

1 **Carbon stock recovery and greenhouse gas shifts following wetland restoration: a global**  
2 **meta-analysis**

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40 **Abstract**

41 Wetland restoration is widely promoted as a complementary nature-based climate solution, but its  
42 net carbon and GHG effects across wetland types and interventions remain poorly quantified at  
43 the global scale. We address this gap with a global meta-analysis spanning all major wetland types  
44 and restoration strategies. We conducted a global meta-analysis of 617 restored-altered pairs from  
45 149 studies on five continents to assess how restoration influences major carbon stocks and  
46 greenhouse gas (GHG) fluxes relative to altered wetlands. Overall, across all wetland types  
47 studied, restoration significantly increased aboveground biomass, belowground biomass and soil  
48 carbon. Restored wetlands also exhibited significantly lower CO<sub>2</sub> fluxes, consistent with  
49 increased primary production and reduced aerobic decomposition following hydrological and  
50 vegetation recovery, but also higher CH<sub>4</sub> emissions particularly in peatlands where rewetting  
51 enhances anaerobic conditions. By contrast, neither N<sub>2</sub>O flux (though close to) nor, particularly  
52 dissolved organic carbon concentration showed statistically significant overall changes. These  
53 global patterns were robust to sensitivity and leave-one-out analyses but varied strongly among  
54 wetland types and restoration approaches. Restored mangroves and peatlands more clearly  
55 exhibited significant biomass and soil carbon gains, whereas restored freshwater wetlands and  
56 peatlands significantly displayed strong belowground biomass and decreases in both CO<sub>2</sub>  
57 fluxes, though the later showed significant CH<sub>4</sub> flux increases. Other wetland types displayed  
58 more variable responses to restoration, if any. Hydrological restoration (mainly rewetting)  
59 produced the strongest improvements in aboveground biomass, soil carbon and N<sub>2</sub>O flux  
60 reductions, though it significantly increased CH<sub>4</sub> flux. Vegetation recovery was significantly  
61 effective action increasing above- and belowground biomass, soil carbon, and decreasing CO<sub>2</sub>  
62 fluxes. Data were dominated by studies from Asia, Europe and North America, by far made in  
63 the Northern hemisphere, highlighting major geographical gaps in Southern areas. Overall, our  
64 synthesis shows that wetland restoration reliably regenerates carbon stocks and reduces CO<sub>2</sub>  
65 emissions, supporting its inclusion in climate mitigation portfolios and nature-based solution  
66 frameworks. Short-term CH<sub>4</sub> emission increases and ecosystem-specific responses highlight the  
67 need for long-term monitoring, context-dependent restoration design and improved global  
68 coverage to optimise carbon benefits and inform on climate's friendly restoration policies and  
69 actions.

70

71 **Keywords:** Wetland restoration; Carbon sequestration; Greenhouse gas fluxes; Soil carbon;  
72 Methane emissions; Nature-based solutions; Climate change mitigation; Global meta-analysis

73

## 74 **Introduction**

75 Wetlands, defined by their waterlogged soils, hydrophytic vegetation, and distinct biogeochemical  
76 conditions, are among the most carbon-dense ecosystems on Earth (Mitra et al., 2005; Adhikari  
77 et al., 2009; Canadell & Monteiro 2021). These ecosystems play a key role in the global carbon  
78 cycle through their capacity to accumulate organic matter, regulate greenhouse gas (GHG) fluxes  
79 and stabilise hydrological regimes (Mitsch et al., 2013, IPCC, 2023). Despite covering a relatively  
80 small proportion of the planet's surface, wetlands store disproportionately large amounts of  
81 carbon in vegetation and soils, particularly in long-term reservoirs such as peat (Blodau, 2002).  
82 These systems also influence the atmospheric balance of CO<sub>2</sub>, CH<sub>4</sub> and N<sub>2</sub>O, making them critical  
83 components of climate-change mitigation and global biogeochemical dynamics (Mitsch et al.,  
84 2013; Rosentreter et al., 2021). However, decades of drainage, land-use change, pollution and  
85 hydrological modification have resulted in the widespread degradation of wetlands worldwide,  
86 impairing their capacity to store carbon and exacerbating GHG emissions (Beaulieu et al., 2019;  
87 Tan et al., 2020; Bonaglia et al., 2025). This has generated increasing interest in wetland  
88 restoration as a Nature-based Solution (NbS) to support both biodiversity recovery and climate  
89 mitigation goals (Chausson et al., 2020; Bertolini & da Mosto, 2021).

90 Different wetland types contribute in distinct ways to the global carbon cycle, shaped by their  
91 vegetation characteristics, hydrological regimes and soil properties, as well as by its conservation  
92 status (Morant et al, 2020a, 2020b; Camacho-Santamans et al., 2024). Peatlands represent the  
93 largest terrestrial organic carbon reservoir, accumulating carbon over millennia given the slow  
94 anaerobic decomposition of mosses and other peat-forming vegetation (Vasander et al., 2003).  
95 Mangroves sequester large quantities of carbon as both above- and belowground biomass and trap  
96 organic-rich sediments in their dense root systems (Alongi et al., 2012). These ecosystems are  
97 recognised as “blue carbon” hotspots with some of the highest carbon accumulation rates globally  
98 (Alongi, 2020). Saltmarshes and seagrass meadows also contribute to blue carbon stocks,  
99 capturing fine sediments and storing carbon in deep, mineral-rich soils with relatively low CH<sub>4</sub>  
100 emissions due to salinity constraints on methanogenesis (Morant et al., 2020a). Freshwater  
101 marshes and floodplain wetlands, by contrast, support high plant productivity and rapid carbon  
102 turnover in soils, but their saturated conditions can promote CH<sub>4</sub> emissions, making their role in  
103 the carbon cycle highly dynamic (Rosentreter et al., 2021). Shallow lakes and lacustrine wetlands  
104 contribute to carbon cycling mainly via organic sediment accumulation and linkages between  
105 aquatic and terrestrial habitats (Kenney et al., 2010), though their responses to disturbance and  
106 restoration remain understudied. These functional differences underline the need to evaluate  
107 carbon responses to restoration across a wide diversity of wetland types, because the balance  
108 between carbon sequestration and GHG emissions is likely to vary among them.

109 Given the key role of wetlands in natural carbon storage and GHG regulation, their degradation  
110 has major implications for climate change (Tan et al., 2020). Over the past century, wetlands have  
111 experienced widespread degradation driven by land conversion, drainage, hydrological alterations  
112 and other anthropogenic pressures (van Asselen et al., 2013; McCauley et al., 2015; Newton et  
113 al., 2020, Fluet-Chouinard, et al., 2023). For instance, when drained or disturbed, peatlands shift  
114 from long-term carbon sinks to major CO<sub>2</sub> sources due to the oxidation of peat (Waddington &  
115 McNeil, 2002), whereas degradation of inland saline lakes and climate shifts may strongly  
116 enhance CH<sub>4</sub> emissions (Camacho et al., 2017; Morant et al, 2020b; 2024). In response, wetland  
117 restoration has emerged as a key strategy to recover ecosystem functions, enhance biodiversity,  
118 and reinforce climate-related benefits (Zedler, 2000; Erwin, 2009; Griscom et al., 2017). Restored  
119 wetlands are expected to exhibit improved vegetation structure, enhanced primary productivity,  
120 and greater carbon accumulation in soils and biomass (Ferreira et al., 2015; Azman et al., 2021;  
121 Mander et al., 2024). However, restoration may also influence anaerobic microbial processes that  
122 affect GHG emissions (Günther et al., 2020; Taillardat et al., 2020; Rosentreter et al., 2021;  
123 Schuster et al., 2024). In particular, the re-establishment of saturated conditions and organic  
124 matter inputs can lead to short-term increases in CH<sub>4</sub> emissions (Kustina et al., 2025), while the  
125 responses of CO<sub>2</sub>, N<sub>2</sub>O and dissolved organic carbon (DOC) may remain highly variable across  
126 systems (Wickland et al., 2007; Armstrong et al., 2012; Morse et al., 2013; Kluber et al., 2014).  
127 This complexity reflects the strong role of hydrological regimes (Rochera et al., 2025), vegetation  
128 traits (Ge et al., 2024), nutrient availability (Moran et al., 2024) and management practices in  
129 shaping biogeochemical dynamics.

130 Although numerous case studies have examined carbon pools and fluxes in restored wetlands  
131 (Hemes et al., 2018; Xu et al., 2019a; He et al., 2024; Schuster et al., 2024; Kustina et al., 2025),  
132 evidence remains fragmented, localized and often methodologically inconsistent. Such  
133 heterogeneity limits the ability to draw generalizable conclusions about whether restoration  
134 consistently enhances carbon storage, how it alters different GHG fluxes which wetland types  
135 respond most strongly, and which type of restoration actions are more effective for C-storage and  
136 GHG concentrations abatement. Furthermore, the growing policy emphasis on ecological  
137 restoration further underscores the need for robust, quantitative evidence. The European Union  
138 Nature Restoration Regulation (European Union, 2024), adopted as part of the EU Green Deal,  
139 mandates the restoration of degraded ecosystems, including a specific focus on wetlands, to  
140 enhance biodiversity, strengthen resilience and contribute to climate change mitigation. Similarly,  
141 global climate frameworks such as the Paris Agreement, the UN Decade on Ecosystem  
142 Restoration and the expansion of NbS within national climate commitments, identify wetland  
143 restoration as a cornerstone strategy for achieving net-zero emissions and enhancing natural  
144 carbon sinks (Seddon et al., 2020; 2021), though the need to a full decarbonisation of economy

145 cannot be forgotten. Despite the policy momentum, the research focus was mostly on single  
146 wetland types or individual carbon components, which prevents direct comparison of restoration  
147 outcomes across ecosystems, carbon pools and gases. In particular, it remains unclear (i) how  
148 strongly restoration modifies carbon stocks and GHG fluxes relative to degraded conditions, (ii)  
149 whether these effects differ consistently among major wetland types and (iii) how they depend on  
150 restoration strategies. Understanding these large-scale patterns is especially important given the  
151 functional and biogeochemical contrasts among wetland types, which may lead to divergent  
152 responses in biomass accumulation, soil carbon recovery and GHG fluxes.

153 To address these gaps, we conducted a global meta-analysis of key carbon indicators, including  
154 above- and belowground biomass, soil carbon, DOC and the major GHG fluxes (CO<sub>2</sub>, CH<sub>4</sub> and  
155 N<sub>2</sub>O), comparing restored and altered wetlands (drained, converted or otherwise degraded  
156 wetlands) across a wide range of wetland types and restoration strategies. With the integration of  
157 heterogeneous empirical evidence by assessing the direction and significance of restoration  
158 effects, this work aims to (i) quantify the overall effect of restoration on key carbon pools and  
159 GHG-fluxes, (ii) evaluate how they vary with restoration strategies, and (iii) assess how these  
160 effects differ among major wetland types. Based on existing knowledge, we hypothesise that  
161 restored wetlands will show increased carbon accumulation in biomass and soils, reduced CO<sub>2</sub>  
162 emissions due to enhanced primary productivity, and potentially elevated CH<sub>4</sub> emissions linked  
163 to the recovery of anaerobic conditions. This synthesis provides a comprehensive assessment of  
164 carbon outcomes following wetland restoration and offers a foundation for improving the role of  
165 wetland management in climate mitigation initiatives.

166

## 167 **Methods**

### 168 **Literature search**

169 The evidence base for this meta-analysis was assembled following the PRISMA 2020 guidelines  
170 (see Page et al., 2021). We searched for peer reviewed studies that report original field data on the  
171 effects of wetland restoration on carbon storage and GHG pathways. Target variables were soil  
172 (including sediment) organic carbon, aboveground and belowground biomass, DOC, and GHG  
173 fluxes of CO<sub>2</sub>, CH<sub>4</sub> and N<sub>2</sub>O.

174 Searches were run in the Web of Science Core Collection and Scopus in January 2024 and updated  
175 on October 7th, 2024, to include newly published work. In both databases we used a structured  
176 search string that combined three groups of terms:

177 1. Carbon and GHG related terms: carbon stock\*, carbon sequestration, carbon storage,  
178 carbon flux\*, GHG, organic matter\*, greenhouse\*, CO<sub>2</sub>, CH<sub>4</sub>, N<sub>2</sub>O, DOC, DOM,  
179 recalcitrant carbon, soluble organic matter, soluble organic carbon, methane, nitrous  
180 oxide, carbon dioxide, aboveground biomass, belowground biomass;

181 2. Ecosystem terms capturing wetlands: wetland\*, lagoon\*, estuar\*, delta\*, mangrove\*,  
182 peatland\*, seagrass, "sea grass", fen\*, bog\*, swamp\*, marsh, floodplain\*;

183 3. Terms related to restoration and management: restoration\*, rehabilitation\*, revitali\*ation,  
184 renaturali\*ation, management\*.

185 The searches were restricted to publications in English language and including only research  
186 articles and data papers using the database filters on language and document type. The exact  
187 database specific syntax is provided in the **Supplementary Methods**. Records from the two  
188 databases were merged and duplicates removed in R (R 4.3.2; R Core Team, 2023) using the  
189 package *dplyr* (Wickham et al., 2020). After duplicate removal, 13,913 unique records remained.

## 190 **Eligibility and data extraction**

191 The collected studies were screened for eligibility against predefined inclusion and exclusion  
192 criteria (**Supplementary Table S1**). Studies were eligible when they all into the following criteria:

- 193 - the study system is a wetland, including both coastal (mangroves, saltmarshes, seagrass  
194 meadows, brackish wetlands) and inland systems (peatlands, freshwater marshes,  
195 floodplain and riverine wetlands);
- 196 - the work was based on field measurements rather than laboratory or mesocosms  
197 experiments;
- 198 - the design included an explicit comparison between altered (degraded or impacted) and  
199 restored conditions, using Before After (BA), Control Impact (CI), Before After Control  
200 Impact (BACI) or analogous paired designs;
- 201 - the study reported enough information to calculate effect sizes for at least one target  
202 variable (means, some measure of dispersion and sample size for both altered and restored  
203 sites).

204 Studies that only compared restored sites with undisturbed reference wetlands, as well as purely  
205 modelling studies, reviews and meta-analyses, non-wetland ecosystems and cases where neither  
206 variance nor sample size could be obtained or derived, were removed.

207 All records were screened in two stages against the inclusion and exclusion criteria. In the first  
208 stage, titles and abstracts were checked. Records were split among co-authors for reviewing, and  
209 each record was independently assessed by two reviewers. If both agreed that a study did not fulfil

the inclusion criteria, it was excluded at this step; any disagreement or uncertainty led to retention for full text assessment. In the second stage, full texts were read by a different reviewer than in the first stage, again using the same inclusion and exclusion rules. For studies that still qualified, the same co-author extracted the relevant quantitative data as well as descriptive information about the study, the site history and the restoration context. Every extracted dataset was verified by a second contributor, who checked for consistency with the original publication and requested corrections wherever needed. When key statistics were missing, the corresponding authors of the original paper were contacted for clarification if the study had been published within the last 5 years. Older papers for which essential information could not be reconstructed, and papers lacking statistics for which we did not receive feedback from the corresponding author within three months were excluded. Number of studies per stage are included in *Supplementary Figure 1*.

For all 149 eligible publications the data were organised in a structured spreadsheet at the level of pairwise comparisons. A “comparison” was defined as one altered versus one restored site (or experimental unit) for a single parameter. For each comparison we recorded:

- study level metadata (region, wetland type, restoration and alteration type, time since restoration);
- the response variable, its units, mean, dispersion and number of replicates for both altered and restored sites;
- relevant methodological notes (sampling approach, analytical methods, instruments, flux calculation method).

When a study reported several measurements that were not clearly independent (e.g., sediment cores from the same restored marsh or multiple sampling dates), we followed the original authors’ aggregation or, when necessary, calculated a single mean and dispersion per site and parameter following standard formulas. The final dataset thus contained one value per parameter, per altered site and per restored site. The database was extended with 55 additional comparisons generated as part of the RESTORE4Cs project (Cabrera-Brufau et al., 2025). Based on the coordinates of the wetlands reported, each comparison was assigned to a Köppen Geiger climate class using the global maps of Beck et al. (2018). To simplify interpretation, the temperate classes Cfa, Cwa and Csa were merged into a broader “subtropical” category, while retaining the other climate zones as defined in the original classification.

## **Descriptive summaries and mapping**

We mapped all study locations using the reported latitude and longitude of each site and coloured points by wetland type to highlight where different ecosystem types have been studied. In

addition, we summarized geographic coverage with bar plots showing the number of comparisons per region (continent level) and per biogeographical realm. World coastlines and political boundaries were obtained from Natural Earth via the *rnaturrearth* package (South, 2017) and handled as simple features with *sf* (Pebesma, 2018). All maps and other figures (bar plots for wetland type, climate zone, region and biogeographic realm, as well as forest plots and moderator plots) were produced with *ggplot2* (Wickham, 2016).

## Effect size calculations

All analyses and calculations for this study were conducted in R (R 4.3.2; R Core Team, 2023). For each altered versus restored comparison, we calculated the standardized mean difference (SMD, Hedges'  $g$ ; Hedges, 1981) using the *metafor* package (Viechtbauer, 2010). SMD captures the significance and direction of the difference in mean response between restored and altered sites. Positive SMD values indicate that the restored site has a higher mean value than the corresponding altered site, whereas negative values indicate lower values after restoration.

A few studies reported exact zeros in means or standard deviations, which would cause problems when computing standardized effect sizes or coefficients of variation. To keep these comparisons in the dataset, we applied two simple rules. First, when a mean or standard deviation was reported as exactly zero, we added a small positive constant ( $10^{-6}$ ). Second, when the reported standard deviation was extremely small relative to the mean, we imposed a lower bound equal to 5% of the absolute mean. These steps prevent infinite or unrealistically large coefficients of variation and excessively large weights, while leaving the substantive results unchanged (Sweeting et al., 2004; Ren et al., 2021; Nakagawa et al., 2022).

## Meta-analytic models

The meta-analytic models were fitted with “rma.mv” from the *metafor* package, using restricted maximum likelihood (REML). The random effects structure accounted for non-independence of comparisons within publications and within experiments through a nested random-intercept term of the form: “ $\sim 1 | \text{study} / \text{substudy}$ ”; where “study” refers to the publication and “substudy” indexes distinct experimental units within a study (e.g., different restoration sites, distinct restoration actions or experimental blocks that share a common design). This structure prevents studies that contribute many related comparisons from being overweighed while allowing for between-study and within-study heterogeneity.

Intercept-only multilevel models (without moderators) provided pooled estimates of the mean effect of restoration on each parameter. For each parameter, we fitted a separate model and reported the pooled SMD with its 95% confidence interval. For each model, we calculated standard heterogeneity statistics: the test of residual heterogeneity ( $Q$ ), its degrees of freedom and

277 p-value, the proportion of total variability attributable to between-comparison heterogeneity ( $I^2$ ),  
278 computed from Q) and the total between-comparison variance (sum of the estimated variance  
279 components,  $\tau^2$ ).

280 To examine which factors modulate restoration outcomes, we fitted multilevel meta-regressions  
281 separately for each response parameter using the same random-effects structure as the intercept-  
282 only models (random intercepts for study and substudy). For each parameter, we included a  
283 common set of ecological and management moderators (**Supplementary Table S2**): alteration type  
284 (hydrology, morphology, hydromorphology, land-use change, vegetation or habitat loss, water-  
285 quality or pollution, natural disturbance), wetland type (peatland, freshwater wetland, seagrass,  
286 saltmarsh, mangrove, brackish wetland), and restoration type (hydrological, morphological,  
287 hydromorphological, soil-focused, vegetation-focused, passive restoration), together with mean  
288 annual temperature, temperature seasonal variation, annual precipitation, precipitation seasonal  
289 variation, and the time since restoration. Continuous moderators were z-standardised before  
290 fitting and one level of each factor was treated as the reference level.

291 Because not all studies reported the full set of covariates, these models were fitted on the subset  
292 of comparisons with complete metadata (k = 99 for soil carbon, 27 for aboveground biomass, 25  
293 for belowground biomass, 59 for  $\text{CO}_2$ , 44 for  $\text{CH}_4$ , 21 for  $\text{N}_2\text{O}$  and 17 for DOC). For each  
294 parameter, we report the omnibus test for moderators (QM, df, p) from *metafor* and interpret  
295 individual coefficients based on their Wald z-statistics and 95% confidence intervals.

296 For an in-depth analysis and visualization, we fitted separate random effects models within each  
297 wetland type and restoration type whenever more than one comparison was available and plotted  
298 the resulting pooled SMDs and confidence intervals alongside the individual comparisons. When  
299 only a single comparison existed for a given parameter/category combination, we showed the raw  
300 effect size without a pooled estimate.

### 301 **Small study effects, publication bias and robustness**

302 We evaluated potential small study effects and publication bias using standard diagnostics  
303 implemented in *metafor*. Funnel plots were constructed for each parameter from the multilevel  
304 intercept-only models to visualize the relationship between effect size and its precision. Because  
305 Egger-type regression tests are defined for single-level random effects models, we additionally  
306 refitted each parameter using “rma” with a single random effect and used these uni-level models  
307 for formal small-study diagnostics.

308 For parameters with at least 10 comparisons, we applied Egger’s regression test (“regtest”, with  
309 the standard error as predictor) to examine funnel plot asymmetry. In parallel, we computed  
310 Kendall’s rank correlation (“cor.test” with method = ‘kendall’ in the *stats* package) between the

311 effect sizes and their standard errors as a non-parametric check for the same pattern. For  
312 parameters where the pooled effect from the intercept-only model was statistically significant, we  
313 also calculated Rosenthal's fail-safe N (fsn) to gauge how many hypothetical unpublished studies  
314 showing a null effect would be required to attenuate the overall effect to non-significance.

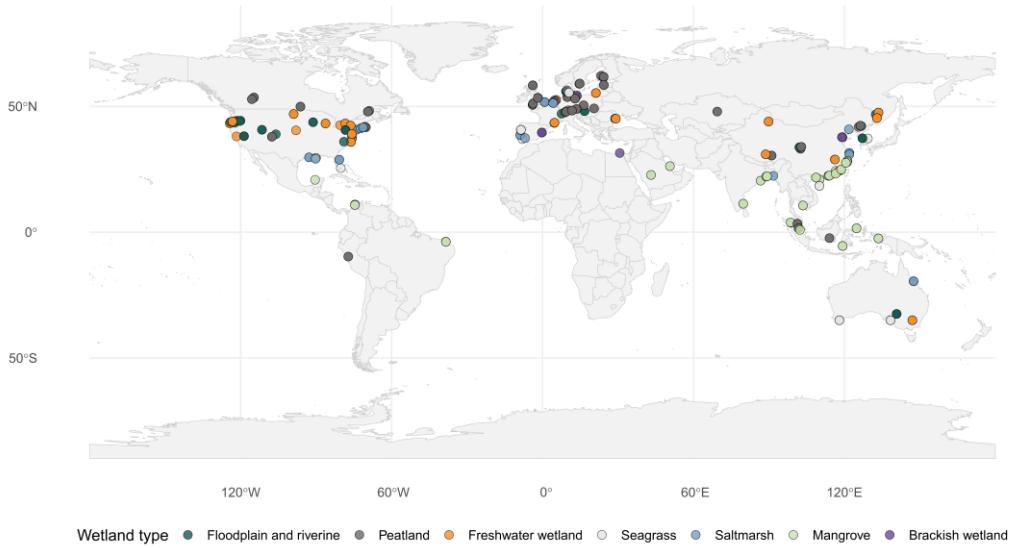
315 To assess robustness to individual studies, we carried out leave-one-out analyses based on uni-  
316 level random effects models. For each parameter, we successively omitted each comparison,  
317 refitted the model and recorded the resulting pooled SMD and its confidence interval. We then  
318 counted how often omitting a single comparison changed the sign or statistical significance of the  
319 pooled effect. Finally, we computed influence diagnostics with *metafor*'s "influence" function  
320 and flagged comparisons as influential when they triggered any of the built-in influence criteria.  
321 At the parameter level, we summarized the number and percentage of comparisons classified as  
322 influential. These diagnostics are reported in the some of the **Supplementary materials** and used  
323 to qualify the interpretation of any apparently strong or highly significant effects.

324

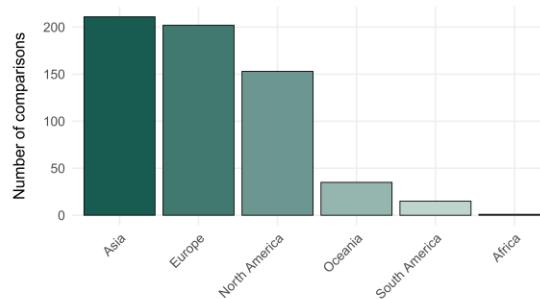
325 **Results**

326 **Distribution of wetlands included in the meta-analysis**

**A. Restoration projects by wetland type**



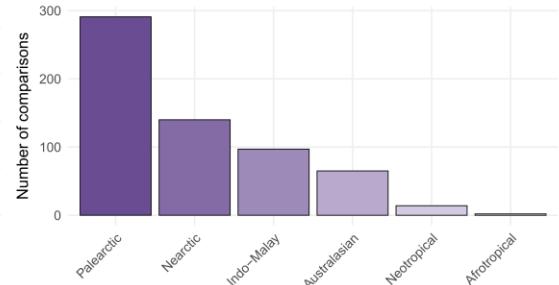
**B. Region**



327

328

**C. Realm**



329 **Figure 1. Global distribution of wetlands included in the meta-analysis (A). Points are color-**  
330 **coded by wetland type. The barplots illustrate geographic coverage across continents (B) and**  
331 **major biogeographic realms (C). Data represent 617 paired comparisons extracted from 149**  
332 **studies.**

333

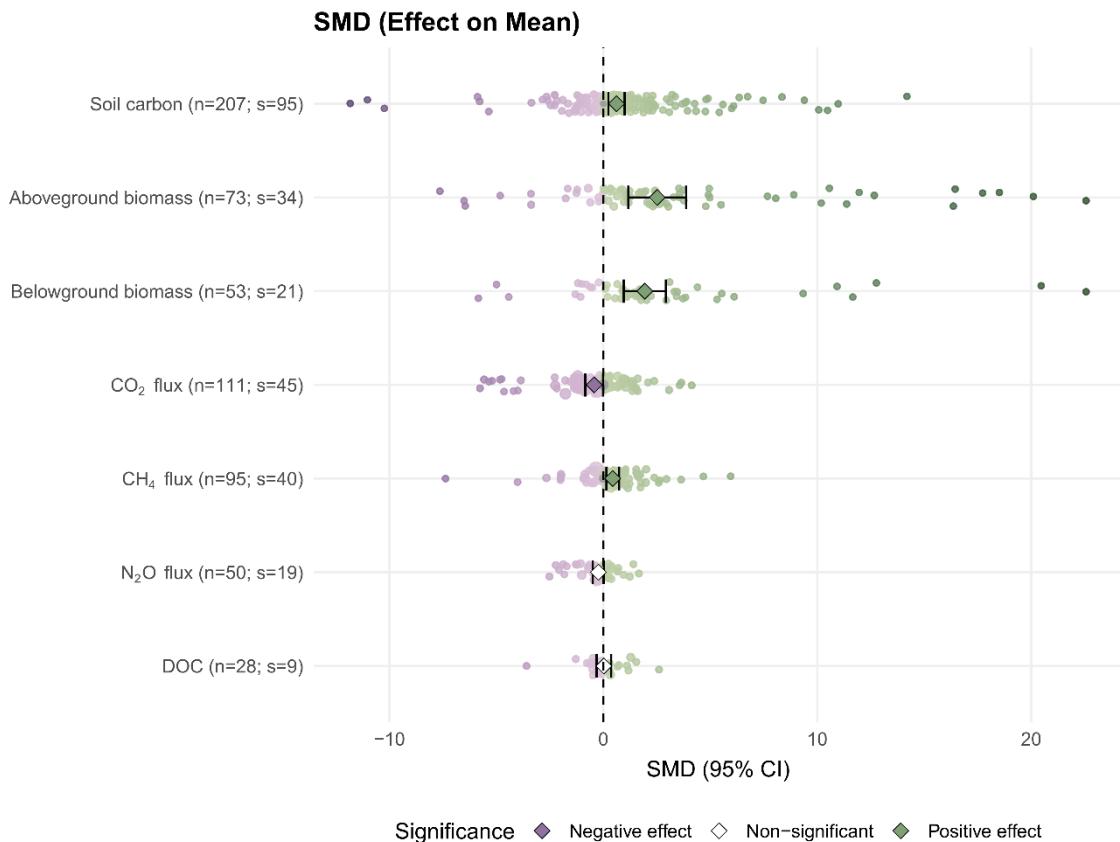
334 A total of 617 paired comparisons were extracted from 149 studies spanning six continents and  
335 all major biogeographic realms (**Figure 1; Supplementary Figure S1**). Most data originated by  
336 far from the Northern Hemisphere (all but a dozen), with Asia (n = 211) and Europe (n = 202),  
337 followed by North America (n = 165). Wetland types were dominated by peatlands (n = 182),  
338 mangroves (n = 134) and freshwater wetlands (n = 118), with fewer cases for saltmarshes (n =  
339 68), floodplain/riverine wetlands (n = 57), seagrass beds (n = 42), and brackish systems (n = 28).  
340 Climate zones were primarily subtropical (n = 209) and temperate (n = 163), with tropical (n =  
341 93), cold (n = 121), and arid (n = 29) regions less represented (**Supplementary Figure 2**).

342

### 343 **Global effects on carbon indicators**

344 Across all wetland types and restoration strategies, SMD indicated significant positive effects of  
345 restoration on biomass and soil carbon accumulations, all of them displaying statistically  
346 significant increases, while contrasting responses were found among GHG fluxes (**Figure 2;**  
347 **Supplementary Table S3**). Aboveground biomass showed the largest effect size (SMD = 2.52;  
348 95% CI: 1.17–3.87; k = 73), followed by belowground biomass (SMD = 1.94; 95% CI: 0.95–  
349 2.92; k = 53). Soil carbon increased significantly (SMD = 0.62; 95% CI: 0.23–0.99; k = 207).  
350 CO<sub>2</sub> flux decrease was statistically significant (SMD = -0.43; 95% CI: -0.84 to -0.01; k = 111),  
351 indicating reduced emissions after restoration. Thus, the responses of all the above parameters  
352 demonstrate a link of wetland restoration to their contribution to climate change mitigations by  
353 increasing the C-sink and GHG-abatement. In contrast, CH<sub>4</sub> flux increased (SMD = 0.44; 95%  
354 CI: 0.14–0.74; k = 95), displaying an opposite pattern. N<sub>2</sub>O flux showed instead a trend though  
355 non statistically significant towards reduction (SMD = -0.23; 95% CI: -0.49 to 0.03; k = 50).  
356 Dissolved organic carbon (DOC) concentration exhibited no significant change neither any visible  
357 trend (SMD = 0.03; 95% CI: -0.31 to 0.36; k = 28). Taken together, these patterns indicate that  
358 restoration strongly rebuilds biomass and soil carbon, reduces CO<sub>2</sub> and (potentially) N<sub>2</sub>O fluxes,  
359 but leaves DOC largely unchanged, while introducing a moderate increase in CH<sub>4</sub> emissions.

360



361

362 **Figure 2. Global standardized mean differences (SMD) for carbon indicators comparing**  
 363 **restored to altered wetlands. Indicators include soil carbon, aboveground biomass,**  
 364 **belowground biomass, CO<sub>2</sub> flux, CH<sub>4</sub> flux, N<sub>2</sub>O flux, and dissolved organic carbon (DOC). The**  
 365 **number of used comparisons are indicated as “n” and the number of studies as “s”. Error bars**  
 366 **denote 95% confidence intervals. Positive values indicate higher values in restored wetlands;**  
 367 **negative values indicate reductions after restoration.**

368

### 369 Heterogeneity and sensitivity

370 Heterogeneity was high for all indicators ( $I^2 > 79\%$ ; **Supplementary Table S4**), with the greatest  
 371 variability observed for CO<sub>2</sub> flux ( $I^2 = 97.3\%$ ) and aboveground biomass ( $I^2 = 93.3\%$ ). Publication  
 372 bias was detected for soil carbon and both biomass indicators (Egger's test  $p < 0.001$ ), while  
 373 heterogeneity tests for GHG fluxes were generally non-significant, except for a weak asymmetry  
 374 signal in CO<sub>2</sub> based on Kendall's tau (**Supplementary Figure S3**; **Supplementary Table S5**).

375 Leave-one-out analyses confirmed robustness of pooled estimates, with no significance flips  
 376 across 617 iterations (**Supplementary Table S6**). Sensitivity analyses indicated no influential  
 377 cases (prop\_influential = 0% for all indicators; **Supplementary Table S7**), this is, there were no  
 378 cases with strongest weighting in the results of the meta-analyses than others. The range of pooled

379 estimates under leave-one-out scenarios was narrow ( $\Delta SMD \leq 0.22$  for aboveground biomass;  $\leq$   
380 0.16 for belowground biomass;  $\leq 0.08$  for DOC;  $\leq 0.05$  for soil carbon; and  $\leq 0.05$  for the three  
381 GHG fluxes included), confirming stability of global trends. In other words, although individual  
382 studies span a wide range of contexts and effect sizes, the direction and significance of the overall  
383 restoration effects are remarkably stable.

384

385 **Moderators analysis**

386 The full moderator set did not improve fit for above- and belowground biomass or  $CH_4$  flux. For  
387 these parameters, the omnibus test for moderators was clearly non-significant (aboveground  
388 biomass:  $Q_M = 4.4$ ,  $df = 13$ ,  $p = 0.99$ ; belowground biomass:  $Q_M = 3.0$ ,  $df = 12$ ,  $p = 1.00$ ;  $CH_4$   
389 flux:  $Q_M = 16.1$ ,  $df = 16$ ,  $p = 0.44$ ; *Supplementary Table S8*), indicating that the multivariable  
390 models did not explain a substantial share of between-comparison variability beyond the overall  
391 mean.

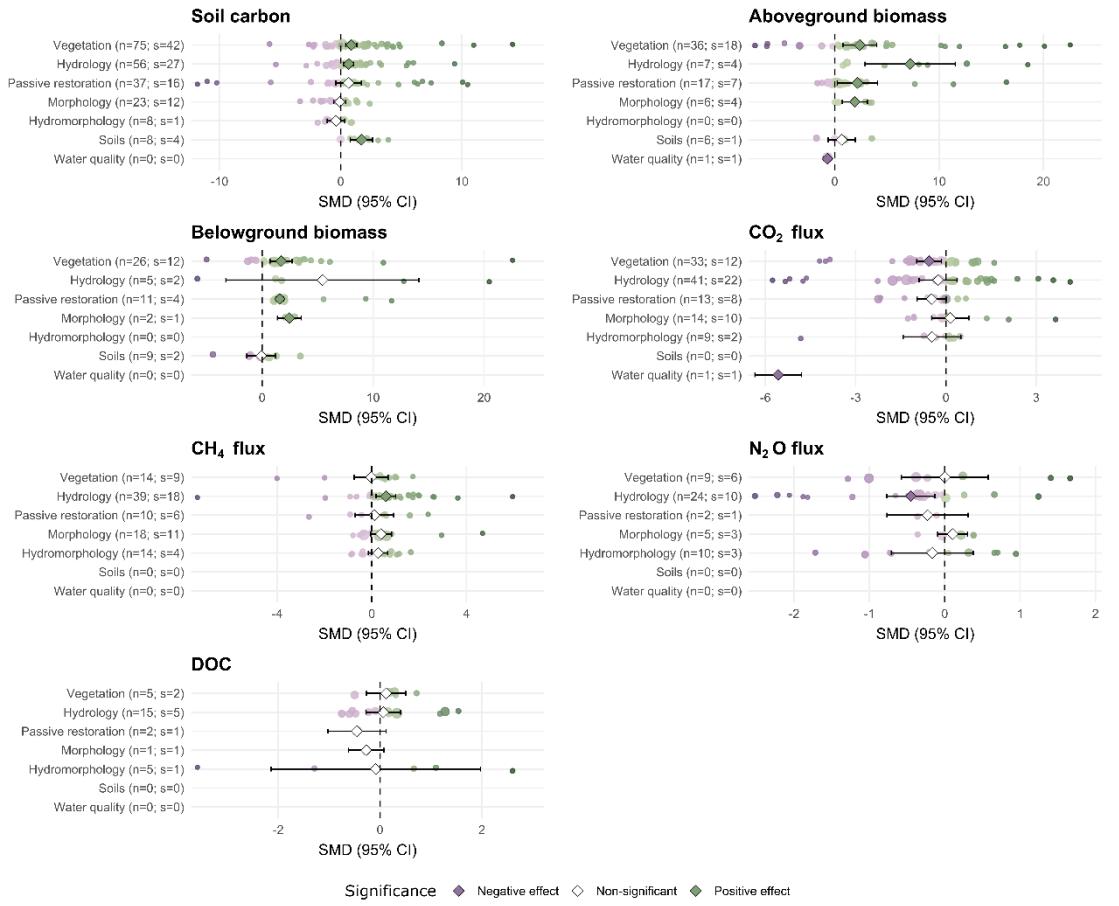
392 By contrast, for  $CO_2$  flux, the full model explained significant heterogeneity ( $Q_M = 56.7$ ,  $df =$   
393 17,  $p < 0.001$ ; *Supplementary Table S8*). Among individual predictors, warmer climates were  
394 associated with stronger  $CO_2$  flux reductions after restoration: the coefficient for standardised  
395 temperature was negative ( $\beta = -3.67$ , 95% CI:  $-6.46$  to  $-0.88$ ,  $p = 0.01$ ), meaning that, per one  
396 standard deviation increase in temperature, the restored–altered contrast in  $CO_2$  flux became more  
397 strongly negative. For soil carbon, the omnibus moderator test was only marginal ( $Q_M = 31.6$ ,  
398  $df = 21$ ,  $p = 0.07$ ; *Supplementary Table S8*), but two climate variables showed consistent positive  
399 effects: both mean annual temperature ( $\beta = 1.63$ , 95% CI:  $0.54$ – $2.73$ ,  $p = 0.003$ ) and temperature  
400 seasonality ( $\beta = 1.43$ , 95% CI:  $0.44$ – $2.43$ ,  $p = 0.005$ ) were associated with larger soil carbon gains  
401 in restored relative to altered wetlands. In other words, within this dataset, restored sites in warmer  
402 and more temperature-seasonal climates tended to show stronger soil carbon recovery.

403 For  $N_2O$  flux and DOC, the moderator blocks were also statistically significant ( $N_2O$ :  $Q_M =$   
404 51.3,  $df = 9$ ,  $p < 0.001$ ; DOC:  $Q_M = 30.0$ ,  $df = 8$ ,  $p < 0.001$ ; *Supplementary Table S8*), but  
405 individual coefficients were generally estimated with wide confidence intervals. The only  $N_2O$   
406 term reaching conventional significance was the time since restoration variable ( $\beta = 0.60$ , 95%  
407 CI:  $0.00$ – $1.20$ ,  $p = 0.05$ ), suggesting that  $N_2O$  responses to restoration change along the  
408 restoration-time gradient.; however, the direction and magnitude of that pattern remain uncertain  
409 given the small sample size ( $k = 21$ ). For DOC, the intercept of the multivariable model was  
410 positive ( $\beta = 3.09$ , 95% CI:  $0.34$ – $5.85$ ,  $p = 0.03$ ), but no single moderator stood out as a clear  
411 driver; the significant omnibus test therefore reflects the combined, diffuse contribution of the  
412 full set of covariates rather than a strong effect of any one factor.

413 Drop-one model comparisons based on AIC supported the idea that restoration time is an  
414 important cross-cutting predictor. Removing the time since restoration variable produced by far  
415 the largest increase in AIC for all parameters ( $\Delta\text{AIC} = 29\text{--}437$  across indicators), whereas  
416 dropping other moderators led to much smaller changes. This pattern indicates that, even when  
417 individual coefficients are imprecise, the restoration-time gradient captures a substantial part of  
418 the residual heterogeneity in the multivariable models. Overall, climate variables and time since  
419 restoration were the most consistent predictors of  $\text{CO}_2$  and soil carbon responses, whereas  
420 variation in biomass and  $\text{CH}_4$  was less systematically explained at the global scale.

421 Effect sizes varied substantially across restoration strategies (**Figure 3**) and wetland types (**Figure**  
422 **4**). For the restoration type (**Figure 3**), vegetation-based and hydrological restoration were the  
423 most influential as significantly changing four parameters each. Vegetation-based restoration  
424 significantly increased both aboveground and belowground biomass, as well as soil carbon,  
425 whereas it significantly decreased  $\text{CO}_2$  fluxes. Hydrological restoration (most generally  
426 rewetting) generated the largest positive effects on aboveground biomass and also increased  
427 significantly soil carbon. Hydrological interventions also produced the most pronounced changes  
428 in GHG dynamics, characterised by significant reductions in  $\text{N}_2\text{O}$  flux (negative SMD) and  
429 elevated  $\text{CH}_4$  flux (positive SMD) in hydrologically restored sites relative to altered sites (**Table**  
430 **I**). Instead, both passive restoration and morphological restoration significantly increased both  
431 aboveground and belowground biomass but had no significant effects on the other parameters.  
432 Actions on soils, much less represented in the meta-analysis, only showed significant increases  
433 of soil carbon, whereas the single study that covered water quality improvements displayed  
434 significant decreases in both parameters tested thereby, namely aboveground biomass and, mostly,  
435 of  $\text{CO}_2$  fluxes. Full statistical support for moderator differences is provided in **Supplementary**  
436 **Tables S8 and S9**.

437



438

439 **Figure 3. Subgroup analysis of restoration strategies.** Forest plots show SMD for each  
 440 indicator grouped by restoration type: hydrological interventions (rewetting), vegetation-  
 441 based restoration, passive restoration, morphological modifications, and soil amendments.  
 442 Error bars represent 95% confidence intervals.

443

444 **Table 1. Overall trends by the different types of restoration actions for the parameters**  
 445 **studied in the meta-analysis. Green arrows mean cooling capacity created by the**  
 446 **restoration-linked changes, either by increasing carbon storage or by GHG abatement,**  
 447 **whereas red arrows mean warming capacity appearing associated to restoration, either by**  
 448 **decreasing carbon storage capacity or by increasing GHG fluxes towards the atmosphere.**

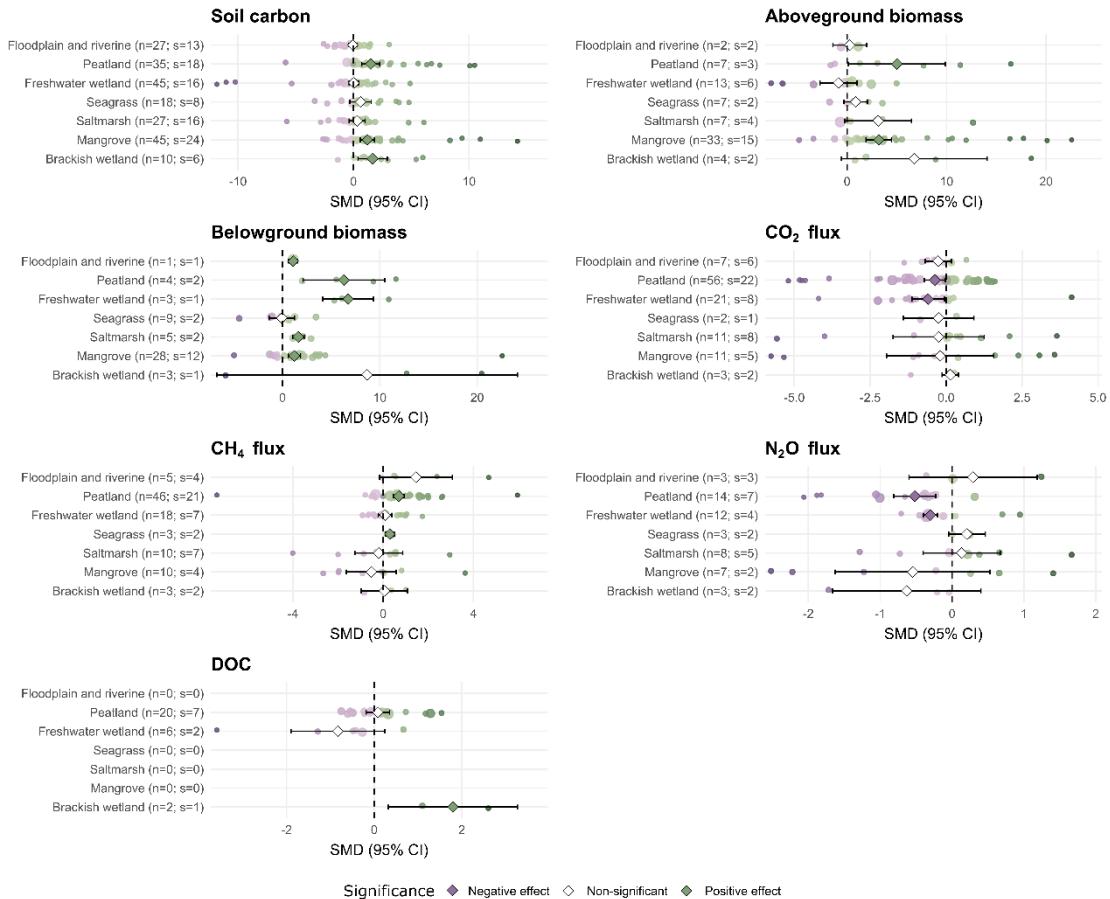
TYPE OF RESTORATION	Aboveground biomass	Belowground biomass	Soil C	DOC	CO <sub>2</sub>	CH <sub>4</sub>	N <sub>2</sub> O
Vegetation	↑	↑	↑		↓		
Hydrology	↑		↑			↑	↓
Passive	↑	↑					
Morphology	↑	↑					
Hydro-morphology							
Soils			↑				
Water quality	↓				↓		

449

450

451 A moderator analysis performed by wetland type revealed substantial variation in restoration  
 452 outcomes among ecosystems (*Figure 4*). Peatland restoration achieved statistically significant  
 453 increases in above and belowground biomass and soil carbon, also promoting significant CO<sub>2</sub> flux  
 454 reductions, though CH<sub>4</sub> fluxes significantly increased reflecting strong anaerobic conditions after  
 455 rewetting. Mangroves also exhibited substantial increases in above- and belowground biomass,  
 456 as well as in soil carbon, though no significant responses were found for GHG fluxes. Freshwater  
 457 wetlands showed significant improvements in aboveground biomass and significant decreases of  
 458 CO<sub>2</sub> fluxes. For the rest of the wetland types, generally much less represented in the dataset,  
 459 seagrasses only displayed significant increases of CH<sub>4</sub> fluxes whereas both saltmarshes and  
 460 floodplain and riverine wetlands only increased significantly belowground biomass, and brackish  
 461 wetlands only increased significantly soil carbon (*Table 2*).

462



463

464 **Figure 4. Subgroup analysis by wetland type. Forest plots display SMD for carbon**

465 indicators across peatlands, mangroves, saltmarshes, freshwater wetlands, seagrass beds,

466 brackish wetlands, and floodplain/riverine systems. Positive values indicate increases in

467 restored sites relative to controls; negative values indicate decreases.

468

469

470 **Table 2. Overall trends of the effects of restoration over different types of wetlands for the**

471 **parameters studied in the meta-analysis. Green arrows mean cooling capacity created by**

472 **the restoration-linked changes, either by increasing carbon storage or by GHG abatement,**

473 **whereas red arrows mean warming capacity appearing associated to restoration by**

474 **increasing GHG fluxes towards the atmosphere.**

WETLAND TYPE	Aboveground biomass	Belowground biomass	Soil C	DOC	CO <sub>2</sub>	CH <sub>4</sub>	N <sub>2</sub> O
Floodplain & Riverine		↑					
Peatlands	↑	↑	↑		↓	↑	
Freshwater wetlands		↑			↓		
Seagrasses						↑	
Saltmarshes		↑					
Mangroves	↑	↑	↑				
Brackish wetlands			↑				

475

476

477

478

479 **Discussion**

480 Our global meta-analysis shows that wetland restoration consistently rebuilds carbon stocks,  
481 especially soil carbon and above- and belowground biomass, and alters GHG fluxes in predictable  
482 ways. Carbon pools respond strongly and robustly, indicating that restoration reactivates key  
483 processes that drive long-term carbon accumulation. CO<sub>2</sub> emissions decline soon after restoration,  
484 whereas CH<sub>4</sub> emissions typically rise in the short-term, with emerging evidence that these  
485 increases weaken as ecosystems mature (Kayranli et al., 2010; Mitsch et al., 2013; He et al.,  
486 2024;).

487 **Restoration enhances carbon stocks through improved hydrology and morphology,  
488 vegetation recovery and passive processes**

489 The consistent increases in above- and belowground biomass in restored wetlands, especially in  
490 mangroves, reflect the re-establishment of hydrological regimes and vegetation communities that  
491 support high primary productivity (Callaway et al., 2003; Kayranli et al., 2010; Azman et al.,  
492 2021; He et al., 2024). All actions related to the restoration of the vegetation cover resulted in  
493 statistically significant increases of the biomass and soil carbon variables, as vegetation is the  
494 main component incorporating the biomass as plant components and soil carbon as producing the  
495 organic matter, highly recalcitrant, that is accumulated in wetland soils. Active vegetation  
496 restoration can further accelerate canopy development in some systems (Callaway et al., 2003;  
497 Ferreira et al., 2015). Hydrological restoration, mostly rewetting, relieves the water physiological  
498 stress, facilitates colonisation by native wetland plant species and enhances nutrient cycling,  
499 which overall drives rapid biomass accumulation (Waddington et al., 2010; Geurts et al., 2020).  
500 Passive restoration, where natural hydrological and ecological processes are allowed to recover  
501 after removal or reduction of disturbances, together with morphological restoration, were also  
502 particularly effective for increasing biomass and soil carbon. When barriers for plant colonization  
503 are removed, natural regeneration rebuilds plant cover and diversity (De Steven et al., 2010), and  
504 re-establishes soil feedback, successional dynamics and soil carbon inputs (An et al., 2021).

505 Soil carbon gains are closely tied to the return of saturated conditions, which suppresses (the  
506 faster) aerobic decomposition and promotes the accumulation of particulate and dissolved carbon,  
507 with permanently flooded wetlands often storing the largest soil carbon stocks (Yin et al., 2019).  
508 In peatlands, raising water tables slows peat oxidation and enables peat-forming vegetation to  
509 expand (Vasander et al., 2003), whereas in mangroves, hydrological rehabilitation enhances  
510 sediment trapping and dense root growth (Twilley & Rivera-Monroy, 2005). These processes  
511 explain the strong and consistent recovery of biomass accumulation and soil carbon across  
512 restored sites, even though some studies suggest that restored wetlands may not fully reach the  
513 carbon levels of intact reference sites (Xu et al., 2019a).

514 Even if some publication bias was detected for soil carbon and both biomass indicators, there was  
515 no overall differential weighting of any of the data sources. Further, leave-one-out analyses  
516 narrow scenarios confirmed the robustness of pooled estimates. These confirm that the patterns  
517 demonstrated in the meta-analysis concerning biomass and soil carbon accumulation are  
518 consistent and represent stable global trends in the direction and significance of the overall  
519 restoration effects on these parameters.

520 **Divergent GHG dynamics: immediate reductions in CO<sub>2</sub> but short-term increases in CH<sub>4</sub>**

521 The clear biomass and carbon stock improvements contrast with the more modest GHG flux  
522 responses achieving statistically significant significance. Though, both the fact that heterogeneity  
523 tests for GHG fluxes were generally non-significant, as well as the narrow range of the estimates  
524 under leave-one-out scenarios, support the strength of the meta-analysis results for GHG fluxes  
525 and show that the obtained patterns are robust.

526 CO<sub>2</sub> emissions drop as linked to restoration (Figure 2), which is consistent with most studies  
527 stating that CO<sub>2</sub> fluxes decreased almost immediately after restoration (e.g. Darusman et al., 2023;  
528 Schuster et al., 2024). This is driven by reductions in aerobic decomposition (Crase et al., 2013;  
529 Valach et al., 2021) and, principally, by increased plant CO<sub>2</sub> uptake (Waddington et al., 2010),  
530 where the meta-analysis clearly identified the direct vegetation restoration as the main actions  
531 achieving statistically significant CO<sub>2</sub> flux reductions.

532 In contrast, CH<sub>4</sub> emissions increased significantly in restored wetlands, reflecting the re-  
533 establishment of anoxic conditions that stimulate methanogenesis (Urbanová et al., 2011; Hemes  
534 et al., 2018). Rewetting increases soil saturation and organic substrate availability, creating  
535 favourable conditions for CH<sub>4</sub> production (Rosentreter et al., 2021; Darusman et al., 2023; Cui et  
536 al., 2024). According to our meta-analysis, this short-term CH<sub>4</sub> pulse is consistent across wetland  
537 types and restoration strategies. However, long-term datasets suggest that CH<sub>4</sub> emissions may  
538 decline as vegetation communities stabilise, root oxygenation increases and microbial processes  
539 tend to adapt to new equilibrium conditions (Mitsch et al., 2013; Delwiche et al., 2025). Restored  
540 wetlands require several decades to develop carbon sequestration capacities comparable to those  
541 of natural systems (Kayranli et al., 2010), which may result in the maintenance of a certain  
542 warming potential of the restored wetlands in the short-term (in some cases, decades), though  
543 overall restoring wetlands are crucial for long-term cooling ability (Taillardat et al., 2020). In  
544 fact, CH<sub>4</sub> emissions in wetlands can peak initially but tend to stabilize after ~10 years (He et al.,  
545 2024). In peatlands, for instance, restoration via rewetting is effective despite initial CH<sub>4</sub> spikes  
546 (Günther et al., 2020; Kustina et al., 2025), thus considering the rewetting a viable approach for  
547 long-term carbon sequestration (Mander et al., 2024). However, it must be recognized that the  
548 support for our interpretation does not come from the meta-analysis itself, but instead uses specific

549 results from particular studies, as our CH<sub>4</sub> moderator analysis was not able to find significant  
550 effect of time-since restoration with our dataset. The limited duration of most post-restoration  
551 studies prevents firm conclusions referring the role of aging, highlighting the need for long-term  
552 monitoring to better characterise CH<sub>4</sub> trajectories as wetlands mature after restoration. This is not  
553 trivial, since various studies (Mistch et al., 2013; Neubauer et al., 2014; Schuster et al, 2024) show  
554 controversies about the time frame after which the effects of restoration can transform an altered  
555 wetland from a source of heating to a cooling capacity

556 N<sub>2</sub>O and DOC responses were not statistically significant overall, implying that these indicators  
557 are more strongly governed by local nutrient availability, soil chemistry and hydrological  
558 variability than by restoration status alone, though also size effects and other aspects can explain  
559 the results as both parameters behave very differently. For N<sub>2</sub>O, the number of related studies  
560 provided less than half of pairwise comparison than for CO<sub>2</sub> and CH<sub>4</sub>, respectively. Given that  
561 nearly significant reductions of N<sub>2</sub>O fluxes were shown in the meta-analysis, the trend relating  
562 restoration to N<sub>2</sub>O, reductions could be considered as a likely possibility, which could be  
563 associated to the induced changes in the biogeochemistry and complexity of nitrogen cycling  
564 pathways, which depend on the balance between nitrification and denitrification, and the  
565 availability of inorganic nitrogen substrates (White and Reddy, 2009). Many restored wetlands  
566 experience rapid changes in oxygen availability and water table height, but N<sub>2</sub>O production is  
567 often constrained by low nitrate concentrations, reducing conditions that favour complete  
568 denitrification to N<sub>2</sub> rather than N<sub>2</sub>O (Kluber et al., 2014; Kasak et al. 2021). However, the context  
569 dependency of nitrogen dynamics, driven by factors such as legacy nutrient loads, external inputs  
570 from agriculture, or tidal flushing in coastal wetlands, can partly override the influence of the  
571 restoration, affecting in a different way the soil microbial communities and processes (Morse et  
572 al., 2013; Kluber et al., 2014), thus masking a possible significant improvement by reducing N<sub>2</sub>O  
573 fluxes as a consequence of restoration (Bianchi et al., 2021).

574 Contrarily, no trend could be observed on the DOC responses to restoration, which likely reflects  
575 the interplay between hydrology, vegetation type, organic matter quality and bacterial community  
576 activity (Evans et al., 2005; Mladenov et al., 2005; Armstrong et al., 2012; Strack et al., 2015),  
577 not only in the wetland itself but also in its catchment. DOC dynamics may reflect transient  
578 flushing of dissolved organics, with longer-term patterns depending on plant community  
579 composition, redox oscillations and soil sorption capacity (Wickland et al., 2007). The net effect  
580 across systems is therefore highly variable, so it is not surprising the lack of consistent global  
581 trend in DOC change following restoration.

582 **Restoration strategies affect carbon stocks and GHGs differently**

583 The meta-analysis results highlight the differentiated roles of restoration strategies for carbon  
584 outcomes (**Figure 3** and **Table 1**). Vegetation-based restoration demonstrated the most effective  
585 mitigation effects, as being effective in increasing carbon capture either directly as biomass or as  
586 increasing carbon storage in wetland's soils but also reducing CO<sub>2</sub> fluxes. This demonstrates that  
587 the recolonization by natural vegetation, mainly helophytes, should be a main target for ecological  
588 restoration with a climatic perspective as recovering the role of wetlands as carbon sinks (Morant  
589 et al., 2020, 2024; Camacho-Santamans et al., 2024) reflecting their capacity to restore organic  
590 matter inputs, rhizosphere processes and successional vegetation dynamics that drive long-term  
591 carbon accumulation (Callaway et al., 2003; Suir et al., 2019). Both morphological and passive  
592 restoration also increased biomass accumulation, mainly by increasing vegetation, mainly by  
593 favouring ecological scenarios that allow plant primary production processes to increase (e.g.  
594 higher habitat availability for macrophytes by increasing the rooting surface or by reducing  
595 turbidity and thus the shading effect). Similarly, for similar reasons, hydrological restoration,  
596 mostly rewetting, also favours increases in aboveground biomass and soil carbon accumulation,  
597 though after rewetting of dried lands (e.g. land reclamation from agricultural uses), these restored  
598 wetlands can trigger short-term CH<sub>4</sub> emission spikes (Darusman et al., 2023; Kustina et al. 2025).  
599 Remarkably, N<sub>2</sub>O fluxes are also reduced after rewetting, which was previously reported by Kasak  
600 et al (2021) for rewetted agricultural lands. Though certainly difficult to be attributed to specific  
601 reasons because of the complexity of the nitrogen cycle, given that most hydrological restorations  
602 collected in the papers are associated to rewetting of old agricultural lands, this can be related to  
603 the decrease of nitrate availability when crop fertilization stops, then nitrates remaining in soils  
604 could be quickly removed by runoff and the limited nitrate availability for denitrification in the  
605 restored wetland could be one of the causes of the N<sub>2</sub>O fluxes reduction pattern shown by the  
606 meta-analysis after hydrological restoration. Overall, these distinct effects show that restoring  
607 vegetation is crucial for long-term carbon storage, while hydrological recovery mainly shapes  
608 GHG fluxes.

#### 609 **Wetland types vary in their contribution to carbon outcomes**

610 Different wetland types contributed differently to carbon responses (**Figure 4** and **Table 2**).  
611 Mangroves and peatlands, the most studied wetland types (Schuster et al., 2024), dominated the  
612 dataset and showed some of the strongest positive responses in carbon stocks. Restored  
613 mangroves exhibit substantial increases in above- and belowground biomass, attributable to rapid  
614 tree growth and dense root networks (Alongi, 2012), but also contribute decisively to soil carbon  
615 storage by favouring sedimentation as acting as sediment traps. Peatlands showed pronounced  
616 soil carbon increases when restored, due to halted peat oxidation and resumed peat accumulation  
617 following rewetting (Günther et al., 2020), though they also achieve significant CO<sub>2</sub> fluxes  
618 reduction as reactivating peat accumulation, though partially counteracted by the above

619 mentioned CH<sub>4</sub> increases linked to rewetting inducing lower redox conditions and thus favouring  
620 methanogenesis given their waterlogged soils, high organic content and the relative potential of  
621 its methanogenic microbial communities (Torres-Alvarado et al., 2005; Liu et al., 2011;  
622 Rosentreter et al., 2021), especially in wetland types where high sulphate concentrations does not  
623 impede methanogenesis activation, (Koebsch et al., 2019, Lin,et al., 2024; Morant et al, 2024,  
624 Miralles-Lorenzo et al, 2025). But even in sulphate-rich environments, such as seagrass beds,  
625 methanogenesis is possible provided that methylotrophic archaea are responsible for this activity  
626 (Schorn et al., 2022), which would explain the CH<sub>4</sub> increases in these ecosystems after restoration  
627 as significantly demonstrated in our meta-analysis (**Table 2**),

628 On the other hand, freshwater wetlands, floodplains, and saltmarshes, when restored, were found  
629 to significantly increase belowground biomass, but the former also demonstrated significant  
630 capacity to reduce CO<sub>2</sub> fluxes. Brackish wetlands and seagrass beds were underrepresented,  
631 which limits generalisation for these systems.

### 632 **Geographic and ecosystem biases constrain global inference**

633 The geographical distribution of available studies is strongly biased towards the Northern  
634 hemisphere, North America, Europe and Asia, accounted for most of the studied cases, with  
635 limited representation from South America, Africa and Oceania, despite their wetlands'  
636 importance (Xu et al., 2019b). On global scale this imbalance is critical because underrepresented  
637 regions contain vast carbon-rich wetlands, such as Amazonian floodplains, Andean peatlands and  
638 the Congo Basin, that may exhibit different restoration trajectories and carbon dynamics. Without  
639 expanding research in the Global South, global restoration policies risk overlooking major carbon  
640 reservoirs and misestimating restoration benefits. Improving monitoring capacity and data  
641 collection in these regions is therefore essential for accurate global carbon assessments.

### 642 **Management Implications and Policy Relevance**

643 The global patterns identified here have direct implications for restoration planning and climate  
644 policy. Consistent increases in soil carbon and above- and belowground biomass across restored  
645 wetlands affirm restoration as an effective nature-based climate solution for carbon storage,  
646 supporting its inclusion in frameworks such as the UN Decade on Ecosystem Restoration, and the  
647 newly adopted EU Nature Restoration Regulation (European Commission, 2024) which  
648 specifically mandates the recovery of degraded wetlands to enhance biodiversity and strengthen  
649 climate mitigation. However, some authors claimed that wetland restoration would not fit the time  
650 frames of the climate agreements, such as the Paris Agreement (Schuster et al., 2024), thus  
651 highlighting the preference of wetlands conservation on a good health status over wetland  
652 restoration (Morant et al., 2020b). In fact, the “so-called” recovery-debt (Moreno-Mateos et al.,  
653 2017) linked to functional and structural losses of restored ecosystems (Moreno-Mateos et al.,

654 2012) may seriously compromise the climatic gains of wetland restoration. Actually, the climatic  
655 effects of restoration are very much dependent on the effects of specific restoration actions on  
656 specific types of wetlands (Meli et al., 2014; Morant et al., 2020b; Camacho-Santamans et al.,  
657 2024, Jones et al., 2024). This is mediated by the biogeochemical changes induced by each  
658 specific restoration action and their interactions, which further can be not only wetland-type  
659 specific (Reddy et al., 2022), but also context specific. For instance, short term increases in CH<sub>4</sub>  
660 emissions could be interpreted in the context of long-term net carbon benefits: although CH<sub>4</sub>  
661 would rise initially, restoration may rapidly reduce CO<sub>2</sub> emissions and enhance carbon storage,  
662 yielding strong climate gains over time (Taillardat et al., 2020; Mander et al., 2024; Delwiche et  
663 al., 2025). In this example, the selection of restoration strategies and their *ad hoc* design can help  
664 minimising CH<sub>4</sub> pulses by avoiding abrupt oversaturation and promoting vegetation assemblages  
665 that moderate anaerobic conditions. Policymakers and wetland managers and restoration  
666 practitioners should therefore avoid uniform prescriptions and instead tailor restoration  
667 approaches to site specific carbon and GHG goals and local biophysical constraints.

668 Climate change is expected to profoundly alter carbon dynamics in wetland ecosystems,  
669 potentially undermining their role as long term carbon sinks (Wang et al., 2018). These findings  
670 highlight the vulnerability of peat-rich systems to climatic variability and underscore the need for  
671 conservation and restoration strategies and policies that integrate climate projections. Therefore,  
672 as demonstrated by this meta-analysis proactive management should prioritise adaptive measures,  
673 such as hydrological stabilisation and vegetation recovery, while considering regional climate  
674 trajectories to safeguard the sink function of wetlands in the coming decades.

675 Ecosystem specific responses should also guide decision making. Mangroves offer particularly  
676 high returns in vegetation biomass, peatlands provide substantial soil carbon benefits and  
677 reductions in CO<sub>2</sub> emissions, whereas some wetland types require targeted strategies for  
678 managing CH<sub>4</sub> fluxes. Funding mechanisms and national climate plans should align investments  
679 with these strengths, prioritising restoration where carbon gains are highest and tailoring GHG  
680 fluxes, particularly CH<sub>4</sub> management where risks are greatest.

681 Finally, addressing geographic research gaps is essential for equitable and effective global climate  
682 mitigation. Expanding conservation and restoration sciences and monitoring in South America,  
683 Africa and Oceania will improve carbon accounting, strengthen restoration outcomes and ensure  
684 that the world's most carbon rich wetlands contribute fully to global climate targets, also  
685 providing economic credits allowing the Global South countries to maintain these world-benefits  
686 under an ecological transition with justice

687

688 **Management and Policy Recommendations**

689 Our findings support several evidence-based recommendations for practitioners, policymakers  
690 and restoration planners:

- 691 1. Conserve before needing to restore. The targets of restoration are recovering the natural  
692 behaviour of the restored ecosystem. By avoiding wetland alteration, the natural role of  
693 wetlands as carbon sinks and (for most wetland types) also GHG mitigation ecosystem  
694 can be maintained.
- 695 2. Prioritise hydrological restoration where climate mitigation is a key objective. Rewetting  
696 produces the clearest and most consistent gains in soil carbon and reduces CO<sub>2</sub> emissions  
697 across wetland types.
- 698 3. Promote vegetation recovery to maximise biomass carbon stocks, especially in peatlands,  
699 mangroves, saltmarshes and freshwater marshes. Passive recovery can be effective, but  
700 active habitat recovery and planting may accelerate carbon accumulation in high-  
701 productivity ecosystems.
- 702 4. Account for short-term CH<sub>4</sub> increases following rewetting, especially in peatlands and  
703 freshwater wetlands, where salinity restrictions would not control the increase in CH<sub>4</sub>  
704 fluxes. Monitoring frameworks and carbon crediting schemes should incorporate  
705 temporal dynamics rather than relying solely on snapshot measurements.
- 706 5. Design restoration strategies to match ecosystem-specific responses. Mangroves and  
707 saltmarshes are high-priority systems for biomass-based mitigation, whereas peatlands  
708 and floodplain wetlands offer substantial soil carbon gains but require careful CH<sub>4</sub>  
709 management and monitoring.
- 710 6. Implement integrated Nature-based Solutions that combine restoration with hydrological  
711 reconnection, sediment management and vegetation rehabilitation, aligning with the EU  
712 Nature Restoration Regulation, the UN Decade on Ecosystem Restoration, and  
713 worldwide and national climate strategies.
- 714 7. Expand long-term monitoring in understudied regions, particularly tropical and  
715 subtropical wetlands outside Asia, to address gaps in global carbon budgets and enhance  
716 the effectiveness of restoration investments.

717 Overall, wetland restoration emerges as a potentially effective approach, though just  
718 complementary to the decarbonisation of the economy, climate mitigation tool, with strong  
719 potential to regenerate carbon stocks, reduce CO<sub>2</sub> emissions and support biodiversity recovery.  
720 To fully realise these benefits, restoration must be implemented at scale, tailored to ecosystem-  
721 specific carbon dynamics and embedded within national and international policy frameworks.

722 Our synthesis provides a quantitative foundation for optimising wetland restoration as a  
723 complementary climate solution and for guiding future research toward the most critical  
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## 751 **Author Contributions**

752 AC, KA, BM, DM, conceived the study. BM, DM, AC, coordinated data compilation and  
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755 analyses. AC, BM, DM, interpreted the results. AC, DM, BM, drafted the manuscript. AC  
756 wrote the final version of the manuscript. All authors contributed to manuscript revision  
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1079 **Supplementary material**

1080 **Supplementary Methods:**

1081 **WOS:**

1082 TS=(( "carbon stock\*" OR "carbon sequestration\*" OR "carbon storage" OR "carbon flux\*" OR  
1083 GHG\* OR "organic matter\*" OR greenhouse\* OR CO2 OR CH4 OR N2O OR DOC OR TOC  
1084 OR DOM OR "recalcitrant carbon" OR "soluble organic matter" OR "soluble organic carbon" OR  
1085 methane OR "nitrous oxide" OR "carbon dioxide" OR "aboveground biomass" OR "belowground  
1086 biomass") AND (wetland\* OR lagoon\* OR estuar\* OR delta\* OR mangrove\* OR peatland\* OR  
1087 seagrass OR "sea grass" OR fen\* OR bog\* OR swamp\* OR *marsh* OR floodplain\*) AND  
1088 (restoration\* OR rehabilitation\* OR revitalization OR renaturalization OR management\*)) AND  
1089 LA=(English) AND DT=(Article OR Data Paper)

1090 **SCOPUS:**

1091 TITLE-ABS-KEY ( ( "carbon stock\*" OR "carbon sequestration\*" OR "carbon storage" OR  
1092 "carbon flux\*" OR GHG\* OR "organic matter\*" OR greenhouse\* OR CO2 OR CH4 OR N2O  
1093 OR DOC OR TOC OR DOM OR "recalcitrant carbon" OR "soluble organic matter" OR "soluble  
1094 organic carbon" OR methane OR "nitrous oxide" OR "carbon dioxide" OR "aboveground  
1095 biomass" OR "belowground biomass") AND ( wetland\* OR lagoon\* OR estuar\* OR delta\* OR  
1096 mangrove\* OR peatland\* OR seagrass OR "sea grass" OR fen\* OR bog\* OR swamp\* OR *marsh*  
1097 OR floodplain\* ) AND ( restoration\* OR rehabilitation\* OR revitalization OR renaturalization OR  
1098 management\* ) ) AND LANGUAGE ( english )

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1103 *Supplementary Table S1. Criteria for inclusion and exclusion during the review process.*

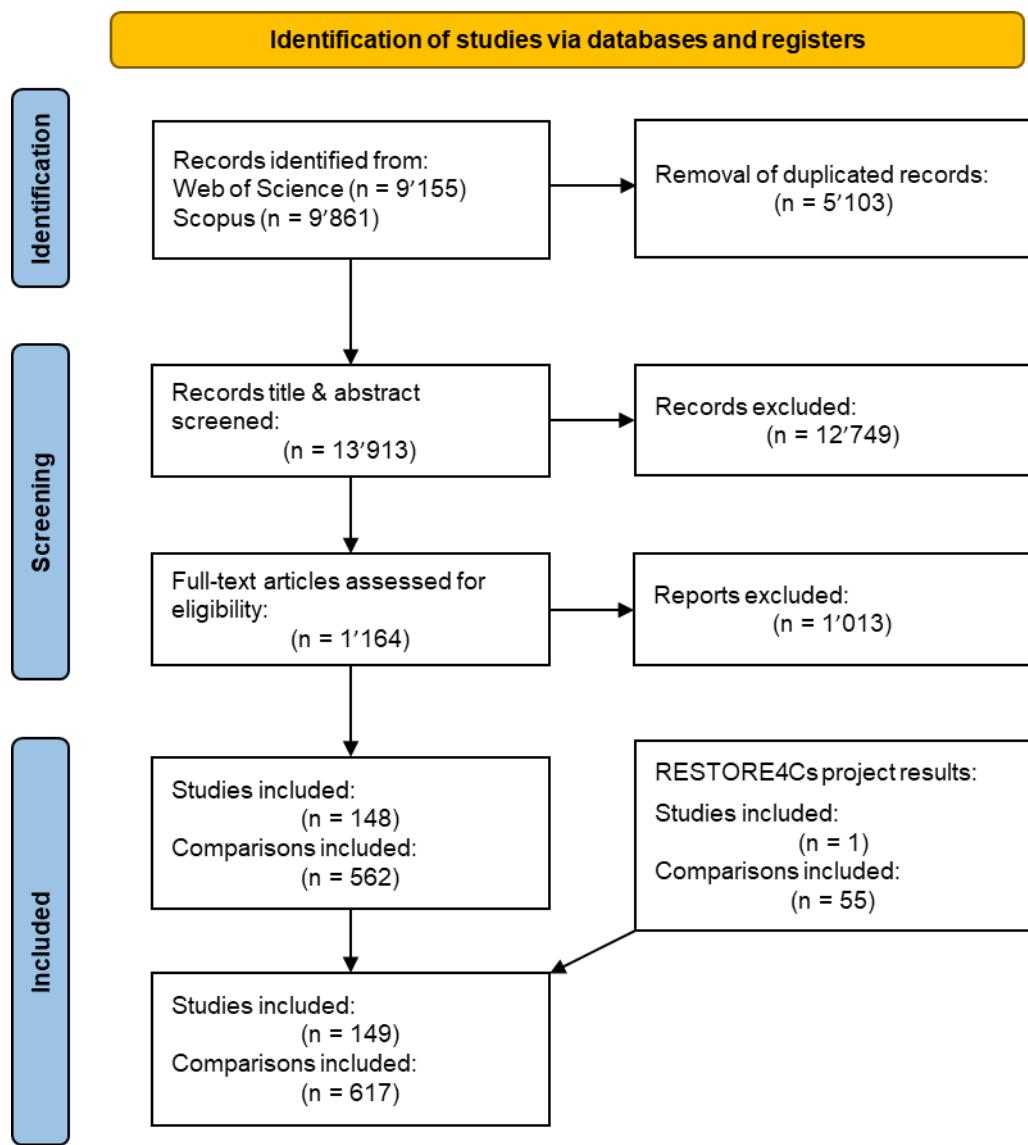
Criteria	Include	Exclude
Source	Web of Science Core Collection and Scopus	Other data sources
Publication type	Research articles or data papers reporting original field measurements	Reviews, meta-analyses, modeling-only studies, extrapolations without field data, laboratory or mesocosm studies
Language	English	All other languages
Ecosystem	Sites historically classified as wetlands	Terrestrial ecosystems, other types of freshwater systems, marine ecosystems
Geography	Studies conducted across the globe	-
Restoration action	Any active or passive restoration or management action that aims to recover coastal wetland structure or function	Monitoring of only natural or only impacted sites with no restored control sites
Variables	Quantified carbon pools (soil or sediment, aboveground or belowground biomass), DOC or GHG fluxes (CO <sub>2</sub> , CH <sub>4</sub> , N <sub>2</sub> O)	Studies that do not report any of these variables
Design	Direct comparisons of altered (degraded, impacted) versus restored sites; BA (Before-After), CI (Control-Impact), or BACI (Before-After-Control-Impact) comparisons	Studies reporting only restored versus natural reference comparisons, or restored sites only
Data provision	Means, SDs, and sample sizes for each group, or convertible statistics allowing SD derivation (SE, CI, IQR, range)	Statistics insufficient to derive SD or unclear sample sizes even after author contact (for studies published in the past 5 years)

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1110 *Supplementary Figure S1. Number of studies included in each stage during the review process.*

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1112 *Supplementary Table S2. Definitions of moderator classes used for grouping studies in the*  
1113 *meta-analysis*

Moderator Category	Class	Definition
Alteration Type	<b>Hydrology</b>	Alterations involving changes to water level, hydroperiod, surface–groundwater connectivity, tidal regime, drainage, or flooding patterns.
	<b>Hydromorphology</b>	Combined hydrological and morphological disturbances, such as channelization, bank stabilization, or modifications affecting both flow regime and geomorphic structure.
	<b>Land-use Change</b>	Conversion of wetlands to agricultural, urban, industrial or other terrestrial uses, including grazing, cropping, infrastructure development or forestry.
	<b>Morphology</b>	Physical reconfiguration of wetland shape or structure, such as excavation, infilling, dredging, shoreline modification, or changes to microtopography.
	<b>Vegetation</b> or <b>Habitat Loss</b>	Removal, degradation or fragmentation of native wetland vegetation, through harvesting, overgrazing, invasive species encroachment, or other processes reducing plant biomass or cover.
	<b>Water-quality</b> or <b>Pollution</b>	Chemical degradation through nutrient enrichment, eutrophication, salinisation, heavy metals, organic pollutants or acidification.
	<b>Natural Disturbance</b>	Alteration resulting from naturally occurring events such as storms, wildfires, droughts or floods.
Wetland Type	<b>Peatland</b>	Terrestrial wetland ecosystems in which waterlogged conditions prevent plant material from fully decomposing.
	<b>Floodplains and riverine wetlands</b>	Flat areas of land next to a river or stream permanently or temporary flooded.
	<b>Freshwater Wetland</b>	Non-peat freshwater systems such as marshes, swamps and ponds.
	<b>Seagrass</b>	Submerged marine angiosperm meadows occupying shallow coastal areas.
	<b>Saltmarsh</b>	Intertidal coastal wetlands dominated by halophytic vegetation and subject to tidal flooding.
	<b>Mangrove</b>	Intertidal forested wetlands in tropical and subtropical coasts dominated by mangrove tree species.

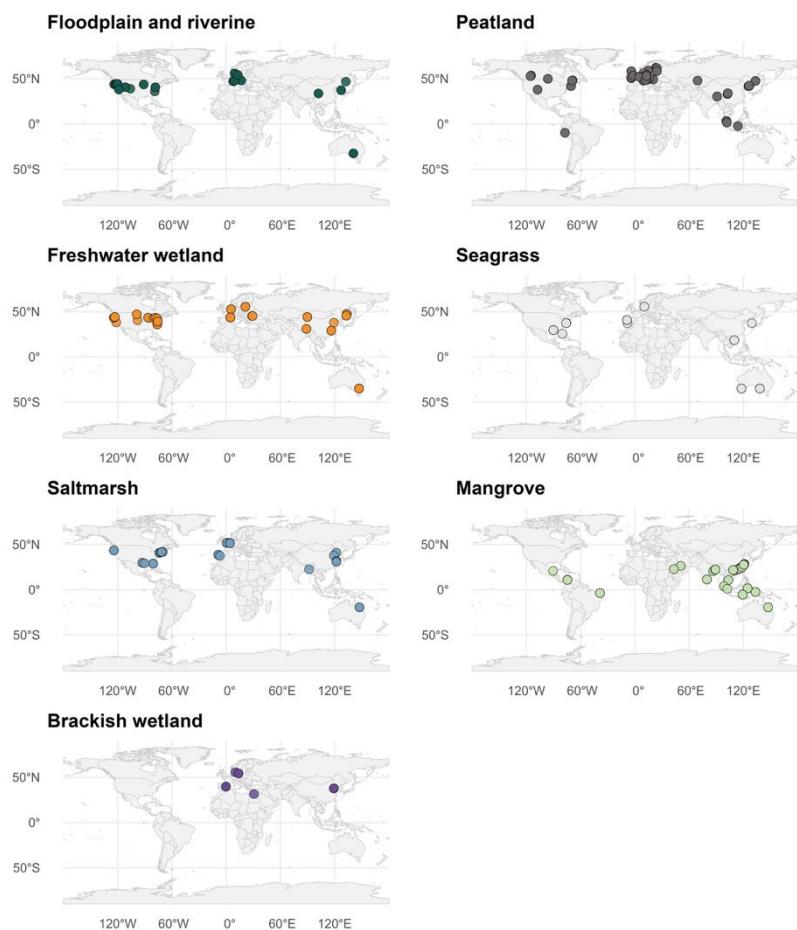
	<b>Brackish Wetland</b>	Wetlands influenced by mixed freshwater and saline conditions, including estuarine and lagoonal systems.
<b>Restoration Type</b>	<b>Hydrological Restoration</b>	Actions focused to restore natural water flow patterns, water levels, and hydroperiods within wetland ecosystems.
	<b>Morphological Restoration</b>	Actions targeting geomorphic structure, including channel reconfiguration, topographic re-construction, shoreline reshaping or creation of microtopographic features.
	<b>Soil-focused Restoration</b>	Measures focused on soil properties or processes, such as re-establishing sediment inputs, changing soil organic matter, stabilising substrates or reducing erosion.
	<b>Vegetation-focused Restoration</b>	Direct interventions involving planting, seeding, translocation or protection of native vegetation to accelerate recovery of above- and belowground biomass and habitat structure.
	<b>Passive restoration</b>	Process that eliminates the factors of degradation and disturbance and permits the natural regeneration of the ecosystem.

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1120 *Supplementary Figure S2. Distribution of restoration projects by wetland type*

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1124 *Supplementary Table S3. Pooled effect sizes (standardized mean differences, SMD) for carbon*  
1125 *indicators comparing restored and unrestored wetlands. Values include point estimates, 95%*  
1126 *confidence intervals, significance category, and number of comparisons and studies*  
1127 *contributing to each estimate.*

Parameter	Estimate	Lover CI	Upper CI	Significance	Nr. Comparisons	Nr. Studies
Soil carbon	0.615119621	0.234490317	0.995748925	Positive effect	207	95
Aboveground biomass	2.518918974	1.166814623	3.871023326	Positive effect	73	34
Belowground biomass	1.938163518	0.952974364	2.923352672	Positive effect	53	21
CO <sub>2</sub> flux	- 0.427299424	- 0.841017922	- 0.013580926	Negative effect	111	45
CH <sub>4</sub> flux	0.439658673	0.141630129	0.737687218	Positive effect	95	40
N <sub>2</sub> O flux	- 0.229069545	- 0.490360119	0.032221028	Non-significant	50	19
DOC	0.027659372	- 0.309282062	0.364600805	Non-significant	28	9

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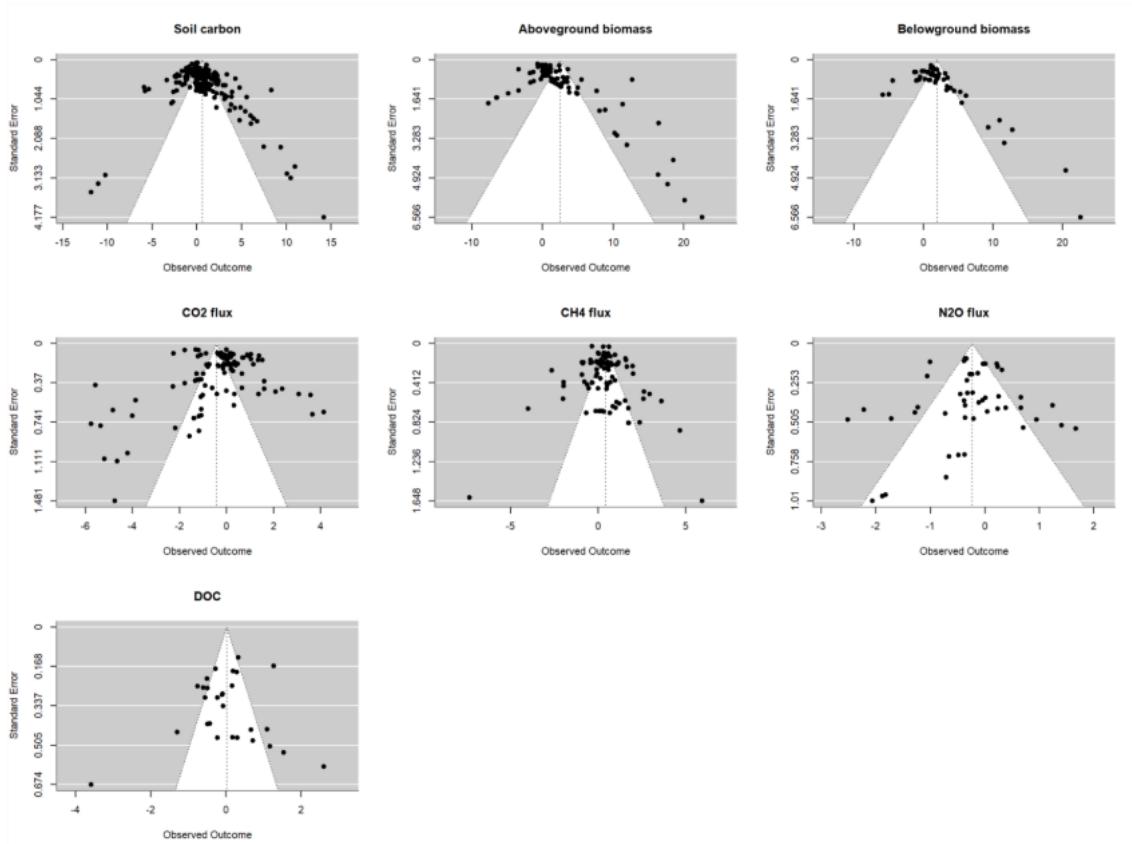
1131 *Supplementary Table S4. Heterogeneity statistics for meta-analyses of carbon indicators.*  
1132 *Reported metrics include Q statistic, degrees of freedom, p-value, I<sup>2</sup> (percentage of total*  
1133 *variability due to heterogeneity), and τ<sup>2</sup> (between-study variance).*

Parameter	k	Q	df_Q	p_Q	I <sup>2</sup> _Q	τ <sup>2</sup> _total
Soil carbon	207	1930.65	206	1.85E-277	89.3	3.6217
Aboveground biomass	73	1070.39	72	1.20E-177	93.3	17.79
Belowground biomass	53	352.53	52	3.01E-46	85.2	6.4046
CO <sub>2</sub> flux	111	4108.56	110	0	97.3	2.8538
CH <sub>4</sub> flux	95	1494.81	94	7.52E-251	93.7	1.1224
N <sub>2</sub> O flux	50	243.09	49	1.57E-27	79.8	0.4433
DOC	28	178.78	27	2.52E-24	84.9	0.6477

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1138 *Supplementary Figure S3. Funnel plots for publication bias assessment across all indicators.*

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1141 *Supplementary Table S5. Publication bias assessment for each indicator. Egger's regression*  
1142 *test and Kendall's tau rank correlation are reported with corresponding p-values.*

Parameter	k	Egger_Z	Egger_p	Kendall_tau	Kendall_p
Soil carbon	207	4.049	7.28E-05	0.274	4.68E-09
Aboveground biomass	73	4.071	1.20E-04	0.452	1.53E-08
Belowground biomass	53	2.315	0.024644	0.361	1.33E-04
CO2 flux	111	0.911	0.364319	-0.19	0.00309
CH4 flux	95	1.045	0.298928	0.095	0.172812
N2O flux	50	0.127	0.899318	-0.116	0.234876
DOC	28	-0.781	0.441625	0.138	0.316765

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1146 **Supplementary Table S6. Leave-one-out sensitivity analysis for pooled effect sizes. For each**  
1147 **indicator, the table shows the full-model effect estimate with 95% CI, range of estimates under**  
1148 **leave-one-out scenarios, maximum absolute change in SMD, and number of significance flips.**

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Parameter	k_effects	Full effect (95% CI)	Range LOO est.	Max  ΔSMD	Sig. flips
Soil carbon	207	0.63 [0.37, 0.88]	0.58 - 0.65	0.05	0
Aboveground biomass	73	2.46 [1.46, 3.46]	2.24 - 2.55	0.22	0
Belowground biomass	53	1.67 [0.92, 2.43]	1.52 - 1.74	0.16	0
CO <sub>2</sub> flux	111	-0.39 [-0.69, 0.09]	-0.42 - -0.33	0.05	0
CH <sub>4</sub> flux	95	0.36 [0.13, 0.60]	0.33 - 0.40	0.03	0
N <sub>2</sub> O flux	50	-0.24 [-0.45, 0.03]	-0.27 - -0.20	0.04	0
DOC	28	0.03 [-0.30, 0.37]	-0.04 - 0.11	0.08	0

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1154 *Supplementary Table S7. Influence diagnostics for meta-analyses of carbon indicators. The*  
1155 *table reports the number and proportion of influential cases identified for each indicator.*

Parameter	k_effects	n_influential	prop_influential
Soil carbon	207	0	0.00%
Aboveground biomass	73	0	0.00%
Belowground biomass	53	0	0.00%
CO2 flux	111	0	0.00%
CH4 flux	95	0	0.00%
N2O flux	50	0	0.00%
DOC	28	0	0.00%

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1161 **Supplementary Table S8. Fit statistics for multivariable meta-regression models of restoration**  
 1162 **effects on carbon indicators.** Summary of full multilevel meta-regression models fitted separately  
 1163 for each response variable (soil carbon, aboveground biomass, belowground biomass, dissolved  
 1164 organic carbon, and  $CO_2$ ,  $CH_4$  and  $N_2O$  fluxes). For each parameter, the table reports the number  
 1165 of comparisons included ( $k$ ), the omnibus test for moderators ( $QM$ , degrees of freedom and  $p$ -  
 1166 value), measures of residual heterogeneity (total  $\tau^2$ ) and overall model fit (AIC and BIC).  
 1167 Significant  $QM$  values indicate that effect sizes vary systematically along at least one of the  
 1168 moderators.

parameter	k	AIC	BIC	tau2_total	QM	QM_df	QM_p
Soil carbon	105	362.5622	420.6144	2.984231	31.56513	21	0.064753
Aboveground biomass	28	109.0777	119.3027	75.55709	4.39554	13	0.98621
Belowground biomass	26	108.6303	117.1046	23.99945	3.045648	12	0.995213
$CO_2$ flux	59	443.7996	478.0711	4.061636	56.73534	17	3.60E-06
$CH_4$ flux	44	143.4998	168.1207	1.780619	16.14293	16	0.443031
$N_2O$ flux	21	33.65455	38.42929	0.040141	51.28654	9	6.17E-08
DOC	17	27.60671	28.48057	0.063485	30.03669	8	2.08E-04

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1172 **Supplementary Table S9. Regression coefficients for full multivariable meta-regression models.**  
1173 *Estimated fixed effects from the full multilevel meta-regression models for each response variable.*  
1174 *The table lists the intercept and all moderator terms (alteration type, wetland type, restoration*  
1175 *type/category, sampling and restoration years, and climatic variables), together with their*  
1176 *estimates ( $\beta$ ), standard errors, Wald z-values, 95% confidence intervals and p-values. Continuous*  
1177 *moderators were z-standardised prior to modelling, so coefficients represent the change in*  
1178 *standardized mean difference (SMD) associated with a one standard deviation change in the*  
1179 *predictor. Positive coefficients indicate larger values in restored compared to altered wetlands*  
1180 *along that gradient; negative coefficients indicate smaller restored–altered differences.*

term	estimate	se	zval	pval	ci.lb	ci.ub
<b>Soil carbon</b>						
intercept	-0.89085	1.529534	-0.58243	0.560274	-3.88868	2.106979
Alteration type: Hydrology	1.590559	1.592605	0.998715	0.317933	-1.53089	4.712006
Alteration type: Hydromorphology	1.430628	1.830522	0.781541	0.434484	-2.15713	5.018386
Alteration type: Land use change	2.260088	1.472632	1.534727	0.124851	-0.62622	5.146393
Alteration type: Morphology	2.206393	1.838534	1.200083	0.230107	-1.39707	5.809855
Alteration type: Vegetation/habitat loss	1.914139	1.62475	1.178113	0.238752	-1.27031	5.098591
Alteration type: Water quality/pollution	2.622893	1.69931	1.543505	0.122708	-0.70769	5.953479
Wetland type: Peatland	1.722315	1.073261	1.60475	0.108549	-0.38124	3.825867
Wetland type: Freshwater wetland	-0.16156	1.28472	-0.12576	0.899924	-2.67957	2.356442
Wetland type: Seagrass	0.520746	1.314861	0.396046	0.692071	-2.05633	3.097827
Wetland type: Saltmarsh	-0.84809	1.251197	-0.67782	0.497886	-3.30039	1.604215
Wetland type: Mangrove	-1.25008	1.306007	-0.95717	0.338479	-3.8098	1.309649
Wetland type: Brackish wetland	0.461961	1.909605	0.241914	0.808846	-3.2808	4.204718
Restoration type: Morphology	-1.01048	1.017063	-0.99353	0.320453	-3.00389	0.982926
Restoration type: Passive restoration	0.50658	1.112898	0.45519	0.648972	-1.67466	2.68782
Restoration type: Soils	-0.94808	1.567216	-0.60495	0.545214	-4.01977	2.123603
Restoration type: Vegetation	-0.55756	0.6385	-0.87324	0.382533	-1.809	0.693874
Temperature	1.633681	0.557375	2.931025	0.003378	0.541245	2.726116
Temperature seasonality	1.434123	0.507371	2.826575	0.004705	0.439693	2.428552
Precipitation	0.013217	0.241126	0.054815	0.956285	-0.45938	0.485816
Precipitation seasonality	0.285491	0.349326	0.817261	0.413779	-0.39918	0.970158
Time since restoration	0.211918	0.142127	1.491046	0.135949	-0.06665	0.490481
<b>Aboveground biomass</b>						
intercept	-152.85	140.187	-1.09033	0.275567	-427.612	121.9112
Alteration type: Hydrology	-192.583	195.6343	-0.9844	0.324917	-576.019	190.8529
Alteration type: Land use change	9.49456	13.23703	0.717273	0.473206	-16.4495	35.43865
Alteration type: Vegetation/habitat loss	15.84706	27.49035	0.576459	0.564305	-38.033	69.72716

Alteration type: Water quality/pollution	21.49658	13.23729	1.623941	0.104388	-4.44803	47.4412
Wetland type: Peatland	-353.773	348.2545	-1.01585	0.309703	-1036.34	328.7934
Wetland type: Seagrass	359.0109	329.8658	1.088355	0.276439	-287.514	1005.536
Wetland type: Saltmarsh	566.5262	540.6693	1.047824	0.29472	-493.166	1626.219
Wetland type: Mangrove	214.8835	196.1159	1.095697	0.273212	-169.497	599.2635
Temperature	-130.797	131.7795	-0.99255	0.320932	-389.08	127.4859
Temperature seasonality	59.90325	57.2369	1.046585	0.295291	-52.279	172.0855
Precipitation	-29.2548	20.76247	-1.40902	0.158828	-69.9485	11.4389
Precipitation seasonality	129.2651	125.3828	1.030963	0.302558	-116.481	375.0109
Time since restoration	0.162408	0.777043	0.209008	0.834442	-1.36057	1.685384
<b>Belowground biomass</b>						
intercept	56.04572	71.62783	0.782457	0.433946	-84.3422	196.4337
Alteration type: Hydrology	-40.4241	74.25791	-0.54437	0.586184	-185.967	105.1187
Alteration type: Land use change	-8.47452	12.457	-0.6803	0.496313	-32.8898	15.94075
Alteration type: Vegetation/habitat loss	-2.68254	13.02805	-0.20591	0.836865	-28.217	22.85196
Alteration type: Water quality/pollution	34.49373	40.05471	0.861165	0.389147	-44.012	112.9995
Wetland type: Peatland	4.917093	12.80639	0.383956	0.701011	-20.183	30.01715
Wetland type: Seagrass	-63.8298	78.11072	-0.81717	0.413831	-216.924	89.2644
Wetland type: Saltmarsh	-7.32939	11.79357	-0.62147	0.534288	-30.4444	15.78558
Wetland type: Mangrove	-62.3729	81.69822	-0.76345	0.445192	-222.498	97.75267
Temperature	19.97926	36.82915	0.542485	0.587484	-52.2046	92.16308
Temperature seasonality	14.59312	33.0207	0.441938	0.658534	-50.1263	79.31251
Precipitation	13.64304	14.69982	0.928109	0.353351	-15.1681	42.45415
Time since restoration	0.131361	1.313505	0.100008	0.920338	-2.44306	2.705783
<b>CO2 flux</b>						
intercept	0.420679	2.142805	0.196322	0.844358	-3.77914	4.6205
Alteration type: Hydromorphology	0.667678	2.119489	0.315018	0.752748	-3.48644	4.8218
Alteration type: Land use change	1.240837	1.851078	0.670332	0.502646	-2.38721	4.868884
Alteration type: Morphology	0.813639	2.935922	0.277132	0.781679	-4.94066	6.567941
Alteration type: Natural disasters	-3.01672	2.764841	-1.0911	0.275229	-8.43571	2.40227

Alteration type: Water quality/pollution	5.707711	4.673849	1.221202	0.22201	-3.45286	14.86829
Wetland type: Peatland	-3.14267	2.500159	-1.25699	0.208757	-8.0429	1.757548
Wetland type: Freshwater wetland	0.190459	2.638498	0.072185	0.942455	-4.9809	5.361821
Wetland type: Saltmarsh	1.438293	3.306173	0.435033	0.663539	-5.04169	7.918272
Wetland type: Mangrove	2.171613	4.765393	0.455705	0.648602	-7.16839	11.51161
Restoration type: Morphology	-3.15182	2.097892	-1.50237	0.133	-7.26361	0.959973
Restoration type: Passive restoration	1.674649	2.509805	0.667243	0.504617	-3.24448	6.593776
Restoration type: Vegetation	0.80585	0.816793	0.986603	0.323837	-0.79503	2.406734
Temperature	-3.66963	1.422503	-2.5797	0.009889	-6.45768	-0.88158
Temperature seasonality	-0.97287	0.824332	-1.18019	0.237924	-2.58853	0.642792
Precipitation	1.381217	1.290901	1.069963	0.284636	-1.1489	3.911338
Precipitation seasonality	-1.23974	0.674438	-1.83819	0.066034	-2.56162	0.082129
Time since restoration	0.471232	0.318076	1.481508	0.138471	-0.15219	1.09465
<b>CH4 flux</b>						
intercept	4.27503	2.253992	1.896649	0.057874	-0.14271	8.692773
Alteration type: Hydromorphology	0.697404	1.165494	0.598376	0.549589	-1.58692	2.98173
Alteration type: Land use change	0.470002	1.015326	0.462908	0.643431	-1.52	2.460004
Alteration type: Morphology	-4.06889	2.384161	-1.70663	0.08789	-8.74176	0.603984
Alteration type: Water quality/pollution	-2.10085	2.893813	-0.72598	0.467852	-7.77261	3.570922
Wetland type: Peatland	-3.82887	2.021129	-1.89442	0.058169	-7.79021	0.132469
Wetland type: Freshwater wetland	-3.89003	2.19434	-1.77276	0.076269	-8.19086	0.410795
Wetland type: Saltmarsh	-0.97025	2.405812	-0.40329	0.686733	-5.68555	3.745059
Wetland type: Mangrove	-3.9102	2.457168	-1.59135	0.111532	-8.72616	0.905757
Restoration type: Hydromorphology	-0.50152	0.954083	-0.52566	0.599125	-2.37149	1.368445
Restoration type: Morphology	-0.74717	1.01681	-0.73482	0.462449	-2.74008	1.245738
Restoration type: Vegetation	-0.588	0.82037	-0.71674	0.473532	-2.19589	1.019901
Temperature	-0.47904	0.847599	-0.56517	0.57196	-2.1403	1.182228
Temperature seasonality	-0.76585	0.486596	-1.57389	0.115513	-1.71956	0.187862
Precipitation	-0.47282	0.774619	-0.6104	0.5416	-1.99105	1.045401

Precipitation seasonality	-0.28551	0.300321	-0.95067	0.341772	-0.87412	0.303112
Time since restoration	0.028597	0.311937	0.091676	0.926956	-0.58279	0.639983
<b>N2O flux</b>						
intercept	-40.179	47.05962	-0.85379	0.393222	-132.414	52.05618
Alteration type:						
Hydromorphology	-44.4253	50.96977	-0.8716	0.383427	-144.324	55.47366
Wetland type: Peatland	49.38638	58.8077	0.839794	0.401024	-65.8746	164.6473
Wetland type: Freshwater wetland	79.06523	93.50783	0.845547	0.397806	-104.207	262.3372
Wetland type: Mangrove	48.37076	59.07896	0.818748	0.41293	-67.4219	164.1634
Restoration type: Vegetation	18.23793	20.06755	0.908827	0.363442	-21.0938	57.56961
Temperature	-49.3458	57.25985	-0.86179	0.388805	-161.573	62.88147
Temperature seasonality	-44.489	52.97749	-0.83977	0.401037	-148.323	59.34501
Precipitation	-2.22722	4.267234	-0.52194	0.601715	-10.5908	6.136403
Time since restoration	0.59756	0.304824	1.960342	0.049956	1.15E-04	1.195004
<b>DOC</b>						
intercept	3.092653	1.405383	2.200577	0.027766	0.338153	5.847154
Alteration type:						
Hydromorphology	-12.1958	11.07424	-1.10127	0.270778	-33.9009	9.50935
Alteration type: Land use change	44.5532	52.29589	0.851945	0.394245	-57.9449	147.0513
Restoration type: Vegetation	-21.9018	23.01498	-0.95163	0.341284	-67.0103	23.20676
Temperature	-29.1864	32.12721	-0.90846	0.363634	-92.1546	33.78179
Temperature seasonality	-13.9061	15.18062	-0.91604	0.359644	-43.6596	15.84736
Precipitation	2.967315	2.943476	1.008099	0.313407	-2.80179	8.736422
Precipitation seasonality	0.2251	0.77051	0.292145	0.770176	-1.28507	1.735272
Time since restoration	-0.26159	0.186014	-1.40631	0.159633	-0.62617	0.102988

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1183 **Supplementary Table S10. Single-term deletion (drop-one) model comparison for multivariable**  
1184 **meta-regressions.** Results of the single-term deletion analysis used to assess the relative  
1185 importance of individual moderators in the multivariable meta-regression models. For each  
1186 response variable and each moderator (or moderator block, in the case of categorical factors),  
1187 the table reports the change in Akaike information criterion ( $\Delta AIC$ ), change in Bayesian  
1188 information criterion ( $\Delta BIC$ ) and change in total  $\tau^2$  when that term is removed from the full model.  
1189 Positive  $\Delta AIC/\Delta BIC$  values indicate a worsening of model fit when the moderator is dropped,  
1190 with larger increases highlighting moderators that contribute more strongly to explaining  
1191 between-comparison heterogeneity.

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term	AIC	BIC	tau2_red	dAIC	dBIC	d_tau2
Soil carbon						
Alteration type	374.730	419.525	2.729	12.167	-1.089	-0.256
Wetland type	378.244	423.039	2.882	15.682	2.425	-0.103
Restoration type	369.592	418.910	2.685	7.030	-1.704	-0.300
Temperature	371.789	427.698	3.509	9.227	7.084	0.525
Temp. seasonality	371.683	427.591	3.453	9.120	6.977	0.469
Precipitation	363.296	419.205	2.920	0.734	-1.410	-0.064
Pre. seasonality	364.350	420.259	2.850	1.788	-0.355	-0.134
Time since restoration	799.730	879.554	2.920	437.167	458.939	-0.064
Aboveground biomass						
Alteration type	125.326	136.158	65.014	16.248	16.855	-10.544
Wetland type	113.135	123.756	51.666	4.057	4.453	-23.891
Restoration type	109.078	119.303	75.557	0.000	0.000	0.000
Temperature	112.902	123.522	48.563	3.824	4.220	-26.994
Temp. seasonality	112.771	123.392	45.363	3.693	4.089	-30.194
Precipitation	113.591	124.212	40.550	4.513	4.909	-35.007
Pre. seasonality	112.801	123.422	46.244	3.723	4.119	-29.313
Time since restoration	327.999	374.363	24.564	218.922	255.061	-50.993
Belowground biomass						
Alteration type	114.863	124.068	15.279	6.233	6.963	-8.720
Wetland type	112.003	120.950	16.813	3.373	3.845	-7.187
Restoration type	108.630	117.105	23.999	0.000	0.000	0.000
Temperature	108.630	117.105	23.999	0.000	0.000	0.000
Temp. seasonality	108.630	117.105	24.000	0.000	0.000	0.000
Precipitation	108.630	117.105	24.000	0.000	0.000	0.000
Pre. seasonality	108.630	117.105	23.999	0.000	0.000	0.000
Time since restoration	227.637	258.744	19.364	119.007	141.640	-4.635
CO2 flux						
Alteration type	456.348	483.778	3.636	12.548	5.707	-0.425
Wetland type	452.827	481.733	3.717	9.027	3.662	-0.345
Restoration type	452.268	482.599	4.287	8.469	4.528	0.226

Temperature	450.882	483.897	6.458	7.082	5.826	2.396
Temp. seasonality	446.955	479.971	4.258	3.156	1.900	0.196
Precipitation	446.521	479.537	3.874	2.721	1.465	-0.188
Pre. seasonality	448.571	481.587	5.017	4.772	3.516	0.955
Time since restoration	624.257	683.439	4.364	180.457	205.368	0.303
CH4 flux						
Alteration type	152.007	173.517	2.011	8.508	5.397	0.231
Wetland type	150.630	172.140	1.956	7.130	4.019	0.175
Restoration type	145.758	168.177	1.574	2.258	0.056	-0.206
Temperature	144.322	168.302	1.681	0.822	0.181	-0.100
Temp. seasonality	146.257	170.237	1.973	2.757	2.116	0.192
Precipitation	144.406	168.385	1.670	0.906	0.265	-0.111
Pre. seasonality	144.972	168.952	1.663	1.472	0.831	-0.117
Time since restoration	337.709	388.399	1.085	194.210	220.278	-0.696
N2O flux						
Alteration type	33.655	38.429	0.040	0.000	0.000	0.000
Wetland type	31.925	37.259	0.037	-1.730	-1.171	-0.003
Restoration type	33.655	38.429	0.040	0.000	0.000	0.000
Temperature	33.655	38.429	0.040	0.000	0.000	0.000
Temp. seasonality	33.655	38.429	0.040	0.000	0.000	0.000
Precipitation	33.655	38.429	0.040	0.000	0.000	0.000
Pre. seasonality	33.655	38.429	0.040	0.000	0.000	0.000
Time since restoration	99.576	128.256	1.079	65.921	89.826	1.039
DOC						
Alteration type	26.697	28.669	0.068	-0.910	0.188	0.004
Wetland type	27.607	28.481	0.063	0.000	0.000	0.000
Restoration type	26.395	28.367	0.063	-1.212	-0.114	0.000
Temperature	26.313	28.285	0.062	-1.294	-0.196	-0.001
Temp. seasonality	26.327	28.299	0.062	-1.280	-0.181	-0.001
Precipitation	26.503	28.475	0.065	-1.104	-0.005	0.002
Pre. seasonality	25.530	27.502	0.052	-2.077	-0.978	-0.012
Time since restoration	56.756	67.573	3061.445	29.150	39.092	3061.382



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