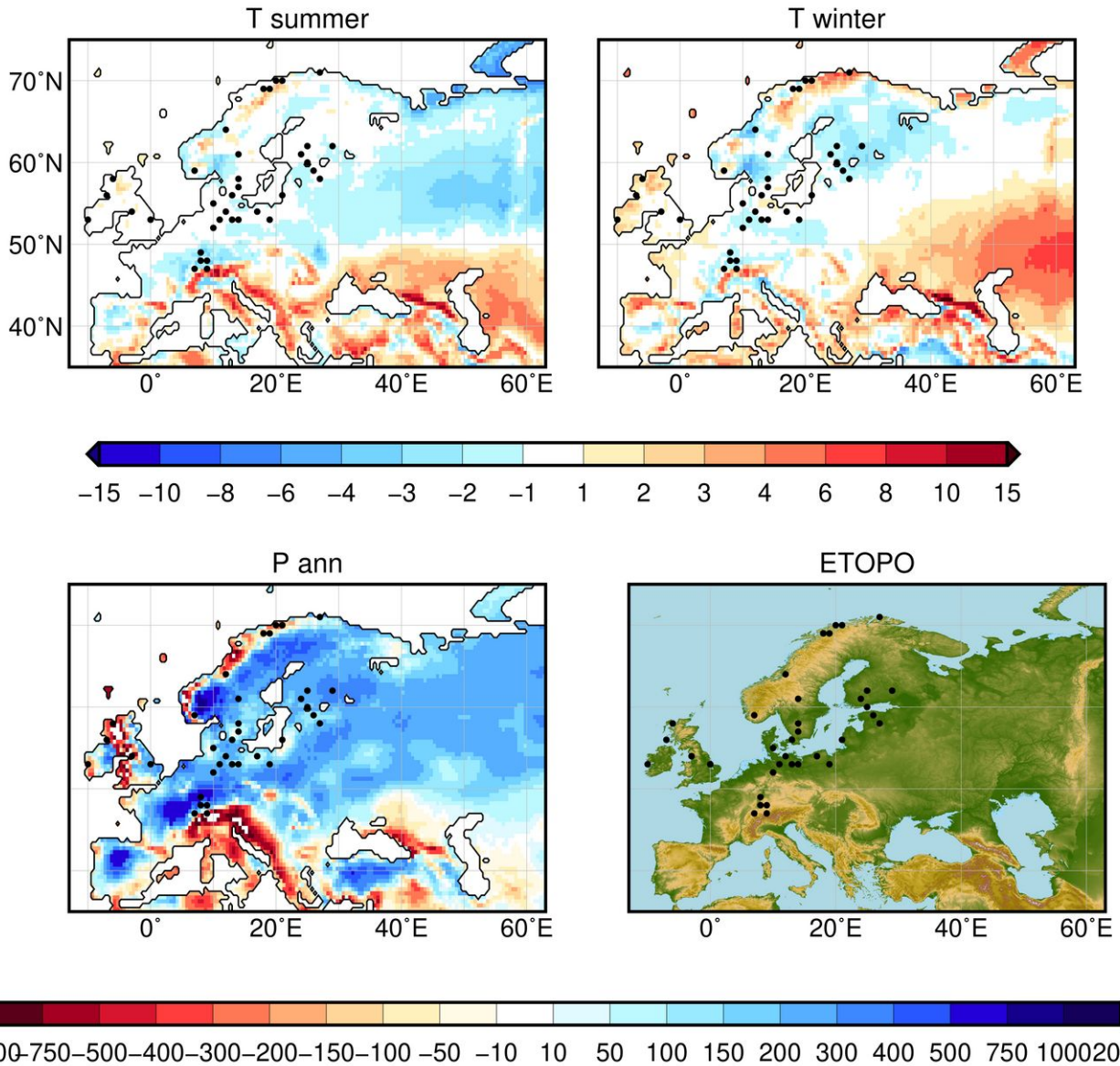
Figure A1: Difference between the pre-industrial climate simulated by MPI-ESM1.2 (bilinearly remapped on a 0.5 grid) and the CRU TS 4.0 dataset (1901-1930) (Harris et al., 2020) chosen as the basic state for the LPJ-GUESS climate forcing. Differences in summer temperature [K] (upper left), winter temperature [K] (upper right), and annual mean precipitation [mm/year] (bottom left) are shown. Bottom right: orography (the more brownish the higher the mountains) based on the ETOPO5 dataset (National Geophysical Data Center, 1993). The black dots display the grid-cells for which pollen-based REVEALS estimates of plant cover are available.

Appendix B List of the chironomid records used in this study

Table B1: List of chironomid records used in this study. Most of the chironomid-based reconstructions used in this study were extracted from the Kaufman et al. (2020) database. For these records, the data set name, site information and references (in form of the doi) were taken from the Temp12k_metadata table provided by Kaufman et al. (2020). For the other reconstructions (marked with *), the information was added accordingly.

Region	Data Set Name	Lat. (°)	Lon. (°)	Elev.(m)	Reference
	Stazersee.Heiri.2015	46,50	9,87	1809	10.1177/0959683614556382
Alpine	Hinterburgsee.Heiri.2015	46,72	8,07	1515	10.1191/0959683603hl640ft
Region	SchwarzseeobSoelden.Ilyashuk.2011	46,97	10,95	2796	10.1016/j.quascirev.2010.10.008
	Egelsee.Larocque.2010	47,18	8,58	770	10.1007/s10933-009-9358-z
	Zabieniec.Plociennik.2011	51,85	19,78	180	10.1016/j.palaeo.2011.05.010
	Zabieniec.Luoto.2019*	51,85	19,78	180	10.3354/cr01543
Central	Zabieniec.Kotrys.2020*	51,85	19,78	180	10.1111/bor.12406
Europe	Lake.Spore.Pleskot.2022*	53,80	16,73	<50	10.1016/j.palaeo.2021.110758
	M25.Smolensk.Mroczkowska.2021*	55,63	31,54	<200	10.3390/w13111611
	STIIIA.Smolensk.Płóciennik.2022*	55,63	31,54	<200	10.1016/j.catena.2022.106206
	VestreOykjamytjorn.Velle.2005	59,82	6,00	594	10.1016/j.quascirev.2004.10.010
	Holebudalen.Seppa.2009	59,83	6,98	1144	10.5194/cp-5-523-2009
Boreal	Gilltjarnen.Antonsson.2006	60,08	15,83	172	10.1002/jqs.1004
Europe	Hirvijaervi.Luoto.2010	60,51	25,23	104	10.1002/jqs.1417
	Medvedevskoe.Nazarova.2018	60,53	29,90	102,2	10.1134/S1028334X18060144
	brurskardstjorni.Velle.2005	61,42	8,67	1309	10.1016/j.quascirev.2004.10.010
	Ratasjoen.Velle.2005	62,27	9,83	1169	10.1016/j.quascirev.2004.10.010
	Spaime.Hammarlund.2004	63,12	12,32	887	10.1191/0959683604hl756rp
Northern	sjuuodjijaure.Rosen.2001	67,37	18,07	826	10.5194/cp-10-1605-2014
Scandi-	AlanenLaanijarvi.Heinrichs.2005	67,97	20,48	365	10.1111/j.1502-3885.2005.tb01015.x
navia	VuolepNjakajaure.Heinrichs.2006	68,33	18,75	409	10.1007/s10933-006-0010-x
	vuoskkujavri.Bigler.2002	68,33	19,10	348	10.1016/j.quascirev.2004.04.006
	850Lake.Shemesh.2001	68,37	19,12	850	10.1191/095968301678302887
	Njulla.Larocque.2004	68,37	18,70	999	10.1023/A:1022850925937
	Tsuolbmajavri.Korhola.2002	68,41	22,05	526	10.1016/S0277-3791(02)00003-3

British Isles	Toskaljavri.Seppa.2002	69,20	21,47	704	10.1002/jqs.678
	LoughMeenachrinna.Taylor.2018	54,73	-8,68	286	10.1016/j.palaeo.2018.06.006
	TalkinTarn.Langdon.2004	54,92	-2,71	130	10.1023/B:JOPL.0000029433.85764.a5
	Lochnagar30.Dalton.2005	56,96	-3,23	788	10.1016/j.palaeo.2005.02.007
no specific region	BasadelaMoraLake.Tarrats.2018	42,54	0,33	1914	10.1177/0959683618788662
	LagoVerdarolo.Samartin.2011	44,36	10,12	1390	10.1038/NGEO2891
	Gemini.Samartin.2017	44,39	10,05	1349	10.1038/NGEO2891
	TauldintreBrazi.Toth.2015	45,40	22,90	1740	10.1177/0959683614565953
	TaulMuced.Diaconu.2017	47,57	24,55	1360	10.1016/j.palaeo.2017.05.007
	Hypkana.Hajkova.2016	48,91	22,16	820	10.1016/j.quascirev.2016.04.001
	Topptjonna.Paus.2011	62,38	9,67	1316	10.1016/j.quascirev.2011.04.010
	Berkut.Ilyashuk.2005	66,34	36,66	25	10.1191/0959683605hl865ra
	Kharinei.Jones.2011	67,36	62,75	108	10.1007/s10933-011-9528-7
	Sokli.Shala.2017	67,81	29,28	220	10.1177/0959683617708442

Code and data availability

955 The primary data, i.e. the model code for MPI-ESM, are freely available to the scientific community and can be accessed with a licence (https://mpimet.mpg.de/en/science/modeling-with-icon/code-availability, last access: 24 November 2021). The simulation (simulation identity: slo0021) will be published soon on the Earth System Grid.

The educational version of LPJ-GUESS is available for download (https://web.nateko.lu.se/lpj-guess/education/, last access: 06.12.2022) and a fully functional version is available for researchers from Department of Physical Geography and 960 Ecosystem Sciences at Lund University upon request.

The REVEALS-based vegetation estimates are stored in PANGAEA (Marquer et al. (2019): https://doi.pangaea.de/10.1594/PANGAEA.900966)

Most of the chironomid-based temperature reconstructions have been downloaded from the Temperature12k database available here: https://lipdverse.org/Temp12k/current_version/, by Kaufman et al. (2020).

965   In addition, secondary data and scripts that may be useful in reproducing the authors' work are archived by the Max Planck Institute for Meteorology and are accessible without any restrictions (https://hdl.handle.net/21.11116/0000-000D-4E03-9, MPG.PuRe, 2023). ~~The link will be published with the final version of the paper.~~

Author contribution

All authors planned the study and were involved in the analysis and discussion of the results. AD, AP, LM, and MJG wrote
970   the manuscript. AP, AD, LM and AS prepared the figures. All authors commented on, discussed, and edited the final manuscript.

Competing interests

AD is currently acting as a Guest Editor for the Climate of the Past Special Issue "Past vegetation dynamics and their role in past climate changes".

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