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Mineral protection regulates long-term global preservation of natural organic carbon

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18 The balance between photosynthetic organic carbon production and respiration controls atmospheric composition and climate^{1,2}. The majority of organic carbon is respired back to 19 20 carbon dioxide in the biosphere, but a small fraction escapes remineralization and is preserved over geologic timescales³. By removing reduced carbon from Earth's surface, 21 22 this sequestration process promotes atmospheric oxygen accumulation² and carbon dioxide 23 removal¹. Two major mechanisms have been proposed to explain organic carbon 24 preservation: selective preservation of biochemically unreactive compounds^{4,5} and protection resulting from interactions with a mineral matrix^{6,7}. While both mechanisms can 25 play a role across a range of environments and timescales, their global relative importance 26 on 10³- to 10⁵-year timescales remains uncertain⁴. Here we present a global dataset of the 27 distributions of organic carbon activation energy and corresponding radiocarbon ages in 28 29 soils, sediments, and dissolved organic carbon; we find that activation energy distributions 30 broaden over time in all mineral-containing samples. This result requires increasing bondstrength diversity, consistent with the formation of organo-mineral bonds⁸ but inconsistent 31 32 with selective preservation. Radiocarbon ages further reveal that high-energy, mineral-33 bound organic carbon persists for millennia relative to low-energy, unbound organic carbon. Our results provide globally coherent evidence for the proposed⁷ importance of 34 35 mineral protection in promoting organic carbon preservation. We suggest that similar 36 studies of bond-strength diversity in ancient sediments may elucidate how and why organic 37 carbon preservation-and thus atmospheric composition and climate-has varied over geologic time. 38

39 Two classes of mechanisms—selectivity and protection—have been proposed to explain why some organic carbon (OC) escapes remineralization in soils and sediments⁴⁻⁷. Biochemical 40 41 selectivity hypotheses state that intrinsically bioavailable compounds such as sugars and amino 42 acids are rapidly respired, whereas "recalcitrant" (macro)molecules such as lignin are selectively preserved due to their low energy yield, large size, and/or a lack of enzymes that can decompose 43 them^{4,5}. Selective preservation has been extensively documented in dissolved OC (DOC)⁹, 44 45 decaying woody tissue¹⁰, and sapropel sediments containing almost exclusively organic matter⁵. 46 In contrast, protection hypotheses state that particles shield OC from respiration regardless of intrinsic recalcitrance, potentially due to occlusion within pore spaces that are inaccessible to 47 microbes and their extracellular enzymes^{4,8,11-14}. Specifically, protection often involves 48

49 physicochemical interactions with a mineral matrix. Mineral protection is evidenced by

50 laboratory incubation experiments¹⁵, non-selective OC preservation in sinking marine particles¹⁶,

51 and strong observed correlations between OC content, iron content 17,18 , and mineral surface

52 area^{7,8,19} in soils and sediments.

Although both selectivity and protection must play some role across a range of environments and timescales, directly measuring their relative importance using traditional geochemical techniques has proven difficult. To address this issue, first we develop a conceptual framework that predicts how chemical compositions, as recorded by OC bond strength, and preservation timescales evolve when OC is exposed to each mechanism. We then compare predictions to results from a global dataset.

59 Figure 1 illustrates our model. Selective preservation states that OC diversity—that is, 60 the distribution of carbon bond strengths contained within a given sample—decreases as certain 61 chemical structures are preferentially respired. OC chemical composition becomes more homogeneous with time (Fig. 1a, top). In contrast, mineral protection allows for respiration of 62 different chemical structures at similar rates. However, some OC molecules become occluded 63 within aggregates^{8,14} or bound to mineral surfaces during aging and persist for longer than they 64 would in the absence of the mineral matrix (Fig. 1a, bottom). Because bound, stabilized 65 66 compounds represent a new chemical and energetic state relative to their unbound counterparts^{20,21}, we expect bond-strength diversity to increase with time when OC is protected. 67 68 Whereas both mechanisms predict a similar decrease in the amount of OC remaining (Fig. 1b, 69 top), they produce opposite trends for the temporal evolution of bond-strength diversity (Fig. 1b, bottom). 70

71 Information on bond-strength diversity is contained in a continuous function p(E) that 72 describes the fraction of OC associated with each activation energy, E. Critically, in addition to 73 the energy needed to break covalent bonds (a function of OC molecular structure), E includes the energy required to remove bound carbon from particles²⁰⁻²². Here, we estimate p(E) by 74 75 determining the bond-strength distribution that best predicts the measured OC decay profile 76 when a sample is exposed to increasing temperature using ramped pyrolysis/oxidation $(\text{RPO})^{22}$. E distributions are unique to each remineralization pathway²²; we therefore treat p(E) calculated 77 by RPO as a proxy for OC bond-strength diversity but emphasize that thermal E is not equivalent 78 79 to the activation energy of any environmental enzymatic respiration reaction (Methods). Our

- 80 interpretation is informed by *E* distribution differences between samples subject to the same
- 81 experimental conditions, not by absolute E values. In addition to providing an estimate of p(E),
- 82 RPO quantifies the OC radiocarbon activity $({}^{14}C/{}^{12}C)$, expressed in ${}^{14}C$ yr), a measure of the time
- 83 that has passed since initial formation, over multiple decay windows. In doing so, RPO results
- 84 allow us to determine how OC age varies as bonds become stronger.

85 To assess global trends across all environments, this study compiles RPO results from riverine and marine DOC, soils, riverine particulate OC (POC), and marine sediments (Extended 86 Data Fig. 1). These samples were primarily obtained from aerobic environments and represent 87 oxic respiration, although some were exposed to periodic anoxia during their diagenetic history 88 89 (e.g., Arabian Sea sediments). This global dataset includes samples analyzed using slightly 90 different protocols (e.g., pyrolysis vs. oxidation), but an uncertainty assessment indicates that 91 methodological variability is small and that complex kinetic behavior during sample heating is unlikely to explain observed trends (Methods)^{22,23}. Furthermore, because our focus is the 92 93 persistence of biospherically produced OC, we excluded samples containing carbonates, 94 anthropogenic OC, and/or rock-derived OC, resulting in a final dataset of 62 samples. This 95 screening procedure removes permafrost-dominated soils since these samples exhibit anomalously old ¹⁴C ages due to protracted storage under frozen conditions. It is possible that 96 97 some retained samples contain minor rock-derived OC inputs. However, such contributions 98 never exceed approximately 5% of total OC and cannot explain resulting E and 14 C trends 99 (Methods).

Figure 2 shows the E distributions for each sample, grouped by environment. We 100 101 observe significant OC bond-strength diversity differences between DOC and all mineralcontaining samples. Here, we quantify OC bond-strength diversity in terms of σ_E , the standard 102 103 deviation of the distribution p(E). DOC exhibits the narrowest distribution of any sample type. In 104 contrast, p(E) is substantially broader in all mineral-containing sample types, indicating greater 105 OC bond-strength diversity. We additionally quantify the mean E value, μ_E , for each sample type 106 as a measure of the average OC bond strength. Across sample types, μ_E increases in the order: soils, river POC, DOC, and marine sediments. Finally, intra-sample ¹⁴C trends for mineral-107 containing samples show a globally coherent increase in 14 C reservoir age with increasing E. 108 High-E OC is consistently thousands of 14 C yr older than low-E OC, especially within riverine 109 110 POC and marine sediments, indicating protracted preservation of high-E material in these

environments. In contrast, DOC samples show either the opposite trend or no relationship between E and ¹⁴C age.

113 To determine whether these observations are diagnostic of a particular preservation 114 mechanism, we compare the observed temporal evolution of σ_E to that predicted by selective 115 preservation and by mineral protection. Because the timescale of exposure to respiration cannot 116 be measured directly, we approximate it using OC content $(\%OC)^{24}$. This approximation assumes that loss by respiration is the dominant control on %OC; it ignores secondary effects 117 118 such as dilution. This is supported at the global scale by the observation that intra-sample ¹⁴C age 119 variances increase with decreasing %OC for all mineral-containing samples following a single 120 power-law relationship (Extended Data Fig. 2). This relationship persists across ~ 3 orders of 121 magnitude in %OC despite large differences in sample latitude, physical environment, grain size, 122 etc., suggesting that samples with lower %OC have been exposed to aerobic respiration for longer timescales $(c.f., Fig. 1b, top)^6$. Furthermore, soils generally contain higher %OC and 123 124 lower ¹⁴C age variances than do river POC and marine sediments, consistent with the idea that 125 sediments represent longer temporal integration. Thus, while we do not expect this relationship 126 to strictly hold at the individual soil/sediment profile level, %OC remains a robust respiration 127 timescale proxy for global theory-data comparisons.

128 We test the hypothesis that selective preservation is the dominant control by quantifying 129 how p(E) evolves following this mechanism. We treat each OC component as decaying according to a rate constant defined by its bond strength. Because high-E OC is generally older 130 131 than low-E OC in our dataset, we assume a decreasing rate constant with increasing E. High-E 132 material is therefore selectively preserved, as expected if stronger bonds are inherently more 133 resistant to degradation⁴. The distribution p(E) can then be obtained at each point in time by 134 solving the first-order differential equation that describes the evolution of p(E) as %OC 135 decreases (Methods). Using a Gaussian distribution for the initial condition, we obtain the 136 results shown in Fig. 3a. In contrast, Fig. 3b shows the observed average p(E) for all mineral-137 containing samples binned by %OC. Although selective preservation is consistent with the observed narrow DOC distribution⁹, it fails to reproduce observed p(E) trends for all mineral-138 139 containing samples, which broaden with decreasing %OC and suggest increasing bond-strength diversity. 140

141 To better quantify bond-strength evolution through time, we bin all mineral-containing 142 samples and plot μ_E and σ_E as a function of %OC (**Fig. 4**). This approach approximates a time-143 series. To estimate the autotrophic biomass starting composition, we include three samples 144 representing mineral-free, vascular-plant OC: woody tissue, cellulose, and a Hawaiian soil 145 organic (O) horizon.

146 **Figure 4b** reveals increasing σ_E with increasing time of exposure to respiration, 147 consistent with low σ_E values for autotrophic biomass and with previously reported p(E)evolution from a Hawaiian soil profile²⁵. In contrast, selective preservation predicts that σ_E must 148 149 decrease with time. Conceivably, selective preservation combined with resupply of respired, 150 low-E material could lead to a nearly constant σ_E over time, but it cannot explain the observation 151 that p(E) broadens beyond its initial distribution. Specifically, resupply fails to predict the observed increase in p(E) at the high-E end of the distribution as OC ages. Rather, mineral 152 153 surfaces likely broaden p(E) relative to autotrophic biomass by protecting otherwise bioavailable OC and by strengthening OC bonding environments^{20,21}. Secondary condensation^{4,26} reactions 154 155 have additionally been hypothesized to occur in the environment; these reactions could similarly 156 form new bonds and thus influence E distributions. However, condensation is likely not the 157 major driver of observed global E trends since (i) this would require DOC to behave similarly, 158 which is not observed, and (ii) this would not necessarily lead to the formation of stronger bonds 159 and thus higher-E OC, which is observed. Still, condensation may play a minor role in OC 160 preservation.

161 We propose that some OC fraction becomes strongly bound to minerals during aging in 162 the environment, potentially due to ligand exchange or direct chelation with iron oxides¹⁸, and 163 leads to the production of high-*E* material that persists for millennia. This is supported by 164 observed intra-sample ¹⁴C trends, which show a global increase in ¹⁴C reservoir age with 165 increasing *E*. High-*E* OC is consistently thousands of ¹⁴C yr older than low-*E* material, especially 166 for riverine POC and marine sediments.

167 All mineral-containing samples additionally include lower *E* OC than that found in DOC 168 and autotrophic biomass, leading to only a weak increase in μ_E with decreasing %OC (**Fig. 4a**) 169 despite significant ingrowth of high-*E* material. This requires production and preservation of 170 low-*E* compounds during heterotrophic alteration in soils and sediments²⁵⁻²⁸. Microbial 171 heterotrophs are known to produce an abundance of small molecules containing

carboxyl/carbonyl (C=O) and hydroxyl (C-OH) carbon atoms (e.g., amino acids, sugars)^{26,28}, for 172 example during extracellular hydrolysis of polymers²⁷. These compounds can then interact with 173 174 particle surfaces via cation bridging and/or hydrophobic interactions and become protected. In 175 contrast, the lack of particle interactions further explains the absence of low-E material and 176 narrow ¹⁴C gradients for DOC; small, bioavailable molecules in mineral-free environments are free to be selectively respired over short timescales. For example, laboratory incubation 177 178 experiments show increased heterotrophic respiration rates once weakly bound OC is desorbed from particles²⁹. Furthermore, Barber *et al.*¹⁷ have shown that C=O and C-OH carbon atoms, 179 which exhibit low inherent E values²¹, are stabilized by binding to iron minerals. 180

181 Combined, these observations imply weak mineral association (possibly due to occlusion 182 within aggregates^{8,14}) as a mechanism to preserve low-*E* OC, consistent with its absence from 183 DOC and vascular-plant biomass. However, unlike high-*E* material, which ages for millennia, 184 low-*E* OC is continuously adsorbed/desorbed and resupplied by alteration of recently produced, 185 14 C-enriched biomass. This leads to the observed increase in 14 C reservoir age with increasing *E*.

186 Mineral protection carries implications for the fate of OC in soils and sediments and for 187 atmospheric composition throughout Earth history. Whereas soils, riverine POC, and marine sediments appear to follow a globally coherent σ_E vs. %OC trend today (Extended Data Fig. 2), 188 these trends could have differed in the geologic past due to changes in oxidant availability, 189 190 mineral surface area, surface charge, and/or OC composition (e.g., availability of organic ligands). We propose that such changes are reflected in the distribution p(E) of geologically 191 192 preserved OC. For example, production of the secondary clay mineral smectite during weathering on land has been shown to promote protracted OC storage³⁰. Furthermore, by 193 194 enhancing the deposition of OC-rich sediments in marine continental margins, smectite 195 production has been proposed as a carbon sequestration mechanism during warm periods such as 196 Cretaceous ocean anoxic events³¹. Assuming the mechanism proposed here equally applies to the 197 geologic past, then our framework provides a means to directly test this hypothesis using the 198 evolution of p(E) through time. In general, we predict that strong interactions with mineral surfaces expand OC bond-strength diversity and increase preservation efficiency, thus decreasing 199 200 atmospheric CO₂ levels and promoting oxygen accumulation.

201

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283 J.D.H., K.E.G., and S.Z.R. performed laboratory measurements; J.D.H. and D.H.R. developed

theoretical models and analyzed data; J.D.H., K.E.G., L.A.D., and V.V.G. provided samples;

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- 287

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291

292 **Data Availability:** All data that support the findings of this study are available on the "RPO

293 Compilation Online Database" with the identifier doi:10.5281/zenodo.1158742.

294 **Figure Legends:**

295 Figure 1 | Schematic representation of two potential preservation mechanisms. Panel (a) 296 shows a cartoon of each mechanism: (top) selective preservation, in which certain compounds 297 are preferentially preserved and (bottom) mineral protection, in which OC becomes altered, 298 attached to particles surfaces, and stabilized. For selective preservation, OC is initially composed 299 of labile (white) and recalcitrant (black) compounds, which decay at different rates. In contrast, 300 for mineral protection, both black and white circles have the same inherent lability. Either type 301 can then become stabilized by clay particles (large parallelograms) and persist. Panel (b) shows 302 how the amount (number of circles) and diversity (different colors and bound/free states) evolve 303 with time for each mechanism.

304

Figure 2 | *E* and ¹⁴C age distributions. (top) Each gray line represents the *E* distribution for one 305 306 sample; the thick black line is the average E distribution for each sample type: (a) DOC (riverine 307 and marine; average $\mu_E = 163.3$ kJ mol⁻¹; average $\sigma_E = 14.8$ kJ mol⁻¹), (**b**) soil OC (average $\mu_E =$ 153.8 kJ mol⁻¹; average $\sigma_E = 20.1$ kJ mol⁻¹), (c) riverine POC (average $\mu_E = 160.1$ kJ mol⁻¹; 308 average $\sigma_E = 25.3$ kJ mol⁻¹), and (d) sedimentary OC (average $\mu_E = 168.9$ kJ mol⁻¹; average $\sigma_E =$ 309 24.0 kJ mol⁻¹). The area under each curve integrates to unity [*i.e.*, p(E) is a probability density 310 311 function]. To compare all sample types side-by-side, p(E) units are not shown. (bottom) Each line represents the ¹⁴C reservoir age as a function of E for one sample; note that reservoir age is 312 313 presented on a logarithmic scale. The number of samples of each type (n) is shown in each panel. Reservoir age for each RPO fraction is equal to the ¹⁴C age for that fraction minus the youngest 314 315 ¹⁴C age of any fraction within a given sample (Eq. 5). OC in mineral-containing samples 316 becomes older with increasing E whereas DOC samples exhibit either the opposite trend or no 317 relationship.

318

Figure 3 | Predicted and observed p(E) evolution. p(E) as a function of %OC for (a) selective preservation predictions and (b) measured observations. The area under each curve integrates to unity [*i.e.*, p(E) is a probability density function]. For panel (b), all mineral-containing samples for which %OC data exist (n = 55) were binned by OC content and average *E* distributions were calculated for each bin: black, OC $\ge 10\%$ (median OC = 15.5%; n = 6); dark gray, $10\% > OC \ge$ 1% (median OC = 1.7%; n = 26); light gray, OC < 1% (median OC = 0.8%; n = 23). For panel 325 (a), the black line is a Gaussian distribution with %OC, μ_E , and σ_E equal to those for the

- 326 observed OC > 10% median distribution in panel (b). This distribution then evolves following
- selective preservation (**Methods**) until OC content reaches 1.7% (dark gray) and 0.8% (light
- 328 gray), the median values for measured $10\% > OC \ge 1\%$ and OC < 1% bins in panel (b). Selective
- 329 preservation cannot explain observed trends.
- 330

331 Figure 4 | Predicted and observed μ_E and σ_E evolution. Measured (black circles) and 332 predicted (dotted line, gray line) (a) μ_E and (b) σ_E evolution as a function of %OC. To visualize 333 global trends, all mineral-containing samples for which %OC data exist (n = 55) were binned and plotted at the mean value for each bin (11 bins containing 5 samples each). Additionally, 334 335 autotrophic biomass samples (cellulose, woody tissue, and O horizon soil; n = 3) were binned 336 and plotted as a white circle at OC = 50 %, a canonical value for photosynthetic biomass. Black 337 bars are $\pm 1\sigma$ uncertainty for each bin. The dotted line shows selective preservation predictions 338 starting from the measured phototrophic biomass composition. The gray line shows selective 339 preservation predictions starting from an initial Gaussian composition with μ_E and σ_E that were 340 chosen to best fit the observed data (Methods). Both predictions clearly cannot explain observed 341 trends, especially for σ_E . The %OC axis is presented on a logarithmic scale and is reversed to 342 emphasize that OC content generally decreases with time.

343 Methods:

Ramped pyrolysis/oxidation instrumental setup. The RPO instrument continuously heats OC, monitors evolved CO₂, and traps CO₂ for isotope analysis. RPO methodology has been described previously^{22,32,33}, and further information describing the detailed protocols used for each sample compiled in this study can be found in the original publications. In summary, all samples were either operated in oxidation mode (carrier gas: 92% He, 8% O₂) or pyrolysis mode (carrier gas:

349 100% He) with a flow rate of 35 mL min⁻¹ and a ramp rate of 5 $^{\circ}$ C min⁻¹.

350

Published data compilation. We compiled all published RPO results that include ¹⁴C

352 measurements^{25,33-47}. All thermograms and blank-corrected RPO fraction data (temperatures,

353 masses, ¹³C contents, and ¹⁴C activities) can be accessed in the online database⁴⁸. Samples

included in this database span five continents and a latitude range between 70.5° N and 64.5° S

355 (Extended Data Fig. 1). Samples were categorized as either: DOC, soil, riverine POC, or marine

sediment. Sample collection and pre-treatment methods are described in detail in each of the

357 original publications and are summarized in **Extended Data Table 1**.

358

New samples analyzed for this study. In addition to previously published results, we included 359 360 in our database five samples presented for the first time in this study. These new samples were 361 collected and treated as follows: (i) and (ii) Congo River DOC and surface POC collected near Brazzaville, Republic of Congo (4.3° S, 15.2° E) in April 2016. Approximately 100 L of water 362 363 was filtered through 0.22 µm polyether sulfone (PES) membrane filters. Sediments were retained 364 on the filters and 2 L of effluent water was collected in acid-leached HDPE bottles (10 % HCl, 1 week) for DOC extraction. Sediments were freeze-dried at -40 °C, whereas DOC was extracted 365 366 using bond-elut PPL cartridges, eluted in MeOH, transferred into pre-combusted quartz reactor inserts (850 °C, 5 hours), and dried under a stream of ultrapure N₂ gas⁴⁹. (iii) Grassland soil 367 368 overlying the Pololu lava flow located on the Kohala Peninsula of Hawaii (20.1° N, 155.7° W). 369 Archived material from 80 to 90 cm depth was freeze-dried at -40 °C, homogenized, and sub-370 sampled for RPO analysis. (iv) Kosi River surface suspended sediment (sample LO765 in Galy et al.⁵⁰]. Surface water was collected near Chatra, Nepal (26.9° N, 87.2° E) in 2007 and was 371 filtered through 0.22 µm PES membrane filters, de-carbonated over HCl fumes at 60 °C for 72 372 hours, rinsed 3× with 18.2 MΩ MilliQ water, and freeze-dried at -40 °C. (v) Suwannee River 373

374 DOC sample NOM II obtained from the International Humic Substances Society (2R101N; 30.8°

 $N, 82.4^{\circ}$ W). Sample material was concentrated by reverse osmosis from large water volumes (~

376 36,000 L), desalted via cation exchange, freeze-dried, and homogenized⁵¹. This sample was

analyzed in freeze-dried powder form as provided by IHSS.

378

379 Activation energy calculations. To quantitatively compare OC bond strengths between samples, 380 we converted observed thermograms into activation energy distributions. In addition to bond strength, thermograms depend on experimental conditions such as oven ramp rate⁵². This 381 dependency can be removed by treating OC decay as a function of E, the activation energy 382 383 required to volatilize a given carbon atom. *E* reflects the energy required to fully oxidize each carbon atom when exposed to a particular oxidation reaction pathway, including energy needed 384 to break both covalent bonds and secondary interactions with mineral surfaces^{20,21,38}, and is 385 386 therefore a suitable quantitative proxy for bond strength. We emphasize that E values calculated 387 here reflect thermal oxidation and are not quantitatively applicable to other (e.g., enzymatic) 388 oxidation pathways.

We calculated *E* distributions by finding the inverse solution to a set of parallel, nonisothermal, first-order kinetic decay reactions²² as implemented in the 'rampedpyrox' package for Python v.3.5 (ref. 53). Model details, mathematical derivations, and sensitivity analyses can be found in Hemingway *et al.*²². As recommended, here we set the Arrhenius pre-exponential factor to be 10^{10} s⁻¹ for all samples. Finally, to compare *E* between samples, we extracted two scalar metrics from each continuous distribution. First, the mean value for a given sample, termed μ_E , was calculated as the first moment of *E*:

$$396 \quad \mu_E = \int_0^\infty Ep(E)dE. \tag{1}$$

397 Second, the standard deviation of *E* for a given sample, termed σ_E , was calculated as the sample 398 variance of *E*:

399
$$\sigma_E = (\mu_{E^2} - [\mu_E]^2)^{1/2}.$$
 (2)

400 **Radiocarbon calculations.** All radiocarbon measurements were initially reported as Fm, the 401 "fraction modern," and were subsequently converted to ¹⁴C years. Fm is equal to the ¹⁴C activity 402 of a given sample (normalized to $\delta^{13}C = -25.0$ % VPDB to remove fractionation effects) relative 403 to 95% of the ¹⁴C activity of NBS Oxalic Acid I in 1950 (ref. 54). The time that has passed since 404 OC was initially fixed from CO₂ was then calculated in ¹⁴C yr as a function of Fm according to: 405 ${}^{14}C$ yr = -8033 ln(Fm), (3)

where 8033 is the Libby ¹⁴C half-life divided by ln(2). We derived two different metrics to
compare the spread in RPO-derived ¹⁴C ages between samples with different mean ages. First,
we calculated the age variance for each sample as

409
$$\operatorname{Var}({}^{14}\operatorname{C}\operatorname{yr}) = \operatorname{E}[({}^{14}\operatorname{C}\operatorname{yr} - {}^{\overline{14}}\operatorname{C}\operatorname{yr})^2],$$
 (4)

410 where $\overline{{}^{14}C}$ yr is the mean ${}^{14}C$ age for a given sample and E[x] denotes the expected value for 411 variable *x*. Second, we determined the reservoir age, R, for each RPO fraction within a given 412 sample relative to the most ${}^{14}C$ -enriched fraction from that sample (typically the fraction 413 collected over the lowest temperature range). Reservoir age was calculated as

414
$$R_f = {}^{14}C \operatorname{yr}_f - \min_f |{}^{14}C \operatorname{yr}_f|,$$
 (5)

415
$$f=1,...,n_f$$

416 where ¹⁴C yr_f is the ¹⁴C age for RPO fraction f, min \int_{f}^{14} C yr_f is the minimum ¹⁴C age measured

for any RPO fraction within a given sample, and n_f is the number of RPO fractions collected for that sample. Because all ¹⁴C decays at the same rate (*i.e.*, according to the ¹⁴C half-life), R is conserved with the passage of time⁵⁵. Finally, R can also be expressed in "fractional" notation

420 analogous to Fm according to:

421
$$F^{14}R = e^{-R/8033}$$
. (6)

422 All OC aging trends discussed in this manuscript are reported in 14 C yr and R notation, whereas

423 sample screening and end-member mixing calculations utilize Fm and $F^{14}R$ notation for

- 424 mathematical convenience (see "Sample screening," below). Finally, all ¹⁴C results reported here
- 425 were corrected for instrument-specific RPO procedural blank contamination 33,56 .

426

427 Uncertainty assessment. Because our dataset contains samples analyzed using multiple

428 methods, we assessed the influence of sample pre-treatment and RPO instrumental conditions on

429 resulting *E* distributions. This was done by subjecting the same samples to various experimental

430 conditions (**Extended Data Table 2**). Specifically, there are three methodological variables that

431 might influence OC thermal stability within our dataset: (i) whether or not a sample was
432 acidified^{38,39}, (ii) the type of acid treatment, and (iii) whether the RPO instrument was operated
433 in pyrolysis or oxidation mode.

First, the overall influence of acidification was determined using 4 carbonate-free samples: one natural Canadian soil³⁸ (sample CaSa_soil in <u>Extended Data Table 2</u>), two Hawaiian soils^{25,33} (PuuEke4 and Pololu1), and one Amazon River suspended sediment³⁹ (AmRi_ju2014_54m). PuuEke4 was analyzed in both pyrolysis and oxidation mode, leading to 5 acidified vs. non-acidified pairs. For these samples, acidification increased μ_E values by 5.6 ± 5.6 kJ mol⁻¹ and σ_E 3.8 ± 2.4 kJ mol⁻¹ (n = 5).

Second, the difference between HCl fumigation³³ and liquid HCl treatment⁴⁰ was assessed using one Hawaiian soil (Pololu1), one Himalayan river suspended sediment (PB-60), and one pure rock-derived OC (OC_{petro}) sample from Taiwan (LW_08-03_petro). Both PB-60 and LW_08-03_petro were analyzed in both pyrolysis and oxidation modes, leading to 5 acid treatment pairs. Fumigation decreased μ_E values by 6.7 ± 10.9 kJ mol⁻¹ and σ_E values by 2.4 ± 4.0 kJ mol⁻¹ relative to liquid acid treatment in these experiments (n = 5).

446 Finally, the difference between RPO pyrolysis and oxidation modes was determined 447 using the abovementioned samples and acidification protocols. For these samples, pyrolysis mode analysis appeared to increase μ_E values by 5.9 ± 4.8 kJ mol⁻¹ and σ_E values by 2.1 ± 2.3 kJ 448 mol⁻¹ relative to oxidation mode (n = 7). Additionally, Grant *et al.*²⁵ report only minor 449 450 differences between pyrolysis and oxidation modes in a set of Hawaiian soil samples. This similarity indicates that complex kinetic reactions such as charring²³ occurring within the RPO 451 instrument cannot fully explain our observed trends. If charring were driving RPO results, then 452 453 we would expect significantly higher μ_E values when the RPO instrument is operated in 454 pyrolysis mode due to the lack of atmospheric O₂; however, this is not observed. Furthermore, charring would act to shift low-E, ¹⁴C-enriched OC toward higher E values, thus artificially 455 dampening E vs. ¹⁴C signals and weakening observed trends. 456

Thus, none of these methodological differences can explain the signals observed in our dataset. Sample results span a μ_E range of 80 kJ mol⁻¹, significantly broader than both the largest bias (6.7 kJ mol⁻¹) and the widest uncertainty (±10.9 kJ mol⁻¹) observed in this assessment. Similarly, observed σ_E values in our dataset span a range of 35 kJ mol⁻¹, much broader than the maximum methodological bias (±3.8 kJ mol⁻¹) and uncertainty (±4.0 kJ mol⁻¹). We therefore

462 assume a conservative uncertainty of ± 10.9 kJ mol⁻¹ for μ_E and ± 4.0 kJ mol⁻¹ for σ_E , resulting in 463 a minimum signal-to-noise ratio above 7 for both metrics.

464

465 Sample screening. Because our focus was on the preservation of naturally produced biospheric 466 OC that has been continuously exposed to heterotrophic respiration since initial formation, we 467 screened the database for samples containing other carbon sources or alternate diagenetic 468 histories. Specifically, samples were omitted if they were shown to contain: (i) carbonates, (ii) 469 anthropogenically produced biospheric OC, (iii) rock-derived or "petrogenic" OC (OC_{petro}), or (iv) OC that has not been continuously exposed to diagenesis due to protracted storage in ice 470 (permafrost) or due to the rapid shutdown of pedogenesis (paleosols)⁴⁵. Screening was performed 471 472 as follows:

473 First, all samples that were not acidified prior to RPO analysis were screened for carbonates. This includes PPL-extracted and ultrafiltered DOC³³, Amazon River suspended 474 sediments³⁹, and a subset of soils^{25,38}. Because DOC cannot contain carbonates by definition, all 475 DOC samples were retained. For Amazon River sediments, Rosengard³⁹ analyzed one sample 476 477 (AmRi ju2014 54m) both un-acidified and after acid fumigation to directly test for the presence 478 of carbonates. Both sample treatments contain nearly identical RPO-fraction mass-weighted ¹⁴C activity (un-acidified: Fm = 0.820 ± 0.002 , fumigated: Fm = 0.770 ± 0.002) and ¹³C content (un-479 acidified: $\delta^{13}C = -29.3 \pm 0.1$ ‰ VPDB, fumigated: $\delta^{13}C = -29.1 \pm 0.1$ ‰ VPDB), a source tracer 480 481 that is highly sensitive to the presence of carbonates. Because of this similarity, especially for 482 ¹³C content, we conclude that any carbonate contribution to these samples is negligible, as discussed in the original publication³⁹. This is further supported by thermogram results, which 483 484 show negligible CO₂ production within the carbonate temperature range for these samples (approximately 600 to 800 °C)³². Furthermore, all soils collected from the Kohala Peninsula of 485 Hawaii²⁵ were retained for two reasons: (i) this landscape contains purely igneous bedrock, 486 487 precluding the possibility of detrital carbonates, and (ii) RPO thermograms do not show 488 significant CO₂ produced within the carbonate temperature range. In contrast, one unacidified Canadian soil likely contains pedogenic carbonates³⁸ and was removed. 489 490 Second, samples were screened for significant anthropogenic influence since the

491 inclusion of agricultural (e.g., fertilizers, pesticides) or urban (e.g., petroleum-derived

492 compounds) OC would likely influence RPO results. This resulted in the exclusion of only a
493 single Canadian soil sample collected from an agricultural plot³⁸.

494 Third, samples were screened for OC_{petro} contribution. Samples were removed if they met 495 any of the following three criteria: (i) If the original publication discusses the influence of OC_{petro} 496 or quantifies OCpetro content. This removes Taiwanese soils, Taiwanese river suspended sediments, and Himalayan river suspended sediments^{37,40}. (ii) If the Fm value for any RPO 497 498 fraction was statistically identical to zero. Because OC_{petro} is formed on timescales significantly longer than the ¹⁴C half-life, it is ¹⁴C-free by definition. Thus, an RPO fraction with Fm = 0499 would indicate purely OC_{petro} contribution to that fraction. This criterion further excludes 500 501 Himalayan river suspended sediments originally presented in this study. (iii) If RPO results 502 suggest that OC_{petro} contributes greater than 5 % of total OC. OC_{petro} contribution can be estimated using RPO results according to: 503

504
$$\% OC_{\text{petro}}^{\text{estimated}} = \sum_{(\mu_{E,f} - 2\sigma_{E,f}) > 185 \text{ kJ mol}^{-1}} (1 - F^{14} R_f) \left(\frac{m_f}{m_{\text{tot}}}\right) \times 100 \%,$$
 (7)

where $\mu_{E,f}$, $\sigma_{E,f}$, $F^{14}R_f$, and m_f are the mean and standard deviation of E, $F^{14}R$ value, and carbon 505 mass for RPO fraction f; 185 kJ mol⁻¹ is the minimum OC_{petro} E cutoff³⁷; and m_{tot} is the total 506 carbon mass for the sample. This estimate treats all OC contained within the OC_{petro} E range as a 507 mixture of OC_{petro} and un-aged biospheric OC (*i.e.*, $F^{14}R = 1$). Because it is likely that high-*E* 508 biospheric OC is described by $F^{14}R < 1$, we treat %OC_{petro} values calculated using Eq. 7 as 509 maximum estimates. This is supported by Himalayan river suspended sediments for which 510 independently calculated %OC_{petro} estimates exist⁵⁷ (samples KoRi sed LO765, 511 NaRi sed LO752, and NaRi sed PB60). For all samples (n = 3), %OC_{petro} values calculated 512 513 using <u>Eq. 7</u> are roughly double those reported previously⁵⁷, confirming that <u>Eq. 7</u> maximizes predicted %OCpetro values and that our screening procedure is therefore conservative. This 514 criterion further removes one Mississippi River suspended sediment⁴² and two Antarctic 515 sediment samples⁴⁴. Retained samples may contain minor amounts of OC_{petro}, especially fluvial 516 517 sediments due to their integrative nature. We therefore assessed the sensitivity of our results to 518 minor OC_{petro} contributions (see "OC_{petro} sensitivity analysis," below). 519 Finally, samples were screened for alternate diagenetic histories. Specifically, a sample

was removed if it has not been continuously exposed to heterotrophic remineralization since the
 time of OC formation due to (i) protracted storage under frozen conditions or (ii) rapid shutdown

of pedogenesis. Inclusion of such samples would conflate multiple preservation mechanisms; for example, since thawed, previously frozen samples would appear old in ¹⁴C space but have only exposed to heterotrophic respiration for a fraction of the time that has passed since OC formation. This resulted in the exclusion of permafrost-containing samples from the Colville River basin^{43,47} as well as paleosol samples from Louisiana⁴⁵. Still, we note that inclusion of these samples does not substantively change observed *E* or ¹⁴C trends. In total, our screening procedure removed 56 potential samples from our analysis (mostly

due to OC_{petro} contributions), resulting in a final database of 62 samples (6 DOC, 11 soils, 27
fluvial sediments, and 18 continental margin and coastal marine sediments). Retained samples
are denoted as such in Extended Data Table 1.

532

OC_{petro} sensitivity analysis. We performed a back-of-the-envelope sensitivity analysis to assess 533 the influence of small OC_{petro} contributions (*i.e.*, less than 5 % of total OC) on *E* distributions of 534 535 retained samples. Biospheric OC E was assumed to be Gaussian with $\mu_E = 140$ kJ mol⁻¹ and σ_E $= 15 \text{ kJ mol}^{-1}$, the minimum observed values for non-DOC samples in our dataset. We then 536 included a 5% OC_{petro} contribution as a Gaussian with $\mu_E = 230$ kJ mol⁻¹ and $\sigma_E = 20$ kJ mol⁻¹, 537 538 the maximum observed values for the OC_{petro} sample included in our uncertainty assessment. 539 This approach maximized the potential OC_{petro} influence on E distributions. Resulting μ_E 540 increased by 5 kJ mol⁻¹ and σ_E increases by 9 kJ mol⁻¹ relative to the purely biospheric OC case. However, this influence is likely significantly smaller in reality, especially for σ_E since this 541 542 metric is sensitive to the difference in petrogenic and biospheric mean E values. Assuming a 543 typical sediment with $\mu_E = 170$ kJ mol⁻¹ and $\sigma_E = 25$ kJ mol⁻¹, adding 5 % OC_{petro} as above increased both μ_E and σ_E by only 3 kJ mol⁻¹, well within the methodological uncertainty. 544 We similarly assessed the potential OC_{petro} influence on F¹⁴R results. To do so, we 545 546 assumed that a given sample contains 5 evenly-spaced RPO fractions (i.e., each contains 20% of total OC), that biospheric OC $F^{14}R = 1$ for all fractions, and that OC_{petro} (5% of total OC) is 547 548 entirely contained within the highest-E fraction. Thus, OC_{petro} contributes one quarter of OC in the highest-*E* fraction (= 5% OC_{petro} total / 20% total OC per fraction), leading to $F^{14}R = 0.75$ 549 550 and a reservoir age of $R = 2,300^{-14}$ C yr. These assumptions maximize the potential OC_{petro} influence on F¹⁴R, but still cannot explain observed reservoir ages as high as 13,000 ¹⁴C yr. 551

Nonetheless, inclusion of OC_{petro} does artificially increase μ_E , σ_E , and ¹⁴C-based reservoir ages relative to purely biospheric OC. Because samples retained in our dataset might contain small OC_{petro} contributions (*i.e.*, less than 5% of total OC), these results should be taken as maximum values.

556

557 Statistical analyses. Because scatterplots generally contain uncertainty in both X and Y
558 variables, all linear relationships presented in this study (*i.e.*, <u>Extended Data Fig. 2</u>) were fit
559 using reduced major axis regression⁵⁸.

560

Selective preservation predictions. Finally, we developed a model to predict the evolution of μ_E and σ_E if OC preservation were controlled exclusively by selective preservation. This model predicts $\rho(t,E)$, the mass of OC remaining at time *t* that is associated with activation energy *E*, as a function of *G*(*t*), the total mass of OC remaining at time *t*. To generate the solid lines in Fig. 4, we then determined μ_E and σ_E at each *t* following Eqs. 1-2. Mathematically, this model was derived as follows:

567 The mass of OC remaining at time *t* that is associated with activation energy *E* can be 568 written as:

569
$$\rho(t,E) = \rho(E)e^{-k(E)t}$$
, (8)

570 where $\rho(E)$ is the initial mass of OC associated with *E* and *k*(*E*) is the first-order decay rate for 571 OC associated with *E* (Ref. 22, 59, 60). OC associated with *E* described by lower *k*(*E*) will be 572 selectively preserved relative to OC associated with *E* described by higher *k*(*E*).

573 Here, we make the simplifying assumption that OC associated with the lowest RPO-574 derived E values would decay most rapidly under selective preservation; that is, higher-E OC is 575 preserved for longer. Although high-E OC has been shown to decay on centennial timescales in 576 some environments³⁷, this assumption is supported by laboratory experiments indicating that thermal and microbial decay rates are positively correlated⁶¹. Nonetheless, we emphasize that 577 578 this assumption will not qualitatively influence σ_E evolution (Fig. 4b); selective preservation 579 requires that the distribution of $\rho(t,E)$ must become narrower with time as some OC is 580 preferentially respired. We therefore define k(E) following the Arrhenius equation:

581
$$k(E) = -\omega e^{-E/RT}$$
, (9)

where ω is the Arrhenius pre-exponential factor for microbial decay, *R* is the ideal gas constant, and *T* is temperature (here assumed to be 25 °C). The total mass of OC remaining at time *t* can then be calculated as⁶⁰:

585
$$G(t) = \int_0^\infty \rho(t, E) dE.$$
 (10)

Because this function is not invertible, we cannot substitute G(t) into <u>Eq. 8</u> and calculate $\rho(f^{-1}(G), E)$ directly. Rather, we choose a value for ω , solve <u>Eq. 8</u> and <u>10</u> over some discretized vector **t**, and calculate μ_E and σ_E for each node in **t**. Importantly, our choice of ω has no effect on resulting *E* distributions nor %OC vs. μ_E and %OC vs. σ_E trajectories. Because ω is a scalar multiplier that does not change as a function of *E*, increasing ω results in more rapid overall OC decay but does not influence the distribution of $\rho(f^{-1}(G), E)$ for a given G(t).

592 To generate the *E* distributions shown in Fig. 3, we solved the model assuming that $\rho(E)$ 593 follows a Gaussian distribution with the following initial conditions: OC = 15.5%, $\mu_E = 150.3$ kJ mol⁻¹ and $\sigma_E = 17.7$ kJ mol⁻¹; that is, the median values for all retained samples in our database 594 with $OC \ge 10\%$ (*n* = 6). We again emphasize that our choice of initial distribution shape will not 595 qualitatively influence μ_E and σ_E evolution. To generate the %OC vs. μ_E and %OC vs. σ_E 596 597 trajectories shown in **Fig. 4**, we used the canonically assumed initial OC content for recently 598 produced organic matter of 50%. We then simultaneously determined the initial μ_E and σ_E values 599 that minimize the root mean square error (RMSE) between theory and observations, again 600 assuming that $\rho(E)$ follows a Gaussian distribution. This resulted in best-fit initial values of $\mu_E =$ 111.3 kJ mol⁻¹ and $\sigma_E = 23.3$ kJ mol⁻¹, with an RMSE of 11.2 kJ mol⁻¹. 601

602

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680 Extended Data Legends:

Extended Data Figure 1 | Sample locations. Marine/riverine dissolved OC (black circles), soil
OC (white squares), riverine particulate OC (red triangles), and marine sediment OC (blue
diamonds). See <u>Extended Data Table 1</u> for GPS coordinates, sample collection information,
and original publication references.

685

Extended Data Figure 2 | Intra-sample ¹⁴C age variances. All individual mineral-containing samples (soil, white squares; river POC, red triangles; marine sediment OC, blue diamonds) plotted against %OC reveal a globally coherent power-law relationship. *r* is the reduced major axis correlation coefficient. Age variance is calculated as the sample variance of ¹⁴C ages for all RPO fractions within a given sample (Eq. 4). Age variance and %OC are presented on logarithmic scales. Uncertainty ($\pm 1\sigma$) is smaller than the marker for all data points. The %OC

- axis is reversed to emphasize that OC content generally decreases with time.
- 693

Extended Data Table 1 | Sampling locations, sample collection details, screening procedure results, %OC, Fm, and original reference for all samples.

- 696 Footnotes:
- ^{*}See Methods for screening procedure details
- [†]Samples that originally contain CO₃ but have been acidified prior to RPO analysis are
 marked 'n'

700[‡]Empty cells represent missing or nonexistent data701[§]If no uncertainty is reported, then std. dev. is taken to be the maximum of 0.01 or 1 % of702measured %OC703[¶]If no uncertainty is reported, then std. dev. is taken to be 0.005 Fm units704¶P = Peat; E = Entisol; UB = Udic Boroll; A = Andisol705706706Extended Data Table 2 | μ_E and σ_E values for samples used to determine methodological707uncertainty.