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1 **Title: Black Death land abandonment drove European diversity losses**

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24 **Data accessibility:** Pollen data were obtained from the Neotoma Paleoecological Database
25 (<http://www.neotomadb.org>) and its constituent databases, in particular the European Pollen
26 Database and the Alpine Pollen Database. All data are freely downloadable and cited in-text.
27 All code used in the analysis, along with input datasets, intermediate data products and
28 outputted model fits (etc.) are open access and available from the Dryad Digital
29 Repository: <https://doi.org/10.5061/dryad.z08kprrrr>.

30

31

1 **Author contributions:** Conceptualization: JDG, BF, JF, LG, NM, CDT; Methodology: JDG,
2 BF, LG, NM, CDT; Investigation: JDG, BF; Visualization: JDG; Funding acquisition: CDT;
3 Supervision: LG, NM, CDT; Writing – original draft: JDG; Writing – review & editing: JDG,
4 BF, JF, LG, NM, CDT.

5

6 **Abstract:**

7 The current prevailing perception is that human impacts on the biological realm have been
8 overwhelmingly negative. Here, we test this narrative by considering the consequences for
9 aspects of floristic diversity of the ‘Black Death era’ (1300–1400 CE), where one third of
10 Europe’s population died within half a decade. Based on evidence from 109 pollen records
11 spanning the Common Era, we find increasing floristic diversity from 0 CE to ~1300 CE as
12 human populations increased, followed by rapid and substantial diversity reductions during
13 the famine- and disease-driven human mortality events of the ‘Black Death era’. As human
14 populations recovered following the mortality shock, diversity also recovered. Strikingly, it
15 was landscapes characterised by cereal cultivation that generated both the overall Common
16 Era increases *and* the Black Death era declines in diversity. The highest diversity levels
17 were achieved in human-generated, mosaic landscapes, highlighting the integral role of
18 human action in biodiverse European landscapes.

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1 Main Text

2 Introduction

3 Mass appropriation of land for food production is widely considered to be the primary driver
4 of biodiversity decline over the last century (IPBES 2019; Krausmann *et al.* 2013; Newbold
5 *et al.* 2015). Yet, over multi millennial timescales, analyses of continental-scale fossil pollen
6 datasets (that contain information on past plant community composition and diversity)
7 suggest the opposite, with increasing diversity associated with intensifying human footprints
8 across multiple biomes and continents (Gordon *et al.* 2024). While not necessarily
9 incompatible, given the different intensities of farming operations during different time
10 periods, these two sets of observations lead to opposing expectations of the biodiversity
11 consequences of land 'abandonment'. Here, we test these alternatives by considering an
12 historical precedent - the 14th Century in Europe, the 'Black Death era' - a period that
13 encompassed the Great Famine (1315 – 1317 CE) as well as the Black Death (1347 – 1353
14 CE) itself. Previous work has used the pollen record to identify spatial variation in Black
15 Death mortality (Izdebski *et al.* 2022), but here we use the pollen record to ask a
16 fundamentally different question: what were the consequences of the decline of cultivated
17 lands and the subsequent return of cereal farming activities for European floristic diversity
18 over the Black Death era? The answer to this question may inform ongoing conservation
19 choices in Europe, specifically the likely longer-term consequences of 'traditional
20 management' versus 'rewilding' conservation strategies.

21

22 The 14th Century represents the best-known pre-industrial period of human depopulation
23 and rapid social change in Europe. Population growth in the 13th Century created
24 considerable pressure on what had been largely subsistence agricultural and pastoral
25 economies across Europe, including meeting the food demand from the growth of urban
26 centres (Astill & Langdon 1997). The Great Famine and Black Death substantially raised
27 mortality rates and generated major social, cultural and economic changes, though the

1 severity of these varied from place to place (Alfani & Murphy 2017; Benedictow 2004;
2 Campbell 2000; Fraser 2011; Jedwab *et al.* 2022). Population levels only began to recover
3 over the late 16th and 17th Century, a recovery paralleled by increasingly specialised
4 agricultural systems and greater market freedoms (Hallam 1988).

5
6 Two sets of scientific literature provide opposing predictions of how floristic diversity might
7 respond in areas that experienced 14th Century depopulation and abandonment of
8 cultivated lands. The first hypothesis is that biodiversity levels should increase following
9 agricultural abandonment, over approximately 10 to 75 years (with the transition from
10 croplands and other anthromes towards intermediate-aged and mature successional
11 vegetation) and then remain high until agriculture returns (Cramer *et al.* 2008; Hudson *et al.*
12 2014; Newbold *et al.* 2015). The second hypothesis predicts declines in floristic diversity in
13 response to depopulation, farmland abandonment and the loss of anthropogenic mosaic
14 landscapes, followed by a recovery in diversity in the following two centuries, alongside
15 human population and cultural recovery (Armstrong *et al.* 2021; Gordon *et al.* 2024; Lomba
16 *et al.* 2020; Niedrist *et al.* 2009; Prangel *et al.* 2023; Root-Bernstein & Svenning 2018).

17
18 We investigate these competing hypotheses using a dataset of 4,616 Common Era samples
19 from 109 standardised European pollen records. Adopting a 'natural experiment' approach,
20 we treat the abandonment of cultivated land associated with increased human mortality and
21 socio-cultural reorganisations of the 14th Century in Europe as a test of historical human
22 influences on biodiversity.

23

24 **Materials and methods**

25 Data processing and standardisation

26 We downloaded all available European pollen records from the Neotoma palaeoecology
27 database (Williams *et al.* 2018) using the `neotoma2` R package (Vidaña & Goring 2023) and

1 standardised these data following the recommendations from Flantua *et al.* (2023), which we
2 briefly outline below (see *Supporting Information: Extended Methods* for further details). We
3 filtered the total Neotoma download based on each pollen record's (i) pollen and
4 chronological control sampling intensity and (ii) inclusion in the Githumbi *et al.* (2022)
5 synthesis of European Holocene vegetation change, in order to model past vegetation cover
6 (discussed in following two sentences). For records that met these criteria, we modelled past
7 vegetation cover for each pollen sample using the REVEALS model (Sugita 2007), with the
8 requisite per-record input data from Githumbi *et al.* (2022). This provided proportional
9 estimates of past vegetation cover for the dominant taxa per sample, which we multiplied by
10 the original pollen counts [following Felde *et al.* (2016)] to help mitigate against the variation
11 in pollen source areas, relative pollen productivity and dispersal, depositional environments,
12 taphonomy, [etc., see Sugita (1994, 2007) and Sugita *et al.* (2010)] across the dataset,
13 which can strongly bias pollen diversity and composition metrics (e.g., Senn *et al.* 2022). We
14 then computed a new Bayesian age-depth model for each pollen record using **Bchron**
15 (Haslett & Parnell 2008), harmonised pollen-type names using an updated harmonisation
16 table (Birks *et al.* 2023), randomly resampled (without replacement) 150 pollen grains from
17 each REVEALS-adjusted pollen sample using the `rrarefy` function from **vegan** (Oksanen
18 *et al.* 2022) and paired each randomly resampled pollen assemblage with a random
19 realisation ('draw') from that record's age-depth model. We then excluded all samples with
20 assigned ages falling outside the Common Era (0 – 1850 CE). We repeated this resampling
21 and age allocation procedure 1,000 times to generate 1,000 Common Era pollen datasets –
22 where each sample is represented by a random age draw and resampled pollen assemblage
23 – thereby propagating uncertainties inherent to the pollen resampling and age-depth
24 modelling steps through our analyses.

25

26 Overall, this procedure resulted in the reduction of the initial Neotoma download down to 109
27 high temporal resolution European pollen records (each resampled 1,000 times), comprising

1 4,616 samples spanning the Common Era (see *Supporting Information: Extended Figures*
2 *and Tables, Fig. 1* for sampling intensities and spatial and temporal coverage of the pollen
3 dataset and *Supporting Information: Extended Methods* for detailed descriptions of each
4 step summarised here).

5

6 Pollen diversity and composition metrics

7 We used these standardised pollen data to evaluate the consequences of the ‘Black Death
8 era’ for floristic diversity by comparing the richness [the number of unique pollen types per
9 sample (Birks & Line 1992)], evenness [how the abundances of pollen grains are distributed
10 among pollen types (Pielou 1966)], and temporal turnover of pollen samples [the change in
11 the abundances and types between successive samples, measured using an interval-
12 adjusted Bray-Curtis index (Bray & Curtis 1957)], before, during and after the 14th Century.
13 Fossil pollen data contain information on past plant compositions, though the pollen –
14 vegetation relationship is non-linear (Davis 1963; Prentice & Webb III 1986). Despite this
15 non-linearity, positive relationships have been identified between pollen and plant diversity
16 across multiple regions, vegetation types and climates of Europe (e.g., Felde *et al.* 2016;
17 Matthias *et al.* 2015; Papadopoulou *et al.* 2022; Reitalu *et al.* 2019; Senn *et al.* 2022). To
18 estimate changes in the extent of cultivated land, we computed the REVEALS-outputted
19 “Agricultural land” plant functional type, which represents modelled counts of pollen grains
20 from cereal taxa (Githumbi *et al.* 2022; Wolf *et al.* 2008).

21

22 We modelled these Common Era diversity and vegetation composition datasets through time
23 using Generalised Additive Models (GAMs), fitted with discretized predictor variables (for
24 computational efficiency) and `bam` from `mgcv` (Li & Wood 2020; Wood 2011). We employed
25 a bootstrapping and posterior simulation procedure to handle the 1,000 resampled pollen
26 datasets (that included variation from the age-depth modelling and resampling) and to
27 incorporate statistical uncertainty adequately in our results. We approximated the first

1 derivative of each simulated temporal smooth using the method of central finite differences
2 and summarised these derivative values across draws, marking timepoints in bold where
3 >95% of the cohort of simulations were represented by exclusively positive or negative
4 values, and others in regular linewidth. We performed these analyses for all combinations of
5 pollen metric, spatial unit (per-vegetation zone or European-scale) and 14th Century cereal
6 trajectory (see below).

7

8 Comparison of landscapes with different agricultural trajectories

9 We compared the temporal patterns of pollen diversity and composition between landscapes
10 with different histories of agricultural activity over the Black Death era. To do this, we
11 decomposed the Common Era pollen diversity and composition data (Fig. 1A-D) into three
12 categories based on the 14th Century cereal farming trajectories of individual pollen records,
13 calculating each record's '14th Century cereal farming trajectory' as the change in average
14 (across samples per record) cereal counts between the period 1200 – 1300 CE and the
15 period 1400 – 1500 CE. We then modelled these categories through time using GAMs, as
16 above. The categories were (i) pollen records that declined in cereal representation, thus
17 indicating a contraction or cessation of arable farming (i.e., some degree of abandonment;
18 Figs. 1E-H, green lines), (ii) pollen records where cereal cultivation continued at the same
19 level or increased over the same period (Figs. 1E-H, orange lines), and (iii) pollen records
20 that showed no evidence of cereal cultivation over the 14th Century (zero cereal grains
21 present in samples over the 13th – 15th Centuries; Figs. 1E-H, purple lines).

22

23 Category (i), pollen catchments where cereal pollen declined over the Black Death era, could
24 in principle include landscapes where there was a complete 'cessation' of cereal farming
25 (i.e., records with cereal counts that declined to zero) and others where there was merely a
26 'decline' (i.e., records with cereal counts that declined to a non-zero value). However, we
27 cannot distinguish such landscapes with confidence due to sample error; given the relatively
28 low cereal counts (on average fewer than one grain per 150 grains in a rarefied resample,

1 Fig. 1D) and the potential for the long-distance transport of individual pollen grains. Hence,
2 we opted for a combined 'abandonment' category, where 'abandoned' refers to any
3 reduction in cereals, rather than necessarily the complete disappearance of cereal pollen
4 from pollen samples.

5

6 Assessing the drivers – human-generated mosaics

7 Finally, we investigated how changes in vegetation structure and landscape (pollen source
8 area) scale heterogeneity over the Black Death era related to these diversity changes (Fig.
9 2). European vegetative types can be (heuristically) characterised along a spectrum between
10 fully wooded and fully open (grassland/herbaceous, including open anthromes) ecosystems.
11 Based on expectations from the intermediate disturbance (Connell 1978; Grime 1973) and
12 habitat heterogeneity (MacArthur 1972) hypotheses, we expect landscapes characterised by
13 intermediate proportions of wooded and non-wooded elements to have the highest floristic
14 diversity. If this is correct, we expect divergent trajectories of diversity change over the Black
15 Death era depending on the 'starting condition' (wooded proportion) of each landscape's
16 floristic assemblage. To investigate this, we firstly modelled the relationship between
17 richness and the tree pollen percentage (which represents the coarse-level structure of
18 Europe's vegetation: wooded and non-wooded elements) for all pollen samples that fell
19 within the 13th Century (1200 – 1300 CE); i.e., prior to the Great Famine and Black Death.
20 We then computed the *predicted* change in richness over the 14th Century (using the model
21 fitted to the spatial data in Fig. 3A) that would be expected, given the (i) initial tree cover (in
22 the 13th Century) and (ii) change in tree cover at each pollen record location across the 14th
23 Century (i.e., between the 13th and 15th Centuries). We then plotted the *observed* change in
24 richness over the 14th Century against these predicted changes (Fig. 3B; see Fig. 2 for a
25 map of actual changes in each metric per pollen record over the 14th Century) to evaluate
26 how well the modelled relationship (Fig. 3A) predicted actual richness change over the Black
27 Death era. We performed these analyses at both whole-Europe (Fig. 3) and also vegetation

1 zone (Fig. 4D-I) scales, since pre-Black Death (13th Century) tree cover varied between
 2 biogeographic regions.

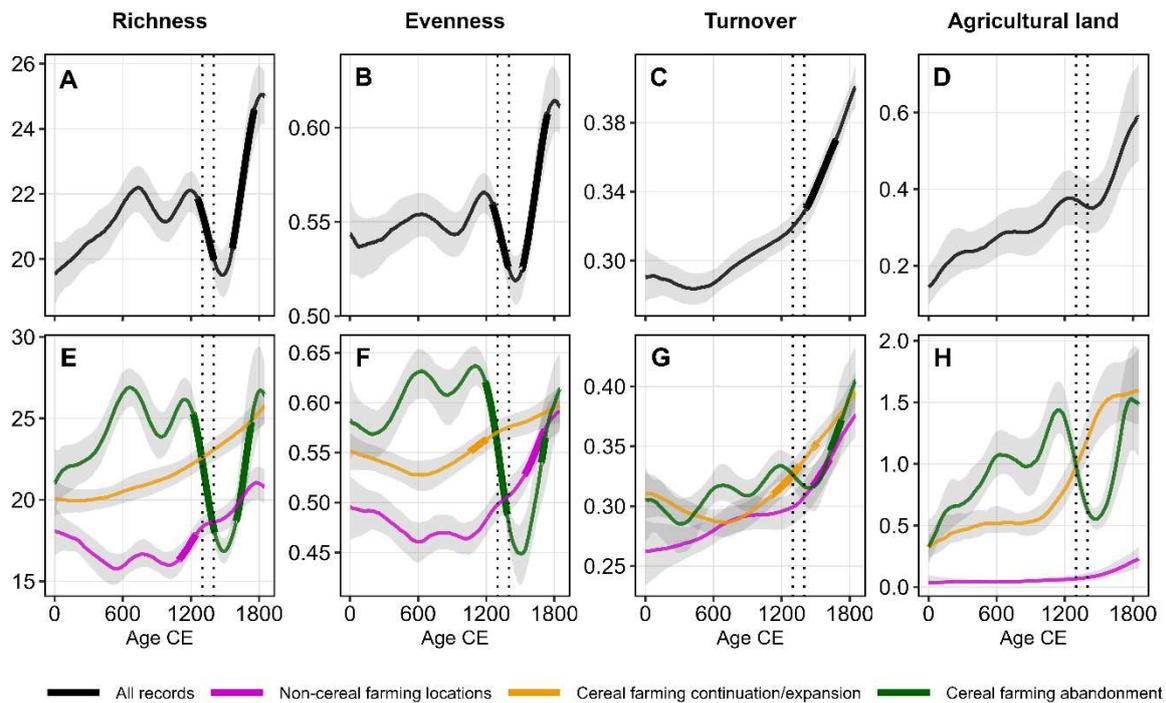
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5 Results

6 Temporal and spatial diversity patterns of diversity change

7 Overall, average pollen richness and evenness increased from the start of the Common Era
 8 up to ~700 CE, followed by a period of decline and then subsequent increase from ~1000
 9 CE up to 1300 CE (Figs. 1A & B). Following the onset of the 14th Century, average richness
 10 and evenness rapidly declined, with both metrics falling to the approximate level they had
 11 been 1,000 years earlier. After ~150 years of steep decline, both richness and evenness
 12 increased again, returning to their late 13th Century peak by ~1650 CE (Figs. 1A & B) and
 13 increasing to the final time point considered here (1850 CE). Floristic turnover remained
 14 relatively stable, or slightly reduced, over the first ~500 years of the Common Era but
 15 accelerated after ~600 CE up to the 1850 CE (Fig. 1C).



1 **Fig. 1. Common Era diversity metrics.** Generalised additive models fit to **(A, E)** richness
2 (number of unique pollen types), **(B, F)** evenness (Pielou Index), **(C, G)** turnover (Bray-Curtis
3 dissimilarity through time) and **(D, H)** Agricultural land (REVEALS-modelled cereal counts
4 per 150 grains) as a smooth function of time. Panels **A-D** show results across all pollen
5 records (black); panels **E-H** show results split into areas where cereal farming declined
6 (green, n = 32) or remained stable/increased (orange, n = 62) over the 14th Century, with
7 non-cereal farming locations shown in purple (n = 15). Central lines represent the average fit
8 across the 1,000 resamples, with sections of thicker lines indicating periods over which
9 >95% of the smooths exhibit positive or negative slopes. Slope percentiles are not shown for
10 Agricultural land because the direction of change is predetermined by the subsetting of
11 records by cereal trajectory. Grey intervals represent interquartile range across fits to each
12 of the 1,000 resamples. Vertical black dotted lines show the bounds of the 14th Century.
13 Sample sizes per coloured line in **(E-H)** are computed by calculating the mean change in
14 cereals over the 14th Century across the 1,000 resamples, then identifying the category of
15 change. Note though that individual resamples (of the 1,000) per pollen record may not all
16 show the same cereal trajectory (given the random variation in pollen assemblage
17 compositions across the 1,000 arising from our resampling procedure). Total pollen record N
18 = 109.

19

20 Agricultural land, estimated from cereal counts, also increased overall from 0 CE up to
21 ~1250 - 1300 CE, with a fluctuation between ~700 and 1000 CE. Agricultural land extent
22 then declined to ~1450 CE and gradually increased thereafter up to 1850 CE (Fig. 1D).

23 Hence, richness and evenness changes follow those of agricultural cereal trends

24 (*Supporting Information: Extended Figures and Tables, Extended Fig. 2*). These two

25 diversity metrics remained lower than the pre-1300 CE levels for ~300 years, a duration

26 mirroring that of human population and of societal and agricultural recovery in Europe after

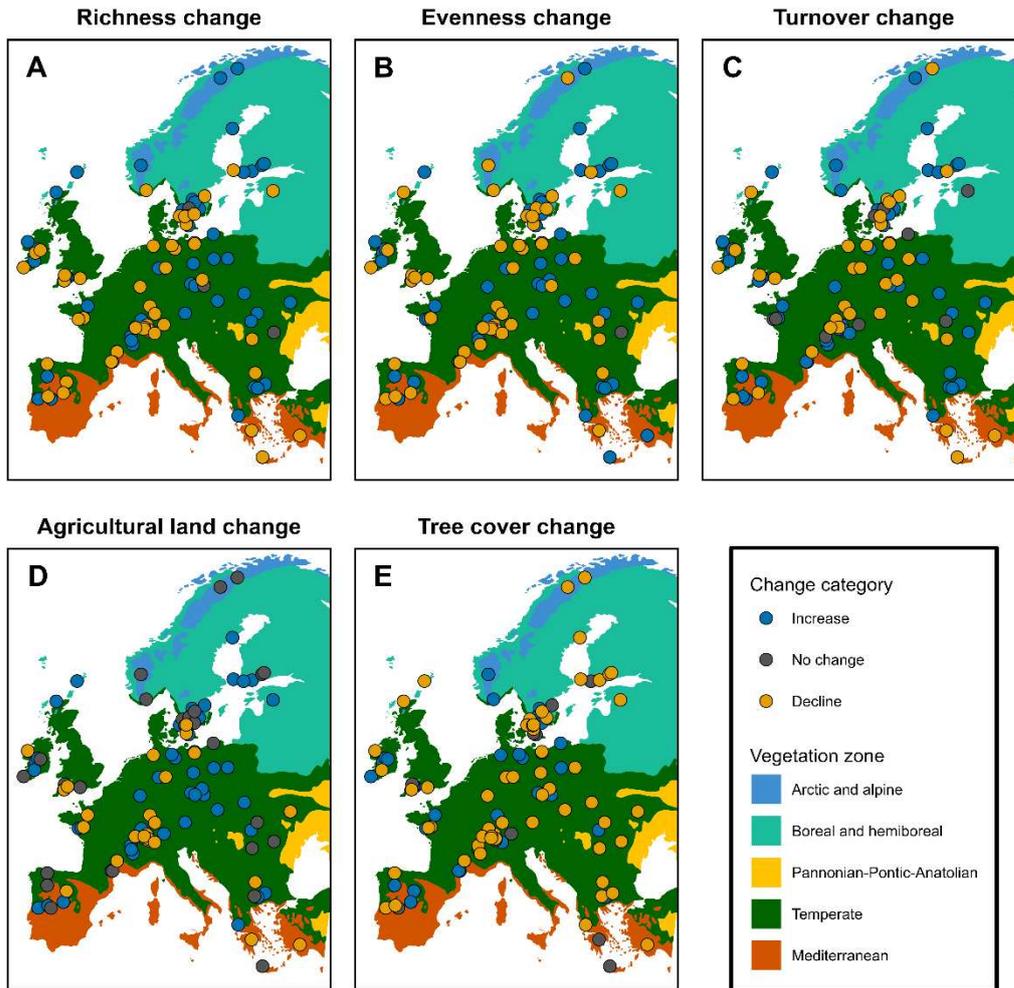
27 the crises of the 14th Century.

28

1 Despite these clear average temporal patterns (Fig. 1), trajectories of change at individual
2 sites over the Black Death era (Fig. 2; calculated as the change per site in each average
3 metric between 1200 – 1300 CE and 1400 – 1500 CE) were spatially heterogeneous. For
4 richness, evenness and turnover, there were some sites that increased and others that
5 decreased over the full range of European latitudes and longitudes (Fig. 2A-C). Change in
6 the extent of agricultural land (Fig. 2D) and tree cover change (Fig. 2E) were also spatially
7 heterogeneous. Different directions of per-site changes in each metric were observed within
8 each broad vegetation type (Fig. 2), suggesting landscape-scale responses, but there were
9 nonetheless different average biodiversity responses that reflected differences in average
10 levels of pre-Black Death forest cover in each vegetation region (see below, Fig. 4).

11

12



1

2 **Fig. 2. Change in diversity metrics, turnover and agricultural land area over the Black**3 **Death era (spanning the 14th Century).** Points represent the change in **(A)** richness, **(B)**4 evenness, **(C)** turnover, **(D)** agricultural land (REVEALS-modelled cereal counts) and **(E)**

5 tree cover (percentage) at individual pollen sites between two time windows: Time window 1

6 = 1200 – 1300 CE and time window 2 = 1300 – 1400 CE (averages of samples per site, for

7 the two centuries; see *Methods*). Yellow points represent declines in each metric, blue points

8 represent increases and grey points represent no change (plus or minus five percent).

9 Coloured polygons represent Europe's primary vegetation zones from Lang *et al.* (2023).

10 Total pollen record N = 109.

11

1 The overall downturn in European cereal pollen was relatively modest (Fig. 1D) because the
2 effects of the Great Famine and Black Death were not spatially uniform across Europe, with
3 the extent and nature of agricultural practices and its impacts varying both within and
4 between regions (Fig. 2). Temporal diversity patterns from pollen records that either
5 remained unchanged or increased in cereal representation over the 14th Century show
6 diversity gains from the 7th Century up to 1850 CE (Figs. 1E-G, orange lines). However,
7 those landscapes that experienced declines in cereal cultivation strongly declined in richness
8 and evenness from ~1300 CE, reaching minima ~1450 CE, but increasing thereafter (Figs.
9 1E-F, green lines). Average richness and evenness values were substantially lower for non-
10 cereal farming locations than for the other two categories over the period 0 – 1300 CE (Figs.
11 1E-F, purple lines), with richness and evenness values at abandoned sites declining to less
12 than or approximately the same values as those non-cereal farming sites over the Black
13 Death era (compare Figs. 1E-F green and purple lines). Differences between the three
14 categories in the magnitude of turnover changes were much smaller (Fig. 1G). Plotting
15 cereals by these same categories (Fig. 1H) shows the similarity of temporal trends for cereal
16 pollen, richness and evenness for those sets of pollen records that reveal declines in cereal
17 farming activity across the 14th Century.

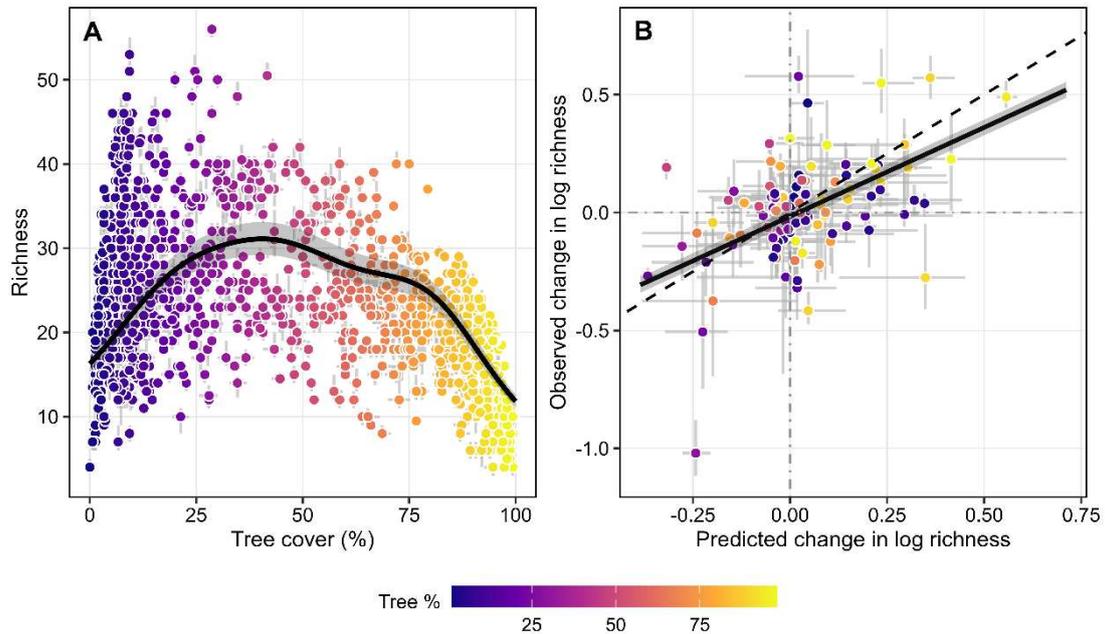
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19 Together, these data show that the diversity declines observable in the full diversity datasets
20 (Figs. 1A-C) were driven by declining diversity metrics in landscapes where cereal cultivation
21 had been at relatively high levels up to ~1300, but where cereal cultivation declined or was
22 abandoned in the 14th Century (Figs. 1E-G). It is worth noting that pollen counts from
23 cereals represent a very small proportion of the total count (between zero and two grains per
24 150 on average, Fig. 1D), so cereal cultivation did not dominate Europe's landscapes over
25 the period considered here - they indicate the presence of human cereal farming and, by
26 implication, other human activities (such as rearing livestock, erecting buildings, establishing
27 boundaries, paths and tracks, and managing woodlands) within multi-use landscapes.

28

1 Changes in landscape heterogeneity

2 Our analyses reveal a 'hump-shaped' relationship between richness and tree cover, with
3 expected average richness maximised at approximately 40% tree cover (Fig. 3A; and also
4 maximised for intermediate levels of cereal pollen, *Supporting Information: Extended Figures*
5 *and Tables, Extended Fig. 3*). This is consistent with the heterogeneity hypothesis
6 (MacArthur 1972), whereby very open (here, <30% tree cover) landscapes are expected to
7 increase in diversity with the expansion of tree cover, while tree cover expansion beyond
8 ~50% is expected to erode diversity. Plotting the observed change in richness over the 14th
9 Century against changes predicted from the change in tree cover over the same period (see
10 Fig. 2 for a map of actual changes in each metric per pollen record over the 14th Century)
11 reveals a positive relationship between the model-predicted and observed richness change
12 values, verifying that the model (in Fig. 3A) is capturing the general trend in changes to
13 richness predicted by changes in tree cover across the 14th Century (Fig. 3B). This shows
14 that landscapes with vegetation that moved towards intermediate tree cover over the Black
15 Death era resulted in richness gains, whereas those landscapes that moved towards the
16 extremes resulted in richness losses. Given that humans were major drivers of changes to
17 European woodland extent from the Neolithic onwards (Roberts *et al.* 2018), these results
18 support our overall interpretation that landscapes with intermediate levels of human
19 landscape-scale modification represent higher diversity floristic communities.



1

2 **Fig. 3. Tree representation and richness relationships. (A)** Richness as a nonlinear3 function of percentage tree cover (1200 – 1300 CE), see *Supporting Information: Extended*4 *Figures and Tables, Extended Table 1* for model summary (N unique pollen samples =5 1,449). **(B)** Predicted versus observed change in log richness across the 14th Century, see6 *Supporting Information: Extended Figures and Tables, Extended Table 1* for model summary7 (N pollen records = 109). The values along the x-axis in panel **B** are computed by calculating

8 the expected richness change given the change in tree cover estimated from the fitted non-

9 linear function shown in panel **A**. For example, if a pollen record increased in tree cover from

10 an initial 5% in the period 1200 – 1300 CE to 10% tree cover in the period 1400 – 1500 CE,

11 there would be an expected richness change of +5. The values along the y-axis in panel **B**

12 are computed by calculating the difference in richness (from the empirical pollen data)

13 between the period 1200 – 1300 CE and 1400 – 1500 CE. In both panels, the central points

14 represent the median and the crosshairs represent the variation (25th and 75th percentiles)

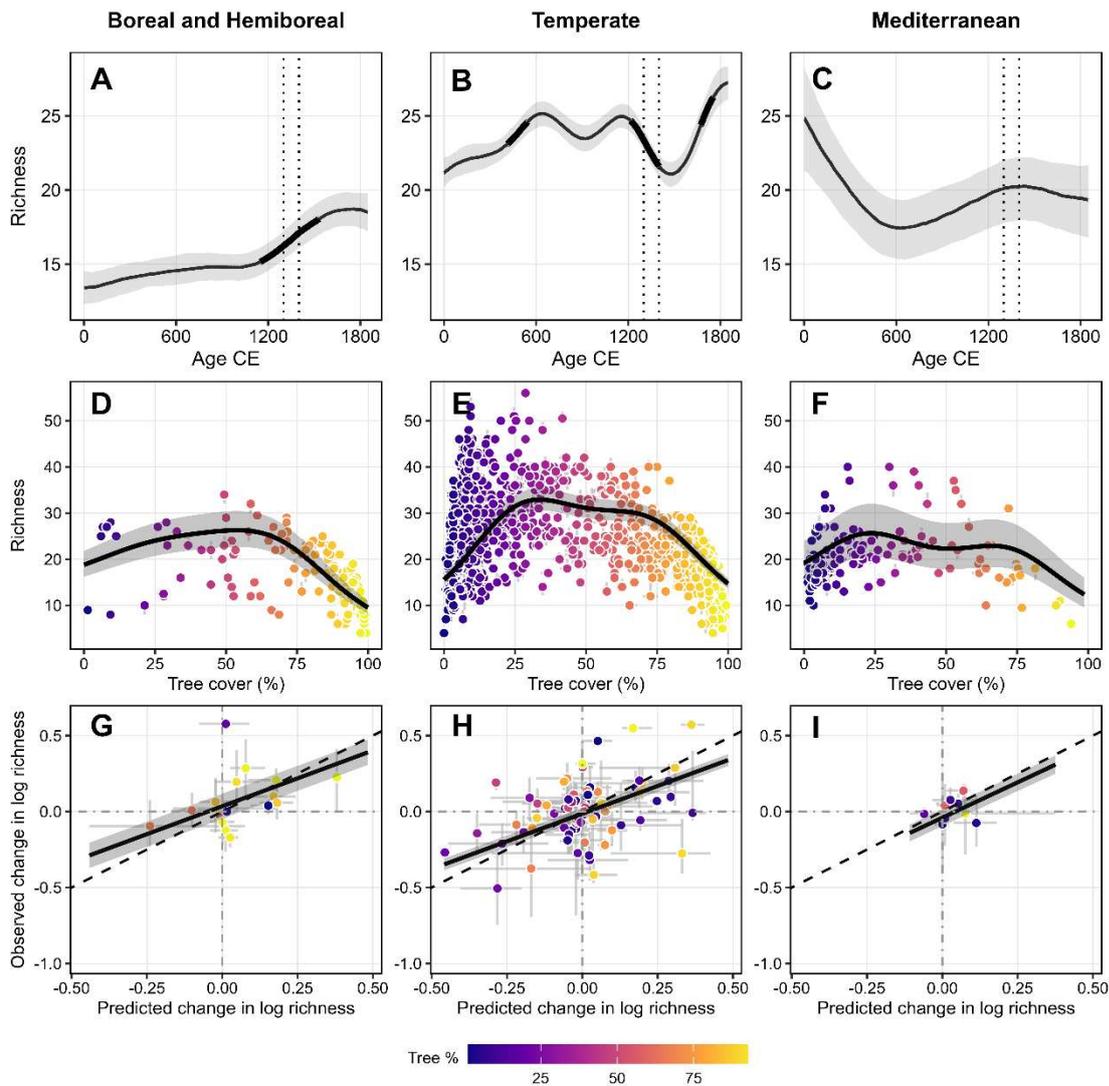
15 across the 1,000 resamples in both (change in) richness and percentage tree cover. Points

16 are coloured by their tree percent in the 13th Century (shown on the x-axis in panel **A**).17 Model summaries are presented in *Supporting Information, Appendix 1*.

18

1 Different vegetation zones exhibited different average trajectories of Common Era temporal
 2 richness patterns (Figs. 4A-C), but the underlying processes appear to be the same: (i)
 3 locations characterised by intermediate tree cover were associated with maximum richness
 4 (Fig. 4D-F) and (ii) change in richness predicted per landscape from the change in tree cover
 5 between the periods 1200 – 1300 CE and 1400 – 1500 CE was strongly related to observed
 6 changes in richness at individual sites (Fig. 4G-I).

7



8

9 **Fig. 4. Per-vegetation zone (A-C) temporal richness patterns and (D-I) tree**

10 **representation and richness relationships. (A-C) Generalised additive models fit to**

11 **richness as a smooth function of time. Central lines represent the average fit across the**

1 fitted distribution (see *Methods*), with sections of thicker lines indicating periods over which
2 >95% of the draws exhibit positive or negative slopes. Grey intervals represent the
3 interquartile range across draws. Vertical black dotted lines show the 14th Century. **(D-F)**
4 Richness as a nonlinear function of percentage tree cover (1200 – 1300 CE) for each
5 vegetation zone, see *Supporting Information: Extended Figures and Tables, Extended Table*
6 *2* for model summaries. **(G-I)** Predicted change in log richness versus observed change in
7 log richness across the 14th Century; see *Supporting Information: Extended Figures and*
8 *Tables, Extended Table 3* for model summaries. In all panels, the central points represent
9 the median and the crosshairs represent the variation (25th and 75th percentiles) across the
10 1,000 resamples. Points in panels **D-I** are coloured by their tree percent in the 13th Century
11 (shown on the x-axes in panels **D-F**).

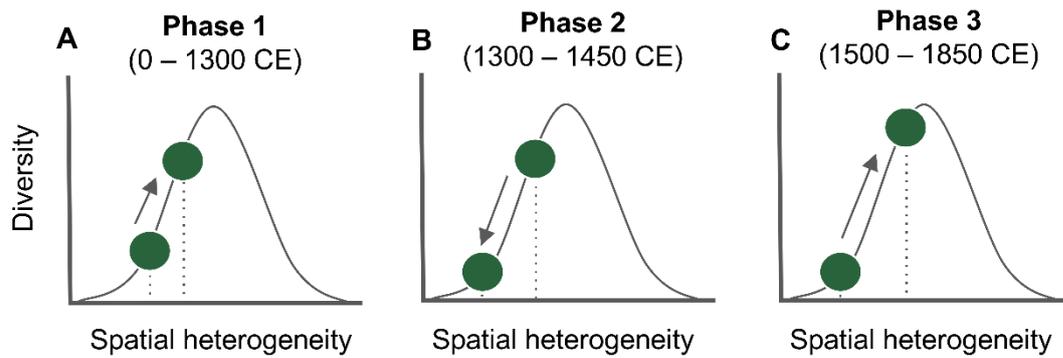
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13 **Discussion**

14 Using 14th Century famine and disease in Europe as a test case, we find that abandonment
15 or decreases in arable farming (with associated farming and rural management activities) led
16 to declines in biodiversity metrics over the following one to two centuries. For those pollen
17 records with evidence of cereal decline (Fig. 1H, green line; Fig 2D, yellow points), we
18 identify three phases of diversity change over the Common Era (up to 1850 CE), which we
19 suggest are directly linked to changing human landscape management practices (at the
20 scale of relevant pollen source areas, ~landscapes) and their impacts on the heterogeneity
21 of habitats per unit area within a landscape (Fig. 5).

22

23



1

2 **Fig. 5. Schematic representation of changes to floristic diversity across the Common**3 **Era in European landscapes that experienced cereal abandonment during the 14th**4 **century.** Green circles represent the average diversity changes in relation to changes in

5 spatial heterogeneity of pollen source areas surrounding European pollen records

6 (~landscapes), for three time periods. See main text for additional interpretation.

7

8 Phase one (0 – 1300 CE) represents an expansion of arable practices (Fig. 1D & H) as

9 human populations grew, changing those landscapes in ways that generated mosaics of

10 different habitats and increasingly diverse floristic assemblages (Fig. 1, Fig. 5A).

11 The period 0 – 900 CE includes the expansion, maximum extent and later the retreat (during

12 the 5th century) of the Western Roman empire from southern Europe and beyond. Roman

13 agricultural systems were highly market oriented and their expansion over the period 0 – 400

14 CE (Fig. 1D & H) was driven by an increasing nutritional demand from the Empire's growing

15 urban populations (administrators, crafts people, etc.) and its substantial military (Allen &

16 Lodwick 2017). The increased tax burden from the Roman administration necessitated the

17 expansion of provincial grain production, with an expanded road network and investment in

18 grain drying and storage facilities facilitating its long-distance transport (Van der Veen 2016).

19 In the Mediterranean zone, production focussed on olives and grapes, with cereal and other

20 crops frequently grown in a diverse matrix. Outside of the Roman frontiers (i.e., much of

21 northern central and north eastern Europe), populations relied heavily upon barley and later

22 rye, often cultivated in mixed systems with a strong emphasis on raising cattle (Behre 1992;

1 Grabowski 2013; Lange 1975). Despite these regional differences, European arable
2 production increased over the first 400 years of the Common Era, a trend that continued into
3 the Early Middle Ages (from 500 CE onwards; Fig. 1). In England, for example, the post
4 Roman period saw little woodland regeneration and settled areas remained open (Rippon *et*
5 *al.* 2015). However, in some areas there were significant changes when agriculture shifted
6 from small-scale relatively intensive regimes to extensive, low-input, larger-scale regimes by
7 800 – 900 CE (Hamerow 2022). The period 680 – 830 CE saw several new features
8 appearing in Europe's landscape such as ditched enclosures to corral livestock and
9 centralised crop processing and storage, suggesting more extensive arable cultivation and
10 grain surpluses, yet still within a very mixed agrarian landscape (Hamerow 2022).

11

12 Our data therefore do not support a 'collapse' of provincial rural agriculture after Roman
13 withdrawal in the 5th Century (Faulkner 2000; McCormick 2002; Ward-Perkins 2005),
14 instead it supports the view of a restructuring of post-Roman markets, economies and elites
15 (Wickham 2009) in unison with continued arable production. Thus, diversity measures
16 continued to increase in association with expanding arable production over the Early Middle
17 Ages, up to ~700 CE (Fig. 1A & E).

18

19 The period 700 – 900 CE had roughly stable (perhaps slightly reducing) arable production
20 and moderately reducing diversity values. Yet, from ~900 CE onwards, European
21 populations grew and the total area under cultivation expanded once again, particularly in
22 the more densely populated areas of north-western Europe (Duby 1962; Hoffmann 2014;
23 Mitterauer 2010). Enabled by innovations such as the heavy mouldboard plough, the use of
24 horses as draught animals, and three-year crop rotations with extensive use of fallowing,
25 arable cultivation was extended onto previously unploughed woodland, wastelands, pastures
26 and wetlands. New crops such as oats and rye were also introduced to exploit poor soils,
27 particularly in upland regions (Litt *et al.* 2009). Open fields with intermingled strips and some
28 degree of collective management are documented in many regions from 1000 CE onwards,

1 such as in the midlands of England (Leturcq 2007; Schroeder 2022; Thoen 2018).
2 Elsewhere, for example across Scandinavia, a single 'infield' of small plots of fields were
3 surrounded by large open areas of meadows, and pastures at some distance from the
4 infield, that could be brought into temporary arable cultivation when needed, or that could be
5 intermingled with other enclosed land, forming a patchwork of cropping regimes (Nesset &
6 Hjelle 2022; Thoen 2018).

7
8 During Phase one, Europe's expanding and diverse mosaics of cultivated lands were
9 typically separated by woods, heath, or uncultivated plots, and often enclosed by hedgerows
10 or trees (Verhulst 2002). Europe's agro-ecosystems thus represented a complex lattice of
11 multiple agricultural practices operating within remnants of 'natural' (e.g., forests, bogs) and
12 'semi-natural' (e.g., grasslands, heathlands) areas in many European regions, and diversity
13 metrics were correspondingly high (Figs. 1 & 4). In addition to the different elements of the
14 mosaic, edge effects and ecotones would have contributed to the overall raised floristic
15 diversity at this time.

16
17 Phase two (1300 – 1450 CE) represents a sharp decline in diversity during and immediately
18 following the Black Death era (Fig. 5B), with the strongest impacts within the temperate
19 zone, particularly north western Europe (Fig. 2). During this period, the decline in human
20 populations and resulting social and cultural reorganization led to the reduction or cessation
21 of arable farming in many regions across Europe (Dodds 2008). Relatively recent
22 settlements that had been established in more marginal locations as the population grew,
23 and which required greater labour and resource inputs to sustain, were more vulnerable to
24 abandonment (Green & Mair 2025), whereas larger more established settlements with
25 preferential trade routes or resources might have experienced consolidation or population
26 growth, as surviving communities migrated. The reduction or full withdrawal of cultivation
27 reduced spatial heterogeneity within many landscapes across Europe, thereby driving
28 floristic diversity declines in these areas (Figs. 1 & 2D).

1 However, biodiversity responses to Black Death land-use changes varied significantly by
2 location (Fig. 2) (Yeloff & Van Geel 2007) due to differences in the 'starting condition' of
3 vegetation compositions (i.e., amount of tree cover) in the century before the Black Death.
4 Our analysis – based on records from across multiple vegetation zones and latitudes –
5 reveals a non-linear relationship where diversity peaked at intermediate levels of tree cover
6 (~40%, Figs. 3 & 4D-I). Consequently, diversity declined in landscapes that shifted towards
7 extremes (either heavily wooded or very open) but increased in those landscapes that
8 shifted towards intermediate tree cover states over the 14th century. Thus, reduction or
9 cessation of arable production did not universally reduce biodiversity. In previously very
10 open landscapes (<30% tree cover), woodland regeneration increased habitat heterogeneity
11 and drove richness gains (e.g., records in Northern Germany, Northern France, Western
12 Spain; increasing richness and tree cover, blue points in Figs. 2A & E). Conversely, in
13 regions with higher initial tree cover (>50%), diversity changes were positively linked to
14 openness. A recent study based on six pollen records from southern Germany finds that
15 reduced arable production was associated with floristic diversity declines over this period
16 (decreasing cereals, yellow points; Fig. 2D) (Spitzig *et al.* 2025), whilst in other areas such
17 as Bohemia, agricultural expansion over the 14th Century and the transition from a forest-
18 based economy to an agricultural one (Myśliwski 2011) resulted in diversity gains (increasing
19 cereals, blue points; Fig. 2D). Our whole-continent analyses reveal that Black Death era
20 diversity responses were determined by whether land-use changes moved a landscape
21 closer to, or further from, an optimal intermediate state of tree cover, rather than by a linear
22 relationship with openness.

23 Finally, Phase three (1450 – 1850 CE) represents a gradual agrarian recovery in the
24 aftermath of the Black Death as economies across Europe regenerated, and population
25 levels gradually responded. Increasing market engagement, including long distance trade,
26 resulted in a specialised and diverse emerging European agricultural economy from the 16th
27 Century e.g. (Overton 1996), the result of which was to drive diversity back up, increasing up

1 to the end of the study period (1850 CE; Fig. 5D). Meanwhile, the changing practices and
2 pressures of the period led to continuing growth of arable land in other landscapes (i.e.,
3 increased cereal pollen frequencies) from before 1100 CE through to ~1500 CE, and with it
4 increased richness, evenness and community turnover (orange lines in Figs. 1E-H). For
5 example, the well-established and persistent mixed farming practices associated with
6 subsistence farming in Central Europe and parts of southern Scandinavia. In non-arable
7 landscapes (i.e., those with zero cereal pollen in samples 1200 – 1500 CE), richness,
8 evenness and turnover also increased, especially after 1500 CE, most likely associated with
9 increased pastoral farming and woodland management in ‘marginal’ landscapes (purple
10 lines in Figs. 1E-H).

11

12 Overall, these results support our hypothesis that increases in human population densities
13 and accelerating farming-associated landscape modification increased Europe’s floristic
14 diversity from the early-mid Holocene onwards, up to the final date considered here, 1850
15 CE (Gordon *et al.* 2024), a trend that was temporarily reversed in some areas by
16 depopulation, land abandonment and the economic and social shifts that followed the Great
17 Famine and Black Death. Diversity declines were associated with locations where cereal
18 representation was reduced or disappeared over the 14th Century (Fig. 1D-H). However, the
19 precise course of events depended on the prior environmental conditions of the landscapes
20 under consideration, i.e., the starting woodland cover (Figs. 3 & 4D-I). The different ‘starting
21 states’ may also help explain discrepancies between the conclusions of biodiversity
22 responses in the palaeoecological versus more recent record (e.g. these results cf. Newbold
23 *et al.*, 2015). Over the full period considered here – and the three types of landscapes that
24 were characterised by different arable farming histories – richness and evenness tracked
25 agriculture and, through it, human development, culture, and population growth, collapse,
26 and recovery. This reveals positive effects of traditional rural activities and human-created
27 mosaics of habitats on biological diversity. In turn, these human-generated (novel), diverse

1 systems supported novel and varied assemblages of other groups of taxa, such as farmland
2 birds (Carroll et al. 2023).

3

4 Remnants of these diverse, human-generated landscapes can still be found, with
5 traditionally-managed locations and regions that escaped intensification in the late 18th to
6 20th Century representing many of the most biodiverse present-day landscapes across
7 Europe (regions categorised as 'High Nature Value' farmlands by the European Environment
8 Agency (EEA 2004). These landscapes typically represent agro-ecological systems
9 characterised by low intensity grazing and mixed farming methods, and include *inter alia* the
10 Iberian *dehesas* and *montados* (low intensity agro-forestry systems), Alpine pastures, Baltic
11 wood pastures, the Hungarian *Tanya* (small-scale mixed farming systems) and the Nordic
12 Alvar pastures (semi-natural grasslands) (Lomba *et al.* 2020). Thus, the legacies of the low
13 intensity, mixed farming systems of Europe's Middle Ages remain present today across the
14 continent. These landscapes have dual cultural and biodiversity value but are today
15 considered at risk of abandonment, with associated loss of heterogeneity and biodiversity.
16 For example, land abandonment and loss of extensive grazing systems has led to
17 encroachment of shrubs and increasing risk of wildfires (Feurdean et al. 2017; Shakesby
18 2011), both of which may erode aspects of biodiversity. This is increasingly recognised by
19 governments and others who offer subsidies to farmers to support the persistence of
20 traditional, or equivalent, management practices such as the reintroduction of extensive
21 grazing to restore heterogeneity and biodiversity, and to reduce fire risks (Rouet-Leduc *et al.*
22 2021; Streifeneder *et al.* 2022).

23

24 In summary, we used the abandonment of cultivated land associated with increased human
25 mortality and socio-cultural reorganisations of the 14th Century in Europe as a test of
26 historical human influences on biodiversity. Our work suggests that human-associated
27 activities underpinned the major European floristic diversity gains that occurred over the last
28 12,000 years, the legacies of which are highly valued today across the continent as

1 biodiversity hotspots and culturally-valued landscapes. These findings underline the
2 interconnections of people and biodiversity and support the reinstatement of certain types of
3 'traditional' agriculture to support biodiversity, an approach that might be particularly suitable
4 in regions where other measures, such as strict protection and landscape-scale disturbance
5 using large animals ('rewilding'), are not feasible or desirable.

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1 **Supporting Information**

2

3 **Extended Methods**

4

5 **Data availability**

6 All pollen and chronological data used in this study are open access and obtained from the
7 Neotoma Paleoecological Database (Williams *et al.* 2018) (hereafter 'Neotoma', accessed
8 July 2024) and its constituent databases, in particular the European Pollen Database and
9 the Alpine Pollen Database (Fyfe *et al.* 2009). *Supporting Information: Extended Figures and*
10 *Tables, Extended Table 4* details the individual pollen records, and their locations, included
11 in these analyses. Data used for the REVEALS steps can be found in Githumbi *et al.* (2022).

12

13 **Code availability**

14 We performed all data manipulations and analyses in this study in **R** (R Core Team 2022)
15 and all 'packages' mentioned throughout are software extensions to **R**. Analysis code is
16 available at: <https://doi.org/10.5061/dryad.z08kprrrr>.

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1 **Pollen data processing and standardisation**

2

3 Data filter

4 We applied criteria relating to each pollen record's chronological and pollen sampling
5 intensities to the total dataset downloaded from Neotoma (data downloaded July 2024). We
6 included pollen records that contain two or more chronological control points (usually
7 radiocarbon dates) from within the Common Era, each separated by no more than 1,000
8 years, and three or more Common Era pollen samples. Each of these pollen samples was
9 required to contain a minimum number of terrestrial pollen grains (excluding spores), the
10 details of which are discussed below (see *Pollen resampling*). If a specific section of any
11 given pollen record conformed with these criteria, but the record as whole did not, we
12 included that conforming section. If multiple sections of any given pollen record conformed
13 with these inclusion criteria, but the record as a whole did not, we included the section with
14 the most chronological control points (that also must satisfy the pollen sampling criteria).
15 This ensured that each pollen record was included only once in our analyses.

16

17 Age-depth modelling

18 We constructed updated Bayesian age-depth models for all records that met our inclusion
19 criteria using the package **Bchron** (Haslett & Parnell 2008). Given that all records included
20 in these analyses are located in the northern hemisphere, we calibrated all radiocarbon
21 dates using the IntCal20 calibration curve (Reimer *et al.* 2020). We ran each age-depth
22 model for 40,000 iterations, with a thin value of 40, discarding the first 10,000 iterations. We
23 did not allow extrapolated age estimates beyond 500 years from the outermost radiocarbon
24 dates. We drew 1,000 age estimates per depth from each pollen record's posterior age
25 distribution, and carried these forwards to the resampling stage. This sampling approach
26 incorporates the uncertainty associated with age-depth modelling procedures and ensures

1 against spuriously confident age estimates (as would be achieved from simply taking the
2 median age estimate per depth).

3

4 Modelling past vegetation covers from pollen counts

5 Fossil pollen data contain information relating to past vegetation communities. Despite this,
6 pollen-vegetation relationships are non-linear and are affected by pollen source area (the
7 relevant geographical area over which pollen is contributed to a record), relative differences
8 in pollen production and dispersal between pollen producing plants, and taphonomy (i.e.,
9 between-pollen type and between-archive differences in pollen preservation in
10 environmental archives) (Prentice 1985; Seppä & Bennett 2003; Sugita *et al.* 2010). Left
11 unaccounted for, these factors have the capacity to strongly bias pollen diversity measures.
12 For example, if a sample is dominated by a numerically dominant pollen type (e.g., from a
13 high pollen producing plant or a well-dispersed pollen type) then the likelihood of sampling a
14 pollen grain that does not belong to that pollen type is reduced, artificially depressing the
15 apparent diversity of that pollen assemblage (Senn *et al.* 2022).

16

17 To overcome these biases (as far as is possible), we modelled pollen counts using the
18 “Regional Estimates of VEgetation Abundance from Large Sites” (REVEALS) model (Sugita
19 2007), using the REVEALS model code from Abraham *et al.* (2014). The REVEALS model is
20 a mechanistic model that takes pollen count data and a number of pollen-type and record-
21 specific abiotic variables [which we obtained from Githumbi *et al.* (2022)] as input data, to
22 provide proportional estimates (per pollen sample) of past vegetation covers [see Githumbi
23 *et al.* (2022) for further detail of method] per pollen sample. Of the total set of pollen records
24 available on Neotoma, 695 of them had the necessary data collected by Githumbi *et al.*
25 (2022) to run the REVEALS model. Thus, for each of these 695 pollen records that also
26 satisfied our data inclusion criteria (a total of 109 records), we ran the REVEALS model
27 across all pollen samples in each record in turn. We then multiplied the original per-sample
28 pollen counts by the outputted REVEALS proportional past vegetation estimates to convert

1 the REVEALS estimates back to pollen counts, following Felde *et al.* (2016). The REVEALS
2 model provides past vegetation estimates for the most numerically abundant pollen types in
3 fossil pollen spectra and we added the adjusted counts for these taxa back to the original
4 dataset, carrying them forwards to the next stage of our analyses.

5

6 Pollen harmonisation

7 We harmonised the names of all pollen types present across our Common Era datasets
8 using an updated European pollen harmonisation table (Birks *et al.* 2023) and summed the
9 REVEALS-transformed counts by these names per sample.

10

11 Pollen resampling

12 These analyses include pollen data originating from 109 individual studies, the sum of which
13 represent a wide range of methodological choices relating to pollen sampling and
14 quantification. Thus, given the heterogeneity in pollen sampling intensities between pollen
15 records, and the bias this would introduce to any computed diversity metrics if left
16 unaccounted for (samples with more pollen grains counted are likely – on average – to have
17 higher diversity estimates), we undertook a pollen resampling procedure to standardise
18 counts across samples and records, resampling 150 grains from each sample using the
19 `rrarefy` function from **vegan** (Oksanen *et al.* 2022).

20

21 REVEALS-transformed pollen counts represent original pollen sums multiplied by estimates
22 of proportional past vegetation cover. These adjusted samples thus represent a dataset of
23 pollen samples with total pollen sums often multiple factors lower than the original (i.e., non-
24 REVEALS-transformed) counts. This is highlighted by *Supporting Information: Extended*
25 *Figures and Tables, Extended Fig. 1F*, which shows the distribution of total terrestrial pollen
26 sums across the 4,616 included pollen samples *before* they were adjusted by the vegetation
27 estimates from REVEALS and underlines the high sampling intensity of the included

1 samples. To prevent the majority of samples (and therefore sites, see *Data filter*) being
2 excluded, we chose 150 grains as the resample value, approximately in line with previous
3 studies using a similar REVEALS methodology (e.g., Felde *et al.* 2016).

4

5 We repeated this resampling procedure 1,000 times. For each of these 1,000 random
6 resamples, we added a random realisation from the posterior distribution of the relevant age-
7 depth model to each sample (which relates to a level [depth] through the pollen sequence).

8 We carried these 1,000 datasets forwards to the next stage of our analyses, thereby
9 including the uncertainty inherent to the age-depth modelling and pollen resampling

10 procedures in our analyses.

11

12

13 **Pollen diversity and composition measures**

14

15 In these analyses, we examine the effects on diversity of the major depopulation and cultural
16 shifts that occurred across Europe during the 14th Century. This required a pollen dataset
17 with pollen samples before and after the disturbances of the 14th Century. We thus filtered
18 our total Common Era pollen dataset to include only records with pollen samples across the
19 period of interest, filtering out those pollen records that did not have at least one pollen
20 sample present in the 13th Century (1200 – 1300 CE) and the 15th Century (1400 – 1500
21 CE); i.e., spanning the 14th Century (see *Supporting Information: Extended Figures and*
22 *Tables, Extended Fig. 1G* summaries of the numbers of samples present in the two time
23 windows across the 109 records). Following the implementation of the criteria described in
24 *Data filter*, the inclusion of only records with the requisite REVEALS information, the
25 resampling procedure (exclusion of all samples with fewer than 150 grains) and the selection
26 of only records with samples spanning the 14th Century, our final Common Era dataset
27 included 109 unique pollen records constituted by 4616 unique REVEALS-adjusted pollen

1 samples spanning the period 0 – 1850 CE (*Supporting Information: Extended Figures and*
2 *Tables, Extended Fig 1*).

3

4 For each of the 1,000 datasets, we computed the three measures of within-pollen record
5 diversity: richness, evenness and compositional turnover (temporal beta diversity).

6

7 Richness and evenness

8 Richness represents the number of unique pollen types present in a standardised pollen
9 sample (Birks & Line 1992); 150 grains in these analyses. Evenness is a measure of
10 dominance, i.e., a measure of how a sample's total pollen sum is shared among pollen
11 types. A pollen sample where the majority of pollen grains are identified as one or a few
12 pollen types is considered 'uneven' (i.e., has low evenness values), whereas a pollen
13 sample where the total pollen sum is shared more equally among pollen types is considered
14 'even' (high evenness values). We use Pielou's evenness (Pielou 1966) in these analyses,
15 computed for each pollen sample using the **vegan** package (Oksanen *et al.* 2022).

16

17 Pollen richness is associated with the evenness of samples, such that strongly uneven
18 samples are likely to have lower richness values than more even samples (Giesecke *et al.*
19 20a), a problem exacerbated by those confounding issues of differential pollen production,
20 dispersal, preservation, etc., mentioned above (see *Modelling past vegetation covers from*
21 *pollen counts*). We mitigate against these biases, as far as is possible, by adjusting pollen
22 counts using REVEALS past vegetation estimates. Pollen richness has been shown to
23 represent the richness of parent plant communities in European studies (e.g. Felde *et al.*
24 2016; Meltsov *et al.* 2011; Papadopoulou *et al.* 2022; Senn *et al.* 2022; Väli *et al.* 2022),
25 though elsewhere this relationship does not always hold e.g., (Goring *et al.* 2013).

26

27

1 Compositional turnover (temporal beta diversity)

2 We computed the compositional turnover of pollen samples between successive samples
3 through each pollen record using the Bray-Curtis index (Bray & Curtis 1957). Bray-Curtis
4 accounts for differences in the abundances as well as identities of pollen types between
5 samples, which we computed using the `vegdist` function from `vegan` (Oksanen *et al.*
6 2022). We adjusted for the time interval between successive samples in the same way as in
7 Gordon *et al.* (2024). Briefly, this involved subtracting the expected difference in Bray-Curtis
8 (given the time interval between samples) from the actual value, and normalising this value
9 to be between [0,1] [see Gordon *et al.* (2024) for full details].

10

11 Agricultural land and tree cover

12 Alongside our pollen diversity metrics, we also analyse two metrics that describe changes to
13 the composition of pollen samples: i) Agricultural land and ii) tree cover (Fig. 2). Table 1 in
14 Githumbi *et al.* (2022) describes the land cover classes (LCCs) and plant functional types
15 (PFT) used by Wolf *et al.* (2008) to group together vegetation estimates from different pollen
16 types. We implement the methodology from Githumbi *et al.* (2022) for our REVEALS pollen
17 count transformation stage and we group the resultant vegetation estimates using the same
18 LCC and PFT groups. Our tree cover group comprises all pollen types included in the
19 'Evergreen trees' LCC and 'Summer-green trees' LCC, whilst our Agricultural land group
20 contains pollen counts from the pollen types *Cerealia t.* and *Secale cereale*, and is identical
21 to the PFT of the same name in Table 1 of Githumbi *et al.* (2022).

22

23 Fourteenth Century agricultural land trajectory

24 To understand how changes in cereal cultivation related to changes in each diversity metric,
25 we subset pollen records into three categories based on their changing cereal pollen
26 representation across the 14th Century. For each resampled pollen record (1,000 resamples
27 per record), we first computed the mean cereal count of all pollen samples that fell into time

1 period one, 1200 – 1300 CE, and for all pollen samples that fell into time period two, 1400 –
2 1500 CE (such that each pollen record was represented by an average cereal count value
3 for both periods). We then computed the difference between these two average cereal count
4 values to provide the ‘trajectory’ of change over the 14th Century for that resampled pollen
5 record. We categorised these change values into (i) ‘cereal farming continuation/expansion’;
6 pollen records that had non-zero cereal counts and that remained the same/showed
7 increases in cereal cultivation, (ii) ‘cereal farming abandonment’; pollen records that showed
8 decreases in cereal cultivation and (iii) ‘non-cereal farming locations’; pollen records that had
9 zero cereal pollen represented in both the 1200 – 1300 CE period and the 1400 – 1500 CE
10 period. We repeated this categorisation procedure for all 1,000 resamples of each pollen
11 record and carried through to the modelling stage (described below).

12

13 It is worth noting that the reduction of REVEALS-transformed cereal counts in a pollen
14 record is most plausibly a consequence of reduced areal coverage of cereal crops, in some
15 cases to zero (8.9% of resampled records showing declines to zero) and in other cases to a
16 lower level (median reduction for resampled records where cereal prevalence reduced =
17 64.1% decline, 25th quantile = 100% decline, 75th quantile = 33.3% decline). Reduced
18 intensity in a similar area cultivated is theoretically possible, but less likely because of the
19 large effort required to cultivate land in pre-industrial agricultural systems.

20

21

22 **Temporal trends**

23

24 Given our large Common Era dataset (n = ~5,000 unique pollen samples x 1,000 resamples)
25 and the computational expense of fitting Generalised Additive Models with many thousands
26 of parameters (smooths and random effects), it was computationally intractable to model the
27 total dataset of 1,000 resamples per diversity metric through time in a single model. To

1 overcome this, we employed a bootstrapping and posterior simulation procedure to ensure
 2 that the variation across all 1,000 resampled diversity metrics and their per-sample age-
 3 distributions, alongside statistical uncertainty associated with model fitting, was carried
 4 through these analyses.

5

6 To ensure computational feasibility, we subset the total dataset of 1,000 resamples per
 7 metric into ten blocks, with each block containing 100 unique resamples. For each block of
 8 100 resamples per pollen metric (richness, evenness, turnover, agricultural land), we fit a
 9 Generalised Additive Model to:

10

- 11 1. The full dataset, i.e., a Europe-scale analysis (Figs. 1A-D)
- 12 2. The full dataset, subset by 14th Century cereal trajectory (Figs. 1E-H)
- 13 3. The full dataset subset (spatially) by each vegetation zone (Lang *et al.* 2023) (Figs.
 14 4A-C).

15

16 For each metric, block and spatial subset of the data (1-3, above) combination (*Metric_i*,
 17 below), we fit a GAM to the pollen data as a function of time and a random intercept term for
 18 each unique dataset ID. These models took the form,

19

```
20 Metric_i ~ s(age_CE, by = "resample", bs = "fs") +
21           s(datasetid, bs = "re"),
22           discrete = T,
23           method = "fREML",
24           nthreads = 50,
25           family = fam.
```

26

27 We chose the Poisson family for the richness models (given the count response) and the
 28 beta family for the evenness and turnover models (given the [0, 1] constraints of the data).

1 We chose the negative binomial family for the Agricultural land models, given the
 2 overdispersed nature of the count response. We discretized all variables and fit these
 3 models using fast restricted maximum likelihood (fREML) and `bam` to speed up model fitting
 4 (Wood 2011). We performed model checks by inspecting histograms of residuals and
 5 residuals versus fitted plots, which did not show any patterns and were thus deemed
 6 acceptable. We also checked the outputs from `mgcv`'s `gam.check` function to ensure that
 7 all models had converged and that the wiggleness provided for each smooth (k) was
 8 sufficient.

9

10 We combined the fitted smooths across the ten blocks per metric and spatial unit using a
 11 posterior simulation approach. For each model fit, we simulated 100 draws from the
 12 posterior of each estimated function using the `fitted_samples` function from `gratia`
 13 (Simpson 2024). Given the estimated model parameters, this function generates posterior
 14 draws (i.e., realisations) for each fitted function to include the uncertainty in the expected
 15 values. This posterior simulation procedure resulted in 100 draws associated with each of
 16 the 1,000 (original) resamples, which resulted in 100,000 individual draws for each metric,
 17 spatial unit (Europe/vegetation zone) and cereal trajectory combination through time across
 18 the total dataset. We summarise and present these 100,000 draws using the median and a
 19 90% credible interval for each metric (Fig. 1, Fig 4A-C and *Supporting Information: Extended*
 20 *Figures and Tables, Extended Fig. 2*).

21

22 To quantify the rate of change (and subsequently check if they are predominantly positive or
 23 negative) through each draw, d , of each metric, m , for each spatial type, s , and cereal
 24 trajectory type, c , we approximated the first derivative (slope) of m , m' , using the method of
 25 central finite differences. That is,

26

$$27 \quad m'_{d,s,c}(t) \approx \frac{m_{d,s,c}(t+h) - m_{d,s,c}(t-h)}{2h},$$

1

2 i.e., the change in the metric between the next time point and the previous time point divided
3 by the total amount of time between the two time points. Here, h is the gap between two time
4 points and is set to be two, as the predictions are made across an evenly spaced grid at an
5 interval of two years over the Common Era.

6

7 We performed this procedure for each of the 100,000 temporal smooths per diversity metric,
8 spatial unit and cereal trajectory combination, which provided a distribution of derivative
9 estimates over the Common Era. We plot in bold time points for which 95% of the estimates
10 were either positive or negative (bold used for both), which provides a level of confidence in
11 the modelled temporal changes. We plot all the other time points in non-bold (Figs 1 & 4).

12

13

14 **Relationships between metrics**

15

16 The analyses described below included pollen samples from within the time period 1200 –
17 1300 CE only and thus included orders of magnitude fewer pollen samples than were
18 analysed in the procedures outlined in *Temporal trends* (above). Given the substantially
19 lower computational requirements to fit these models, for the following analyses we were
20 able to model the full dataset (i.e., all 1,000 resamples) in the *same* model (i.e., one GAM of
21 1,000 resampled datasets vs ten GAMs that each contained 100 resamples as in *Temporal*
22 *trends*). We performed the following analyses for the full European dataset (Fig. 3) and for
23 each vegetation zone (Fig. 4D-I).

24

25 Diversity and tree cover

26 Our expectation was that diversity values should be maximised in landscapes characterised
27 by approximately balanced levels of open versus wooded cover. To investigate this
28 expectation, we modelled i) how tree cover related to richness and, using predictions from

1 this model, ii) how tree cover change related to richness change spanning the 14th Century.
 2 We firstly subset all pollen records that were represented by pollen samples spanning the
 3 Black Death era (samples present in both the period 1200 – 1300 CE and 1400 – 1500 CE).
 4 We then modelled the richness data as a function of tree cover for those pollen samples that
 5 fell into the period 1200 – 1300 CE. We had no *a priori* expectation for the ‘best fitting’
 6 functional form of tree cover, so we used a GAM to estimate this (potentially smooth)
 7 relationship. This model took the form,

```
8
9     richness ~ s(tree_percent, bs = "tp",
10                k = 7, m = 2) +
11                s(datasetid, bs = "re") +
12                s(tree_percent, resample,
13                bs = "fs", k = 5, m = 2),
14     method = "fREML",
15     discrete = T,
16     nthreads = 10,
17     family = poisson.
```

18
 19 We included a random intercept term for each unique pollen record. We also allowed the
 20 functional form for the effect of tree cover on richness to vary by resample. The argument `bs`
 21 `= "fs"` selects the factor-smooth interaction basis type and `m = 2` penalises the spline
 22 (preventing overly wiggly forms) using the second derivative of the function. See Pedersen
 23 *et al.* (2019) for further details of fitting hierarchical generalised additive models using **mgcv**
 24 in **R**. We chose the Poisson distribution as the family, given the count response. Again, we
 25 discretized all variables and fit these models using fast restricted maximum likelihood
 26 (fREML) and `bam` to speed up model fitting (Wood 2011).

27

1 We performed the same model checks as for the temporal models and present the fit of the
 2 global `tree_percent` smooth, plus the data and prediction uncertainty for individual
 3 records across the 1,000 resamples, in Figs. 3A for Europe and in Figs. 4D-F for each
 4 vegetation zone.

5

6 According to our hypothesis, a reduction in tree cover in landscapes dominated by trees
 7 would generate biodiversity gains, whereas in landscapes largely devoid of trees, further
 8 reductions in tree cover would erode diversity measures (and *vice versa*). Using the above
 9 model trained using the 13th Century data only, we predicted the expected (temporal)
 10 change in richness given the actual changes in tree cover between the 13th and 15th
 11 Centuries (for each resample). This prediction provided the expected change in richness
 12 between the 13th and 15th Centuries for each pollen record, given the estimated smooth
 13 relationship between richness and tree cover presented in Fig. 3A. We also computed the
 14 change in the raw (observed) richness data for each pollen record over the same time
 15 period.

16

17 To identify how well the modelled expectation related to the actual changes observed, we
 18 fitted a linear model to the observed change in richness (`delta_observed_richness`,
 19 below) as a function of the predicted change in richness (`delta_predicted_richness`)
 20 across the 1,000 resamples, (Fig. 3B). This model took the form,

21

```
22     Delta_observed_richness ~ delta_predicted_richness +
23                               s(datasetid, bs = "re") +
24                               s(resample,
25                                 delta_predicted_richness, bs = "re"),
26                               method = "fREML",
27                               discrete = T,
```

```

1         nthreads = 10,
2         family = gaussian

```

3

4 We chose the Gaussian family for this model, given the continuous, zero-centred response.

5 We also included a random intercept term for each unique pollen record and a random slope

6 and intercept for each resample. Again, we discretized all variables and fit these models

7 using fast restricted maximum likelihood (fREML) and `bam` to speed up model fitting (Wood

8 2011).

9

10 We present the overall relationship between changes in predicted and observed richness,

11 plus the data and prediction uncertainty across the 1,000 resamples for Europe in Fig. 3B

12 and for each vegetation zone in Fig. 4G-I.

13

14

15 Diversity and cereals

16 We also were interested in how richness and evenness related to cereal representation in

17 the period immediately before the Black Death. We subset pollen records with samples

18 spanning the Black Death era, i.e., with samples in both the 1200 – 1300 CE and 1400 –

19 1500 CE time periods (in the same way as for the *Diversity and tree cover* models; see

20 above). The distribution of raw cereal counts were heavily right skewed, with a high

21 proportion of zeros and we therefore square-root transformed them. Then, we fit a GAM to

22 the 1200 – 1300 CE diversity data (richness and evenness, *Diversity* below), with a

23 smooth function of square-root transformed cereal counts as the explanatory variable. These

24 models took the form,

25

```

26     Diversity ~ s(sqrtcereal, bs = "tp", k = 10, m = 2) +

```

```

27         s(sqrtcereal, resample, bs = "fs",

```

```

28         k = 5, m = 2),

```

```
1     method = "fREML",  
2     discrete = T,  
3     nthreads = 10,  
4     family = fam.
```

5

6 We chose the Poisson family for the richness model (given the count response) and the beta
7 family for the evenness model (given the [0, 1] constraints of the data). Again, we allowed
8 the functional form and intercepts for the effect of `sqrtcereal` on `Diversity` to vary by
9 resample. We did not include a random intercept for each unique dataset ID, as in the tree
10 cover model, because it prevented convergence when included. We present the fit from the
11 global smooth plus the data and prediction uncertainty across the 1,000 resamples, in

12 *Supporting Information: Extended Figures and Tables, Extended Fig. 3.*

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53

1 **Extended Figures and Tables**

2

3

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5 **Extended Figure 1. Fossil pollen record data descriptions. (A)** Map describing locations
6 of fossil pollen records (N = 109) included in these analyses, with the size of the icon
7 representing the number of resamples (maximum 1,000) each pollen record is present in
8 (given variation arising from probabilistic age modelling procedure). **(B)** The number of
9 unique pollen records present in 100-year bins through the Common Era. **(C)** The number of
10 unique pollen samples (N = 4,616) present in 100-year bins through the Common Era. **(D)**
11 The number of unique samples per record in 100-year bins through the Common Era.
12 Central orange points represent the median numbers of samples per record across all
13 records present in each bin plus the 90% empirical interval (across records) around the
14 mean, represented by the grey bar. The dotted line represents the median across bins. **(E)**
15 The temporal coverage of each pollen record; each line represents one of the 109 records,
16 cohort ordered by age and duration. **(F)** The terrestrial pollen sums of all 4,616 pollen
17 samples included in these analyses prior to the REVEALS adjustment step. **(G)** The number
18 of pollen samples in the time window prior to the Black Death era (1200 – 1300 CE) and
19 after the Black death era (1400 – 1500 CE) across the 109 records included in these
20 analyses. Pollen samples in panels **(B - E & G)** are temporally partitioned based on their
21 median age across all 1,000 age draws from the posterior age distribution per sample. Note
22 the logarithmic y-axes for panels **(D, F & G)**. The mean and standard error-based 95%
23 confidence interval of the mean is shown as a yellow diamond and interval in panels **(F & G)**
24 - note that the 95% interval is equal in size to the mean symbol in **(F)** so is not visible.

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Extended Figure 2. Per-vegetation zone temporal patterns in each pollen metric. (A-C) Evenness, **(D-F)** turnover and **(G-I)** cereals. Generalised additive models fit to each metric as a smooth function of time. Central lines represent the average fit across the fitted distribution (see *Methods*), with sections of thicker lines indicating periods over which >95% of the temporal draws exhibit positive or negative slopes. Grey intervals represent the interquartile range across draws. Vertical black dotted lines show the bounds of the 14th Century.

Lack of 'significance' (thicker lines) for Mediterranean trends for richness (Fig. 4C), evenness and turnover reflects smaller sample sizes for this vegetation zone, given the reduced availability of pollen records for drier environments. Nonetheless, increased apparent values around 0 CE may reflect positive influences of earlier Mediterranean cultures on all three biodiversity metrics. It is unclear whether the growth of biodiversity metrics in the Boreal and Hemiboreal (Fig. 4A, panels A, D here) indicates limited impact of the Great Famine and Black Death and social-political changes (continuation of previous growth), or whether increases are directly linked to population relocation.

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Extended Figure 3. Cereal representation and diversity relationships. (A) Richness and **(B)** evenness as nonlinear functions of cereal counts for pollen samples in the period 1200 – 1300 CE for Europe. In both panels, the central points represent the median and the crosshairs represent the variation (25th and 75th percentiles) across the 1,000 resamples in both diversity (richness and evenness) and cereal counts.

1 **Extended Table 1. Model summaries, Fig. 3.** Sheet 1: Tree cover and richness
2 relationship, from Fig. 3A. Sheet 2: Change in predicted and change in observed richness
3 relationship, from Fig. 3B. Parameter estimates and fitness estimates.

4 **Extended Table 2. Model summaries, Fig. 4D-F.** Per-vegetation zone tree cover and
5 richness relationships, from Figs. 4D-F. Parameter estimates and fitness estimates.

6 **Extended Table 3. Model summaries, Fig. 4G-I.** Per-vegetation zone change in predicted
7 and change in observed richness relationships, from Figs. 4G-I. Parameter estimates and
8 fitness estimates.

9 **Extended Table 4. Metadata for fossil pollen records included in these analyses.**

10 Columns: Sitename (unique site name), datasetid (unique Neotoma identifier per pollen
11 record), collunitid (unique Neotoma collection unit identifier detailing pollen record origin),
12 n_samples (number of pollen samples per record), n_chroncontrols (number of chronological
13 controls per record), long (longitude in DD), lat (latitude in DD), citation (author names and
14 DOI for original publication).

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