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1 **Title: Coastal vegetation and estuaries collectively are a greenhouse gas sink**

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33 **Abstract:**

34 Coastal ecosystems release or absorb carbon dioxide (CO₂), methane (CH₄), and nitrous oxide
35 (N₂O), but the net effects of these ecosystems on the radiative balance remain unknown. We
36 compiled a dataset of observations from 738 sites from studies published between 1975 and
37 2020 to quantify CO₂, CH₄, and N₂O fluxes in estuaries and coastal vegetation in ten global
38 regions. We show that the CO₂-equivalent (CO_{2e}) uptake by coastal vegetation is reduced by
39 23-27% due to estuarine CO_{2e} outgassing, resulting in a global median net sink of 391 or 444
40 Tg CO_{2e} yr⁻¹ using the 20 or 100-year global warming potential, respectively. Globally, total
41 coastal CH₄ and N₂O emissions reduce the coastal CO₂ sink by 9-20%. Southeast Asia, North
42 America, and Africa are critical regional hotspots of greenhouse gas sinks. Understanding these
43 hotspots can guide our efforts to strengthen coastal CO₂ uptake while effectively reducing CH₄
44 and N₂O emissions.

45

46

47 Keywords: Coastal vegetation, estuaries, climate change, greenhouse gas budget, radiative
48 balance

49

50 **Main text:**

51 Since the beginning of the industrial era, atmospheric concentrations of the greenhouse gases
52 (GHG) CO₂, CH₄, and N₂O have increased by 47%, 156% and 23%, respectively, and continue
53 to increase at alarming rates due to anthropogenic activities, driving global warming¹. The
54 global terrestrial CO₂ sink is diminished by biogenic CH₄ and N₂O emissions due to
55 anthropogenic activities such as agriculture or biomass burning, which resulted in a shift of the
56 terrestrial biosphere from a net CO₂ sink to a net GHG source². Globally, it is now clear that
57 inland waters are an atmospheric GHG source^{3,4} while the GHG budget of coastal ecosystems
58 is less certain. Located in the downstream portion of the Land-Ocean-Aquatic-Continuum
59 (LOAC)⁵, estuaries receive large amounts of terrestrial carbon (C) and nitrogen (N) through
60 riverine and groundwater flows, but their C and N cycles are also tightly inter-connected with
61 coastal vegetated ecosystems⁶ and the coastal oceans^{7,8}. The GHG radiative balance of
62 estuaries and coastal vegetation are thus a complex spatial and temporal combination of GHG
63 sources and sinks⁷, which complicates the estimate of the net global warming effect and makes
64 the implementation of efficient mitigation strategies difficult.

65

66 Estuaries (tidal systems and deltas, lagoons, fjords) and surrounding coastal vegetation
67 (mangroves, salt marshes, seagrasses) are strongly interconnected but show remarkable
68 variations in the magnitude and direction (sink or source) of GHG fluxes. Estuaries have been
69 estimated to emit 0.4 to 2.2 Pg CO₂ yr⁻¹ globally⁷⁻¹⁰, whereas mangroves and salt marshes
70 collectively take up 0.3 to 1.7 Pg CO₂ yr⁻¹ ^{5,8} (Supplementary Table S1). Together with
71 submergent seagrasses, coastal vegetation potentially store 304 (131-466) Tg CO₂ yr⁻¹ of so-
72 called 'blue carbon' in their sediments^{11,12}. On the other hand, emissions of CH₄ and N₂O from
73 coastal sediments and surrounding waters can reduce some of the coastal vegetation carbon
74 sinks^{13,14}, thereby complicating 'blue carbon' assessments^{15,16}. A recent global synthesis
75 showed that median CH₄ emissions from combined mangrove, salt marshes, and seagrasses
76 (0.52 Tg CH₄ yr⁻¹) exceed those from estuaries (0.23 Tg CH₄ yr⁻¹)¹⁷. However, CH₄ fluxes are
77 highly variable across time and space, causing a large range in global estimates of both coastal
78 vegetation (0.02 to 6.2 Tg CH₄ yr⁻¹)^{17,18} and estuaries (0.02 to 6.6 Tg CH₄ yr⁻¹)^{4,19}
79 (Supplementary Table S2). Coastal N₂O fluxes are less understood and it remains unclear if
80 coastal vegetated ecosystems are a net source or sink of N₂O to the atmosphere^{20,21}. Global
81 estimates of estuarine N₂O emissions are highly uncertain, with large discrepancies for both
82 observation-based (110 to 2,844 Gg N₂O yr⁻¹)^{4,22,23} and modeling (94 to 1,084 Gg N₂O yr⁻¹)^{24,25}
83 approaches (Supplementary Table S3).

84

85 Here we present a data-driven meta-analysis synthesizing CO₂, CH₄, and N₂O fluxes in three
86 major estuary types (tidal systems and deltas, lagoons, and fjords) and three coastal vegetation
87 types (mangroves, salt marshes, seagrasses), globally and in ten world regions. These regions
88 were delineated by the REgional Carbon Cycle Assessment and Processes-2 (RECCAP2)
89 project²⁶, an activity of the Global Carbon Project. We compiled water-air CO₂, CH₄, and N₂O
90 fluxes for estuaries and combined CO₂ fluxes from eddy-covariance with CH₄ and N₂O fluxes
91 at the interfaces of water, sediments, and plants with the atmosphere for coastal vegetation.
92 Our dataset compiles observations from a total of 738 sites from studies published between
93 1975 and end of 2020 and we provide a GHG flux climatology assumed representative for this
94 period. The GHG fluxes are regionalized by combining GHG flux densities with recently
95 published surface areas of coastal vegetation²⁷⁻²⁹ and estuaries³⁰. As such, our global-scale
96 regional assessment applies a consistent framework for all three GHGs and ecosystems.
97 Regional GHG fluxes are then summed to provide global net GHG fluxes for estuaries, coastal
98 vegetation, and both systems combined. We quantify the net contemporary GHG radiative
99 balance in units of CO₂-equivalents (CO₂e) based on the global warming potential (GWP) of
100 each gas for the 20-year and 100-year time horizons¹. Note that our study does not include an
101 analysis of temporal changes in GHG fluxes required to assess a contribution to radiative
102 forcing. Following the RECCAP2 regional segmentation, our coastal GHG budget can be
103 integrated into broader budgets of continents.

104

105 **Estuarine and coastal vegetation CO₂ fluxes**

106 We estimate global median [first (Q1)-third (Q3) quartile] CO₂ emissions from estuaries at 111
107 [73-170] Tg CO₂ yr⁻¹ (Table 1), which is three to five times lower (as is our mean of 121 Tg
108 CO₂ yr⁻¹) than recent mean estimates (370-550 Tg CO₂ yr⁻¹)^{9,10}. Our lower global emissions
109 are due to the inclusion of estuarine surface area³⁰ (Extended Data Figure 1) that is ~30% lower
110 than previously estimated^{31,32}, and includes more sites than previous studies^{9,10}. In particular,
111 CO₂ flux data in fjords have more than doubled since earlier reviews^{8,10} that considered fjords
112 as minimal CO₂ sources. Here we show that fjords take up 66 [96-24] Tg CO₂ yr⁻¹ from the
113 atmosphere, reducing 37% of the CO₂ emissions from global tidal systems and deltas (127 [90-
114 176] Tg CO₂ yr⁻¹) and lagoons (50 [28-81] Tg CO₂ yr⁻¹) (Figure 1). Tidal systems and deltas
115 which account for 40% of the global estuarine surface area (Supplementary Table S4) show
116 higher CO₂ flux densities than lagoon and fjords (Supplementary Table S5), likely due to their
117 strong hydrological connectivity with rivers and groundwater that import CO₂ supersaturated

118 waters³³. In addition, the considerably strong influence by tides³¹ can increase water turbulence
119 and therefore gas transfer velocities³⁴ which in turn enhance CO₂ evasion from tidal systems.
120 At the regional scale, we find distinct trends of CO₂ fluxes between different geomorphic
121 estuary types (Figure 1). For example, North America is a hotspot for atmospheric CO₂ uptake
122 by fjords (57 [76-30] Tg CO₂ yr⁻¹, 86% of the CO₂ uptake by global fjords). Long stretches of
123 lagoons can be found along Africa's coastline that contribute 24% to global lagoon CO₂
124 emissions. Europe accounts for 20% of global CO₂ emissions by tidal systems and deltas
125 although only comprising 5% of the total surface area of this estuary type. This
126 disproportionate contribution stems from the highest median CO₂ flux densities (4.7 g CO₂ m⁻²
127 d⁻¹) of any estuary type or region (Supplementary Table S5), likely fuelled by organic carbon
128 loads from European rivers under strong anthropogenic pressure³⁵. European estuaries were
129 over-represented in previous global analyses^{9,10,36}, which likely led to overestimates in global
130 CO₂ emissions from estuaries.

131

132 Using data exclusively from eddy-covariance long-term studies, we estimate that salt marshes,
133 mangroves, and seagrasses worldwide take up 601 [774-426] Tg CO₂ yr⁻¹ from the atmosphere,
134 a flux which, in absolute terms, is more than five times greater than the estuarine CO₂
135 outgassing (Table 1). Our global estimate is lower but within the uncertainty range of a recent
136 estimate of 843±440 Tg CO₂ yr⁻¹ by mangroves, salt marshes, and seagrasses⁵. Note that our
137 estimate of CO₂ uptake by coastal vegetation should be distinguished from carbon
138 sequestration by coastal vegetation, which has been estimated to be 110-257 Tg CO₂ yr⁻¹³⁷,
139 roughly 1/3 of the CO₂ uptake, the remainder resulting in a lateral export of carbon, a substantial
140 fraction of which can be transported over long distances to the open ocean⁵.

141 Seagrass meadows can be found in tropical, subtropical, and temperate-cold climate zones.
142 Mangroves are only abundant in subtropical and tropical climates, whereas salt marshes
143 dominate in temperate regions (Extended Data Figure 1, Extended Data Figure 2). Highly
144 productive mangrove forests³⁸ contribute the majority (60%) of the global CO₂ uptake by
145 coastal vegetation (359 [434-333] Tg CO₂ yr⁻¹). Globally, seagrasses (192 [354-53] Tg CO₂ yr⁻¹)
146 have a ~4 times higher CO₂ uptake than salt marshes (50 [59-41] Tg CO₂ yr⁻¹) (Figure 2),
147 which is similar to previous findings based on net primary production³⁹. Since mangrove
148 forests are abundant in tropical southeast Asia, we find that this region contributes 37% to
149 global mangrove CO₂ uptake, followed by Africa (20%), and tropical North and South America
150 (both ~15%). Salt marshes are abundant along the west and east coasts of Canada and the US.
151 Therefore, North America contributes 34% of the global salt marsh CO₂ uptake. Other regions

152 that provide a substantial salt marsh CO₂ sink include East Asia (23%) and Australasia (21%).
153 Africa's coastline has the greatest contribution (24%) to global seagrass CO₂ uptake.

154

155 **Estuarine and coastal vegetation CH₄ fluxes**

156 Global estuaries emit 0.25 [0.07-0.46] Tg CH₄ yr⁻¹ (Table 1), a median value that falls at the
157 lower end of the range of previous assessments^{4,19,40} (Supplementary Table S2), yet is close to
158 the recent median estimate (0.23 [0.02-0.91] Tg CH₄ yr⁻¹) by Rosentreter et al.¹⁷. Global
159 estuarine CH₄ emissions are dominated by tidal systems and deltas (0.14 [0.11-0.23] Tg CH₄
160 yr⁻¹, 56%), and lagoons (0.11 [0.05-0.13] Tg CH₄ yr⁻¹, 44%), while fjords contribute <1%. Our
161 regional analysis shows that North America's tidal systems and deltas contribute the majority
162 (50%) of global CH₄ emissions from this estuary type (Figure 1). North America also shows
163 the highest median CH₄ flux densities (2.9 mg CH₄ m⁻² d⁻¹) of tidal systems and deltas,
164 particularly in the USA, where many eutrophic systems undergo seasonal hypoxia or anoxia
165 that fuel CH₄ production⁴¹. South America, Russia, and Australasia contribute around 10%
166 each to global CH₄ emissions from tidal systems and deltas, whereas all other regions are minor
167 contributors (<5%). North America comprises the largest area of lagoons (Extended Data
168 Figure 1), thereby dominating global lagoonal CH₄ emissions (30%). Other regions with
169 substantial lagoon CH₄ emissions are Africa (25%) and South America (14%), whereas the
170 remaining regions contribute <10%. Interestingly, we find that of all estuary types and regions,
171 lagoons in Africa show the highest flux density (5.3 mg CH₄ m⁻² d⁻¹), driven by high CH₄
172 production in anoxic bottom waters in permanently stratified lagoons along the Ivory Coast⁴².

173

174 Globally, mangroves, salt marshes, and seagrasses combined release 0.76 [0.47-1.41] Tg CH₄
175 yr⁻¹ which is more than three times the CH₄ released by global estuaries. Our coastal vegetation
176 estimate is comparable with a recent review¹⁷ but incorporates more data, which results in an
177 increased estimate for salt marshes and mangroves but not seagrasses. We find that mangroves
178 (0.34 [0.21-0.50] Tg CH₄ yr⁻¹) dominate coastal vegetation CH₄ emissions, followed by salt
179 marshes (0.26 [0.14-0.36] Tg CH₄ yr⁻¹), and seagrasses (0.17 [0.09-0.21] Tg CH₄ yr⁻¹) (Table
180 1, Figure 2). The high CH₄ emissions from mangroves (exceeding global estuaries) are
181 promoted by carbon-rich deep anoxic sediments⁴³ and by tidally induced mixing between
182 groundwater rich in CH₄ and surface waters⁴⁴. We find highest mangrove CH₄ flux densities
183 (21 mg CH₄ m⁻² d⁻¹) in East Asia. However, North America's mangrove forests dominate (41%)
184 global mangrove CH₄ emissions because its forest area is 86 times greater than in East Asia
185 and the CH₄ flux density is similarly high (Supplementary Table S6). Globally, Southeast Asia

186 comprises most mangrove forest area, but the region's much lower CH₄ flux density (1.4 mg
187 CH₄ m⁻² d⁻¹) means that it adds only 6% to global mangrove CH₄ emissions. Such a low
188 regional evasion rate may be explained by high monsoonal rainfall events and short water
189 residence times, resulting in low rates of anaerobic organic matter decomposition⁴⁵. The
190 highest median CH₄ flux densities of any coastal vegetation type are found in East Asian salt
191 marshes (44 mg CH₄ m⁻² d⁻¹), with this region contributing 34% of global salt marsh CH₄
192 emissions despite its relatively small marsh coverage. North America's vast salt marsh areas
193 are the second highest contributor (35%) to global salt marsh emissions. Australasia's seagrass
194 meadows dominate (29%) global seagrass CH₄ emissions, a combination of moderately high
195 seagrass area and a regional flux density that is more than twice the global median
196 (Supplementary Table S6).

197

198 **Estuarine and coastal vegetation N₂O fluxes**

199 We estimate that estuaries globally emit 61 [41-94] Gg N₂O yr⁻¹ (Table 1), which is
200 substantially lower than previous observational and modeling estimates (94 to 5,710 Gg N₂O
201 yr⁻¹)^{4,24,25,46} (Supplementary Table S3). Similar to CO₂ and CH₄, our lower estuary N₂O
202 emissions are partially due to lower estuarine surface area used in this study³⁰, but also reflects
203 the ~80% more sites in our analysis than earlier reviews^{23,46}. We find the highest N₂O emissions
204 in tidal systems and deltas (35 [24-48 Gg N₂O yr⁻¹), but in contrast to CO₂ and CH₄, fjords are
205 the second highest N₂O emitters (16 [12-21] Gg N₂O yr⁻¹), almost double of those of lagoons
206 (10 [6-14] Gg N₂O yr⁻¹). Regionally, North America contributes 27%, 37%, and 77% to global
207 emissions from tidal systems and deltas, lagoons, and fjords, respectively, contributing 25 [17-
208 33] Gg N₂O yr⁻¹ (41%) to global estuary N₂O emissions (Figure 1). Other regions of moderate
209 emissions are South America (14%), Russia (12%), and Southeast Asia (9%), with the
210 remaining regions being only minor contributors (< 5 Gg N₂O yr⁻¹). We further compare our
211 data-driven approach with the recently published mechanistic model of global estuary N₂O
212 emissions²⁵ that was regionalised for comparison purposes. The modelled global emission
213 estimates from all three estuary types (94 Gg N₂O yr⁻¹) fall close to the upper uncertainty bound
214 reported in our study. As such, we find an overall good agreement between the two approaches.
215 Fjord emissions represent the largest relative difference between modelled and data-driven
216 estimates (Table 1).

217

218 Coastal vegetation can be sources or sinks of N₂O to the atmosphere. We find that emissions
219 slightly exceed uptake at the global scale, resulting in 6.3 [0.7-18.3] Gg N₂O yr⁻¹ which is only

220 about 10% of the estuarine emission (Table 1). Globally, mangroves emit 8.8 [4.4-19.3] Gg
221 N₂O yr⁻¹, more than four times than salt marshes (2.1 [-0.8-6.1] Gg N₂O yr⁻¹). Seagrasses take
222 up 4.6 [6.0-3.4] Gg N₂O yr⁻¹, thereby offsetting almost half of the N₂O emitted by salt marshes
223 and mangroves (Figure 2). Nevertheless, large variability can be found in local N₂O fluxes in
224 coastal vegetation. For example, current studies report N₂O uptake from salt marshes in North
225 America⁴⁷ and Europe⁴⁸, while studies from East Asia, mainly China, reveal that this region
226 accounts for a substantial 1.3 [0.2-2.8] Gg N₂O yr⁻¹ (62%) of global salt marsh emissions.
227 Mangroves ecosystems are generally a source of N₂O although some mangrove creeks in
228 Australia have been shown to consume N₂O due to low nitrogen concentrations^{20,21}. Our
229 seagrass N₂O flux database is strongly biased towards data from Australia, where studies
230 mostly suggest N₂O uptake from near-pristine seagrasses⁴⁹ (Supplementary Table S7).

231

232 **Implication for regional and global budgets**

233 Our data-driven meta-analysis reveals that estuarine CO₂-equivalent GHG emissions reduce
234 the coastal vegetation CO₂-equivalent GHG uptake, by 27% using the GWP₂₀ and 23% using
235 the GWP₁₀₀. Estuaries and coastal vegetation therefore are collectively a GHG sink for the
236 atmosphere of 391 [647-76] Tg CO₂e yr⁻¹ or 444 [675-175] Tg CO₂e yr⁻¹ based on the GWP₂₀
237 or GWP₁₀₀, respectively (Table 2, Figure 3). Our quantification of the net GHG sink is broadly
238 consistent with previous assessments based solely on CO₂^{5,7,8}. However, we find that this sink
239 results both from a downward revision of coastal vegetation CO₂ fixation and substantially
240 lower estuarine GHG outgassing, mostly due to the CO₂ uptake by fjords and a reduced
241 estuarine surface area³⁰. The net CO₂ sink of estuaries and coastal vegetation is also
242 significantly compensated by CH₄ and N₂O emissions, which offset 20% and 9% for the
243 GWP₂₀ and GWP₁₀₀, respectively. Our global-scale assessment considerably reduces the
244 uncertainty in the contemporary GHG budget⁵ and emphasises that the combined contribution
245 of coastal vegetation and estuaries on the global radiative balance is a cooling effect (Figure
246 4), in contrast to terrestrial sources² and inland waters^{3,4}. However, our analysis does not
247 address whether human activity has changed the radiative balance of these systems since the
248 pre-industrial period. Importantly, we reveal where estuarine fluxes enhance, partially reduce,
249 or exceed coastal vegetation CO₂ uptake, allowing us to identify regional hotspots of GHG
250 uptake and release (Figure 3). We find that 8 out of 10 coastal regions are a net GHG sink for
251 the atmosphere, regardless of the time horizon considered (Table 2). Using the GWP₁₀₀,
252 Southeast Asia shows the greatest net GHG sink (156 [224-106] Tg CO₂e yr⁻¹) because of its
253 extensive and highly productive tropical mangrove forests and seagrass meadows (Extended

254 Data Figure 1), as well as accommodating relatively few estuaries compared to other regions³⁰.
255 A second regional hotspot for GHG sinks is North America (128 [165-77] Tg CO₂e yr⁻¹) due
256 to its expansive mangrove, salt marsh, and seagrass areas, while also hosting the largest area
257 of CO₂-absorbing fjords globally (40% of which are located in Greenland). In Africa, the large
258 CO₂ uptake by coastal vegetation is partially offset by estuarine GHG outgassing, leaving
259 Africa as the third greatest net GHG sink (93 [134-36] Tg CO₂e yr⁻¹), globally. Australasia
260 and West Asia are moderate GHG sinks (both ~25 Tg CO₂e yr⁻¹), while in East and South Asia,
261 coastal vegetation CO₂ sinks are largely offset by estuarine GHG release, resulting in a
262 combined small net GHG sink of <20 Tg CO₂e yr⁻¹. In Europe and Russia, estuarine GHG
263 outgassing across a large estuarine surface area exceeds uptake from the relatively small area
264 of coastal vegetation, making these regions net GHG sources. As such, our regionalized
265 assessment suggests that in addition to being a net global sink of GHGs for the atmosphere, the
266 sink attributed to estuaries and coastal vegetation is also a common feature in many regions
267 across the world.

268

269 The limitations of our GHG synthesis fall largely into four categories: mapping of ecosystems,
270 GHG flux measurements, spatio-temporal variability, and coupling between coastal vegetation
271 and estuaries. Tidal marsh area used in this study²⁹ is likely underestimated due to the recent
272 surge in restoration efforts resulting in a gain in tidal marsh area⁵⁰ and the lack of global
273 mapping of freshwater tidal marshes. We exclude low salinity (< 0.5) tidal river fluxes because
274 of the difficulty to separate these from non-tidal rivers, and because the biogeochemistry and
275 residence times are distinctly different from brackish estuaries¹⁹. Our coastal GHG budget is
276 focused on the interface of the atmosphere with coastal vegetation and estuaries. We do not
277 account for other radiative effects that may arise from lateral transport, such as offsite
278 emissions, or 'blue carbon' burial in marine sediments. The ebullitive and plant-mediated CH₄
279 flux is underrepresented in our analysis due to the scarcity of such data. GHG flux
280 measurements, despite recent progress, are still lacking, particularly from Africa, Russia, and
281 West Asia. To better capture temporal variability, eddy covariance towers and conventional
282 time-series measurements are needed, even in regions with good spatial coverage. High spatio-
283 temporal variability means that techniques such as remote sensing⁵⁰, empirical modeling, and
284 process-based modeling are needed for extrapolating flux densities and projecting future
285 hotspots, particularly in the face of climate change and population growth along the coast.
286 Coastal vegetation and estuaries are intimately coupled. However, the quantity and quality of
287 material transported laterally and the fraction and mechanisms controlling outgassing through

288 the coastal vegetation-estuary interface or further offshore are currently poorly known⁵¹,
289 particularly for CH₄ and N₂O. Despite its limitations, our coastal GHG synthesis addresses the
290 current research gap between local and global scales and identifies regional hotspots. The
291 future role of coastal ecosystems as a sink or source of GHGs in each world region will depend
292 on the adoption of best practices to reduce CH₄ and N₂O emissions while strengthening the
293 CO₂ uptake.

294

295 **Methods:**

296 Definitions of estuaries (tidal systems and deltas, lagoons, fjords) and coastal vegetation
297 (mangroves, salt marshes, seagrasses) can be found in the Supplementary Information.

298

299 **Estuarine surface areas**

300 The estuarine surface areas used in this study were calculated using a novel regionalized
301 approach that combines available national databases and an extrapolation method that derives
302 the total estuarine surface area of a region from the surface areas of its largest systems³⁰. Using
303 a well-established global coastal segmentation comprising 45 regions (MARgins and
304 CATchment Segmentation, (MARCATS)¹⁰), a surface area was determined for each estuary
305 type (tidal systems and deltas, lagoons, or fjords) in each region. Wherever exhaustive regional
306 or national databases were available and covered an entire MARCATS segment (i.e., Australia,
307 New Zealand, and all lagoons surrounding the Mediterranean Sea), the type-specific surface
308 areas were extracted from these databases by assigning each identified system to a given type.
309 In other regions, the surface areas were extrapolated from the 5 to 10 largest systems. The
310 extrapolation method relies on the observation that the cumulative estuarine surface area
311 expressed as a function of the number of estuaries ranked by decreasing size within a
312 sufficiently large stretch of coastline can be fitted by an equation of the form $S = (a * N) / (b +$
313 $N)$ with S being the total estuarine surface area (in km²), N, the number of estuaries and a and
314 b dimensionless calibration coefficients. Using several extensive national estuarine databases
315 (Australia, New Zealand, United States), it was shown that fitting the parameters of this generic
316 formula using only the 10 largest estuaries of a given region generally allows predicting the
317 total cumulative surface area of the region with a 9 % accuracy. This uncertainty due to
318 extrapolation is complemented by an uncertainty associated with the accuracy in estimating the
319 surface areas of the individual systems used to perform the calculations and which range from
320 4 to 15% depending on the estuarine type. Within each MARCATS, the surface areas of the 10
321 largest systems of each estuarine type were gathered from national databases (United Kingdom,

322 Mexico, United States, Australia, South Africa, South Korea), regional surveys (FAO,
323 UNESCO), global databases (Sea Around Us) or calculated individually using GIS. Those data
324 were then sorted and fitted using the equation described above to derive the estuarine surface
325 area for each MARCATS and each estuarine type, and finally summed to obtain the area for
326 each of the ten RECCAP2 regions. For further details, see Laruelle et al.³⁰.

327

328 **Coastal vegetation surface areas**

329 For all three coastal vegetation types, we segmented the vegetation area given as a global
330 collection of polygons²⁷⁻²⁹ into the larger MARCATS regions¹⁰ using the following approach:
331 The MARCTAS regions were first converted from the raster format to polygons using the
332 python package *rasterio*⁵². To find the area of each coastal vegetation type within a given
333 MARCATS region we first dissolved overlapping individual polygons to avoid overestimation
334 of the vegetation area. We found the union of each polygon and the larger MARCATS region
335 and recalculated the area using the EPSG:6933 projection, using the python package
336 *geopandas*⁵³, and finally summed these values. The same process was repeated for each large-
337 scale MARCATS region and coastal vegetation type individually. The available shapefiles for
338 MARCATS regions were converted from gridded outputs to polygons using *rasterio*⁵². Using
339 *geopandas* for each region, we dissolved overlapping polygons and calculated the area of each
340 polygon within the region using the EPSG:6933 projection, and finally summed the area for
341 each of the ten RECCAP2 regions.

342

343 **Estuarine greenhouse gas fluxes**

344 We calculated estuary water-air fluxes of CO₂, CH₄ and N₂O based on data from peer-reviewed
345 publications until the end of 2020. We conducted a literature search in Google Scholar and data
346 publishers (Fluxnet, PANGAEA, MEMENTO, British Oceanographic Data Centre) using the
347 search string '(CO₂ OR CH₄ OR N₂O) AND (tidal system OR delta OR lagoon OR fjord)'.
348 Additionally, we scanned the reference lists of publications. For each site/study location (> 10
349 km apart), we averaged spatial and seasonal variation to a single water-air flux density,
350 resulting in 204 sites for CO₂, 157 sites for CH₄ and 123 sites for N₂O, globally, which are
351 20%, 21% and 85% more sites compared to previous global syntheses for CO₂⁹, CH₄¹⁷, and
352 N₂O²³, respectively. For estuary CH₄ estimates, we included primarily diffusive water-air
353 fluxes (computed from the gas transfer velocity and the concentrations of CH₄ in water and air)
354 and four studies using floating chamber incubations that estimate the total CH₄ flux (diffusive
355 and ebullitive). The low number of total CH₄ flux estimates means that the ebullitive flux is

356 underestimated in our analysis, particularly in the upper estuarine region. Each site/flux density
357 was then distributed following the ten RECCAP2 regions using lat/long coordinates provided
358 in the original studies or from Google Earth based on the study site description. Sites were
359 categorised as either ‘tidal systems and deltas’, ‘lagoons’, or ‘fjords’ based on the authors’
360 characterisation of their study site. In systems for which the site description was ambiguous or
361 not detailed enough, the type was assigned following the hierarchical steps for estuarine type
362 determination provided by Dürr et al.³¹ which are based on hydrological, lithological, and
363 morphological criteria. Estuary CO₂ and CH₄ emissions have not been modelled at global
364 scales because of the lack of an existing spatially explicit global estuarine model designed to
365 perform such a task. However, we supplemented the literature-derived empirical estuarine
366 estimates of N₂O emissions with results from the spatially explicit global model from Maavara
367 et al.²⁵. These authors used a stochastic-mechanistic model representing generalized nitrogen
368 dynamics for water body types directly connected to river networks worldwide to extract
369 relationships predicting N₂O emissions from water residence times and total nitrogen (TN)
370 yields (i.e., area-normalized loads) delivered to each estuary. The global TN loads were
371 calculated using an approach inspired by Global-NEWS models⁵⁴, and spatially routed nitrogen
372 through an inland water network joined to the estuaries accounted for in Dürr et al.³¹. Emissions
373 factors were calculated using scenarios that reflect existing literature assumptions and datasets.
374 The model was aggregated at a spatial resolution of 0.5 degrees, which is the same used by
375 Dürr et al.³¹ for their estuarine typology. Each watershed thus was assigned an estuarine type
376 based on the type of the coastal cell in which the mouth of the watershed was located. Complete
377 model details are available in Maavara et al.²⁵.

378

379 **Coastal vegetation greenhouse gas fluxes**

380 We conducted a literature search in Google Scholar and data publishers (Fluxnet, PANGAEA,
381 MEMENTO, British Oceanographic Data Centre) until the end of 2020 using the search string
382 ‘(CO₂ OR CH₄ OR N₂O) AND (mangroves OR salt marshes OR seagrasses)’. Additionally,
383 we scanned the reference lists of publications. For each site/study location (> 10 km apart), we
384 averaged spatial and seasonal variation to a single flux density, resulting in 37 sites for CO₂,
385 162 sites for CH₄ and 55 sites for N₂O in coastal vegetation (Extended Data Figure 3), which
386 are 15% and 42% more sites compared to previous global syntheses of coastal wetlands for
387 CH₄¹⁷ and N₂O²³, respectively. In our global synthesis, we use only long-term eddy-
388 covariance (EC) CO₂ fluxes for coastal vegetation. We exclude chamber measurements as they
389 do not provide information on CO₂ ecosystem exchange with a temporal coverage that is

390 comparable to EC measurements. Although EC towers are still relatively rare in coastal
391 vegetation, this method provides robust high-resolution long-term monthly, seasonal, and often
392 annual data of net ecosystem exchange. EC measurements of CH₄ and N₂O are extremely rare
393 in coastal vegetation. Therefore, we combined the few existing EC data with water-air,
394 sediment-air, sediment-water, plant (leaf, root, stem)-air, and ebullition CH₄ or N₂O fluxes
395 where available.

396

397 **CO₂-equivalent greenhouse gas fluxes**

398 For the purpose of directly comparing the relative radiative forcing effects of CO₂, CH₄, and
399 N₂O to one another, we express GHG fluxes as CO₂-equivalents for the 20-year and 100-year
400 time horizons using the global warming potential (GWP₂₀ and GWP₁₀₀) including chemical
401 adjustments reported in the Sixth Assessment Report of the Intergovernmental Panel on
402 Climate Change⁵⁵. For CH₄, we use the non-fossil GWP₂₀ and GWP₁₀₀. Accordingly, over a
403 20-year time period, which is important for climate policies based on shorter timescales, 1 kg
404 of CH₄ and N₂O have the same global warming potential as 79.9 kg or 273 kg of CO₂,
405 respectively. Over a 100-year time period, 1 kg of CH₄ or N₂O have the same global warming
406 potential as 27.0 kg or 273 kg of CO₂, respectively. The net contemporary GHG radiative
407 balance for a given region, ecosystem, and time horizon is the sum of the CO₂-equivalents over
408 the three gases.

409

410 **Data analyses and upscaling**

411 Data processing and statistical analyses were performed in R software⁵⁶. A non-parametric
412 bootstrapping method using the package *boot* in R was applied to resample flux densities for
413 each of the 3 gases in each of the 6 ecosystems (3 estuary types and 3 coastal vegetation types)
414 and in each of the 10 RECCAP2 regions (Extended Data Figure 4). The bootstrapping method
415 used 1,000 iterations of the median of samples to produce a smoothed distribution of flux
416 densities, and to generate a full set of statistics. Results from non-parametric bootstrapping
417 were then multiplied by the corresponding surface area of each of the 6 ecosystem types in
418 each of the 10 RECCAP2 regions. If an ecosystem type had less than three sites in a region,
419 we applied the global statistics of this type in this region.

420

421 **Uncertainty analysis**

422 Uncertainties associated with flux densities were derived from the bootstrapping method that
423 generated a full set of statistics including the first (Q1) and third (Q3) quartiles of the datasets.

424 For estuarine surface areas, the normality of the distribution was successfully tested and
425 allowed deriving upper and lower 95% confidence intervals (c.i.) as well as the first (Q1) and
426 third (Q3) quartiles from the standard deviation³⁰. Global mangrove forest has been
427 successfully mapped with an overall accuracy of 99% likelihood that the true value is between
428 93.6-94.5%²⁷. The global extent of salt marshes is relatively uncertain with recent global
429 estimates ranging from 55,000²⁸ to 90,800 km² ⁵⁰. Submerged seagrasses are challenging to
430 map and current global estimates are highly uncertain. The most recent and continuously
431 updated global seagrass mapping effort by UNEP-WCMC and Short²⁹ does not estimate
432 uncertainty. Here, we assume an area per region uncertainty of 5%, 10%, and 20% for
433 mangroves, salt marshes, and seagrasses, respectively, noting that these uncertainties could be
434 substantially larger for the latter two types of coastal vegetation. We then combined the
435 respective uncertainties of GHG flux densities with uncertainties of surface area of either
436 estuaries or coastal vegetation using the ‘Root Sum of the Squares’ method, which calculates
437 the square-root of the linear sum of the squared standard uncertainty components, treating the
438 uncertainty contributors as statistically independent. We present the combined uncertainties of
439 Q1 and Q3 in accordance with the median (Q2) in the main article, because this statistical set
440 represents the most appropriate measure of the non-normally distributed GHG fluxes.

441

442 **Data Availability:**

443 All data included in this study are freely available
444 (<https://doi.org/10.6084/m9.figshare.22351267> and ref 57). Data may be used if cited
445 appropriately.

446

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468

469 **Author Contribution Statement:**

470 J.A.R. and P.R. conceived and designed the study. J.A.R. did the synthesis for CH₄ and N₂O
471 in estuaries, and for CH₄ and N₂O in coastal vegetation. B.D.E. and H.W.B. helped with the
472 synthesis for N₂O in estuaries and coastal vegetation. T.M. provided the mechanistic model for
473 N₂O in estuaries. G.G.L. did the synthesis for CO₂ in estuaries. I.F. did the synthesis for CO₂
474 in salt marshes. B.V.D. did the synthesis for CO₂ in seagrasses. J.A.R. did the synthesis for
475 CO₂ in mangroves. J.J.M.B. segmented the coastal vegetation surface area. G.G.L. segmented
476 the estuarine surface area. J.A.R. produced the results and figures and wrote the original draft
477 of the paper. All authors helped with the interpretation of the data and contributed to the review
478 and editing of the paper.

479

480 **Competing Interest Statement:**

481 The authors declare no competing interests.

482

483 **Table 1. Global estuary and coastal vegetation greenhouse gas fluxes.** Global median [Q1-Q3] estuary (tidal systems and deltas, lagoons,
484 fjords) and coastal vegetation (mangroves, salt marshes, seagrasses) CO₂, CH₄, N₂O, and CO₂-equivalent (CO₂e) greenhouse gas fluxes using the
485 global warming potential for the 20-year (GWP₂₀) and 100-year (GWP₁₀₀) time periods. Estuary N₂O fluxes are also shown based on the
486 mechanistic model (output) by Maavara et al.²⁵.
487

Ecosystem	CO ₂ Tg CO ₂ yr ⁻¹	CH ₄ Tg CH ₄ yr ⁻¹	CH ₄ Tg CO ₂ e yr ⁻¹ GWP ₂₀	CH ₄ Tg CO ₂ e yr ⁻¹ GWP ₁₀₀	N ₂ O Gg N ₂ O yr ⁻¹	N ₂ O Tg CO ₂ e yr ⁻¹ GWP ₂₀ and GWP ₁₀₀
Tidal systems and deltas	126.6 [89.9-175.5]	0.14 [0.11-0.23]	10.8 [8.66-18.1]	3.66 [2.94-6.13]	35.1 [23.6-48.4], (49.7)	9.58 [6.45-13.2]
Lagoons	50.0 [27.8-80.5]	0.11 [0.05-0.13]	8.53 [4.15-10.1]	2.89 [1.41-3.43]	9.93 [6.24-13.8], (14.1)	2.71 [1.70-3.76]
Fjords	-65.6 [-96.4- -23.5]	0.003 [0.002-0.008]	0.25 [0.15-0.62]	0.08 [0.05-0.21]	16.3 [11.8-20.6], (30.5)	4.46 [3.21-5.61]
Global estuaries	111.1 [72.7-170.3]	0.25 [0.07-0.46]	19.6 [5.45-36.9]	6.63 [1.85-12.5]	61.3 [41.4-93.8], (94.4)	16.8 [11.3-25.6]
Mangroves	-359.3 [-433.5- -333.4]	0.34 [0.21-0.50]	26.8 [16.4-39.8]	9.07 [5.57-13.5]	8.85 [4.36-19.3]	2.42 [1.19-5.27]
Salt marshes	-49.6 [-59.0- -41.2]	0.26 [0.14-0.36]	20.3 [11.2-28.3]	6.89 [3.80-9.60]	2.06 [-0.83-6.08]	0.56 [-0.23-1.66]
Seagrasses	-191.6 [-353.5- -53.2]	0.17 [0.09-0.21]	13.7 [7.04-16.9]	4.64 [2.38-5.73]	-4.56 [-5.98- -3.39]	-1.25 [-1.63- -0.93]
Global coastal vegetation	-600.6 [-773.5- -426.2]	0.77 [0.47-1.41]	60.8 [37.4-112.5]	20.6 [12.7-38.1]	6.35 [0.70-18.3]	1.73 [0.19-5.00]
Global estuaries + coastal vegetation	-489.5 [-700.8- -255.9]	1.01 [0.54-1.87]	80.4 [42.9-149.4]	27.2 [14.5-50.6]	67.7 [42.1-112.1]	18.5 [11.5-30.6]

488 **Table 2. Regional coastal CO₂-equivalent greenhouse gas fluxes.** Median [Q1-Q3] estuary and coastal vegetation CO₂-equivalents (CO₂e)
 489 greenhouse gas fluxes using the global warming potential for the 20-year (GWP₂₀) and 100-year (GWP₁₀₀) time periods. Negative values indicate
 490 net uptake of greenhouse gases from the atmosphere and positive values indicate net release of greenhouse gases to the atmosphere.
 491

Region	Estuaries		Coastal vegetation		Estuaries + coastal vegetation	
	Tg CO ₂ e yr ⁻¹ (GWP ₂₀)	Tg CO ₂ e yr ⁻¹ (GWP ₁₀₀)	Tg CO ₂ e yr ⁻¹ (GWP ₂₀)	Tg CO ₂ e yr ⁻¹ (GWP ₁₀₀)	Tg CO ₂ e yr ⁻¹ (GWP ₂₀)	Tg CO ₂ e yr ⁻¹ (GWP ₁₀₀)
North America	-20.1 [-33.6- -5.97]	-25.7 [-33.9- -12.1]	-89.4 [-123.9- -43.0]	-102.0 [-131.2- -64.6]	-109.6 [-157.5- -49.0]	-127.7 [-165.1- -76.7]
South America	13.3 [7.68-26.1]	11.6 [6.54-22.7]	-48.5 [-66.1- -26.9]	-51.1 [-66.1- -32.8]	-35.3 [-58.4- -0.79]	-39.5 [-59.5- -10.1]
Europe	31.3 [28.5-36.2]	30.5 [27.9-34.9]	-14.5 [-19.5- -5.72]	-15.1 [-19.9- -6.46]	16.8 [8.98-30.5]	15.4 [8.07-28.5]
Africa	25.1 [13.4-44.9]	23.3 [13.4-37.3]	-110.5 [-143.4- -56.6]	-116.0 [-147.1- -73.4]	-85.4 [-130.0- -11.7]	-92.6 [-133.8- -36.1]
Russia	30.4 [25.7-41.8]	29.3 [25.4-38.2]	-3.77 [-6.08- -1.17]	-5.23 [-7.05- -3.04]	26.6 [19.6-40.6]	24.1 [18.3-35.2]
West Asia	1.02 [0.91-1.28]	0.97 [0.86-1.23]	-22.8 [-29.4- -12.7]	-23.5 [-29.8- -15.1]	-21.8 [-28.5- -11.4]	-22.5 [-28.9- -13.9]
East Asia	7.59 [6.24-10.4]	7.21 [6.03-9.80]	-10.1 [-18.0- -3.09]	-15.1 [-21.8- -9.81]	-2.51 [-11.8-7.27]	-7.88 [-15.8- -0.01]
South Asia	12.6 [10.6-13.9]	12.5 [10.5-13.8]	-21.5 [-30.7- -8.48]	-24.2 [-31.9- -13.5]	-8.93 [-20.1-5.47]	-11.7 [-21.4-0.34]
Southeast Asia	12.5 [7.97-21.4]	12.3 [7.78-21.2]	-164.9 [-229.2- -122.8]	-168.0 [-231.5- -127.0]	-152.4 [-221.2- -101.4]	-155.8 [-223.7- -105.8]
Australasia	33.7 [22.1-42.8]	32.5 [21.4-41.4]	-51.9 [-69.7- -28.2]	-58.0 [-74.3- -37.4]	-18.2 [-47.6-14.6]	-25.5 [-53.0-3.95]
Global	147.4 [89.5-232.8]	134.5 [85.9-208.4]	-538.0 [-735.9- -308.7]	-578.2 [-760.7- -383.1]	-390.6 [-646.5- -75.8]	-443.8 [-674.8- -174.6]

492

493 **Figure Captions (main text):**

494

495 **Figure 1: Regional and global estuary GHG fluxes.** The box-whisker plots show the median
496 and interquartile (Q1-Q3) range of CO₂, CH₄, and N₂O fluxes in tidal systems and deltas (TD),
497 lagoons (LA), fjords (FJ) in ten RECCAP2 regions and globally. A positive value indicates a
498 flux from the ecosystem to the atmosphere and a negative value a flux from the atmosphere to
499 the ecosystem. Outliers (open circles) are shown only for global plots. The asterisk (*) indicates
500 that the flux was upscaled based on global statistics (n < 3) (see Methods).

501

502 **Figure 2: Regional and global coastal vegetation GHG fluxes.** The box-whisker plots show
503 the median and interquartile (Q1-Q3) range of CO₂, CH₄, and N₂O fluxes in mangroves (MA),
504 salt marshes (SM), and seagrasses (SE) in ten RECCAP2 regions and globally. A positive value
505 indicates a flux from the ecosystem to the atmosphere (emission) and a negative value a flux
506 from the atmosphere to the ecosystem (uptake). Outliers (open circles) are shown only for
507 global plots. The asterisk (*) indicates that the flux was upscaled based on global statistics (n
508 < 3) (see Methods)

509

510 **Figure 3: Global and regional coastal CO₂-equivalent GHG fluxes.** a) Global median GHG
511 sources and sinks of estuaries and coastal vegetation using the global warming potential for the
512 20-year (GWP₂₀, top) and 100-year (GWP₁₀₀, bottom) time horizons. The radiative balance is
513 expressed as CO₂-equivalents in Tg CO₂e yr⁻¹ for CO₂, CH₄, N₂O, and their sum (net GHGs).
514 b) Conceptual diagram showing the magnitude of the CO₂-equivalent GHG fluxes for estuaries,
515 coastal vegetation, and their sum in the ten RECCAP2 regions and relative to each other. The
516 size of the arrows is proportional to the fluxes in Tg CO₂e yr⁻¹ (Table 2). The arrow size is the
517 same for GWP₂₀ and GWP₁₀₀, therefore arrows are not shown for the two GWP time horizons
518 individually.

519

520 **Figure 4: Estuary and coastal vegetation GHG fluxes in the land-to-ocean aquatic**
521 **continuum (LOAC).** a) the LOAC carbon loop model is adapted from Regnier et al.⁵. The
522 green loop connects the terra-firme ecosystems to the open ocean. The two shorter loops
523 connect the terra firme ecosystems to inland waters (grey) and the coastal vegetation, estuaries,
524 and continental shelves to the open ocean (blue). F_{AW} refers to GHG fluxes by tidal wetlands
525 (salt marshes and mangroves) and submerged vegetation and F_{EA} refers to estuarine GHG
526 fluxes. b) Global median [Q1/Q3] CO₂-equivalent GHG fluxes using the global warming
527 potential for the 20-year (GWP₂₀) and 100-year (GWP₁₀₀) time horizons in coastal vegetation
528 and in estuaries. Here we include seagrasses in coastal vegetation fluxes (F_{AW}), whereas in
529 Regnier et al.⁵ seagrasses as submerged vegetation are separated from inter-tidal emergent
530 mangroves and salt marshes. Units are in Tg CO₂e yr⁻¹.

531

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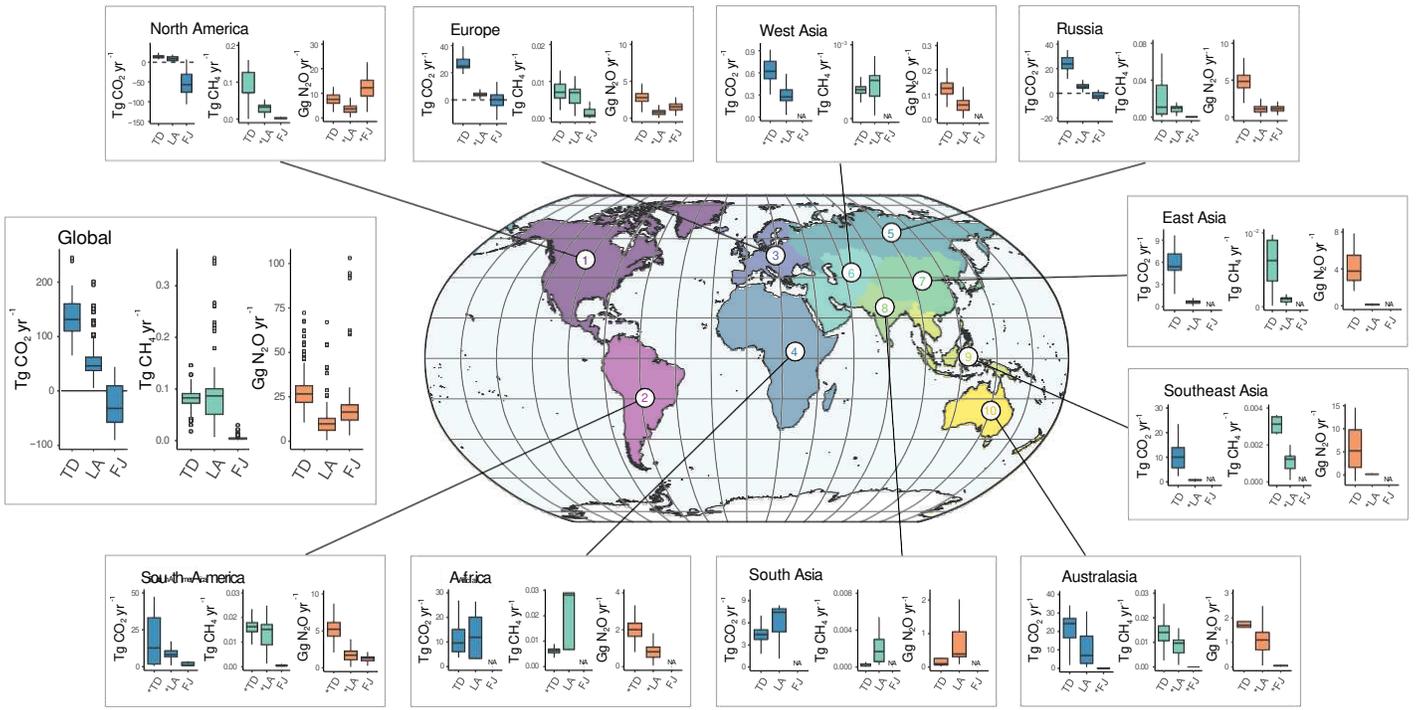
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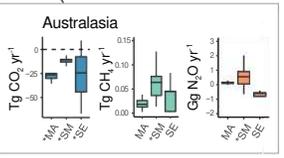
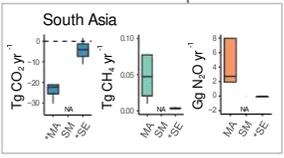
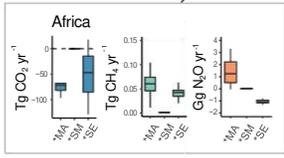
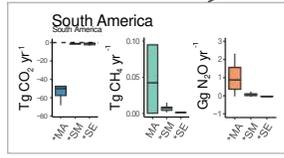
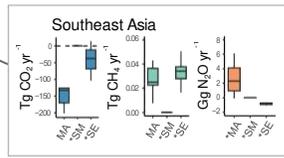
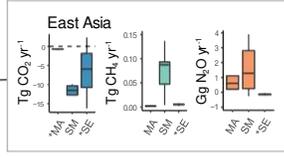
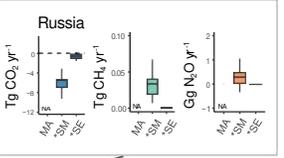
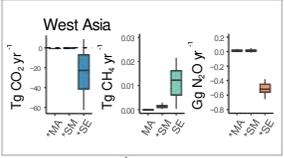
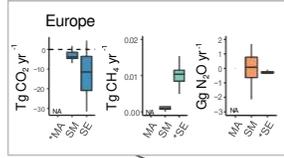
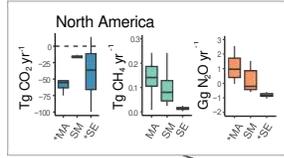
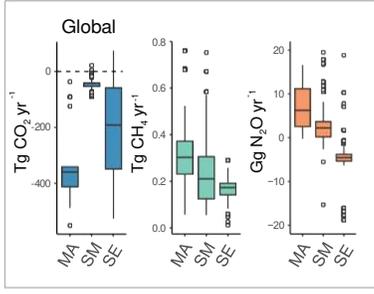
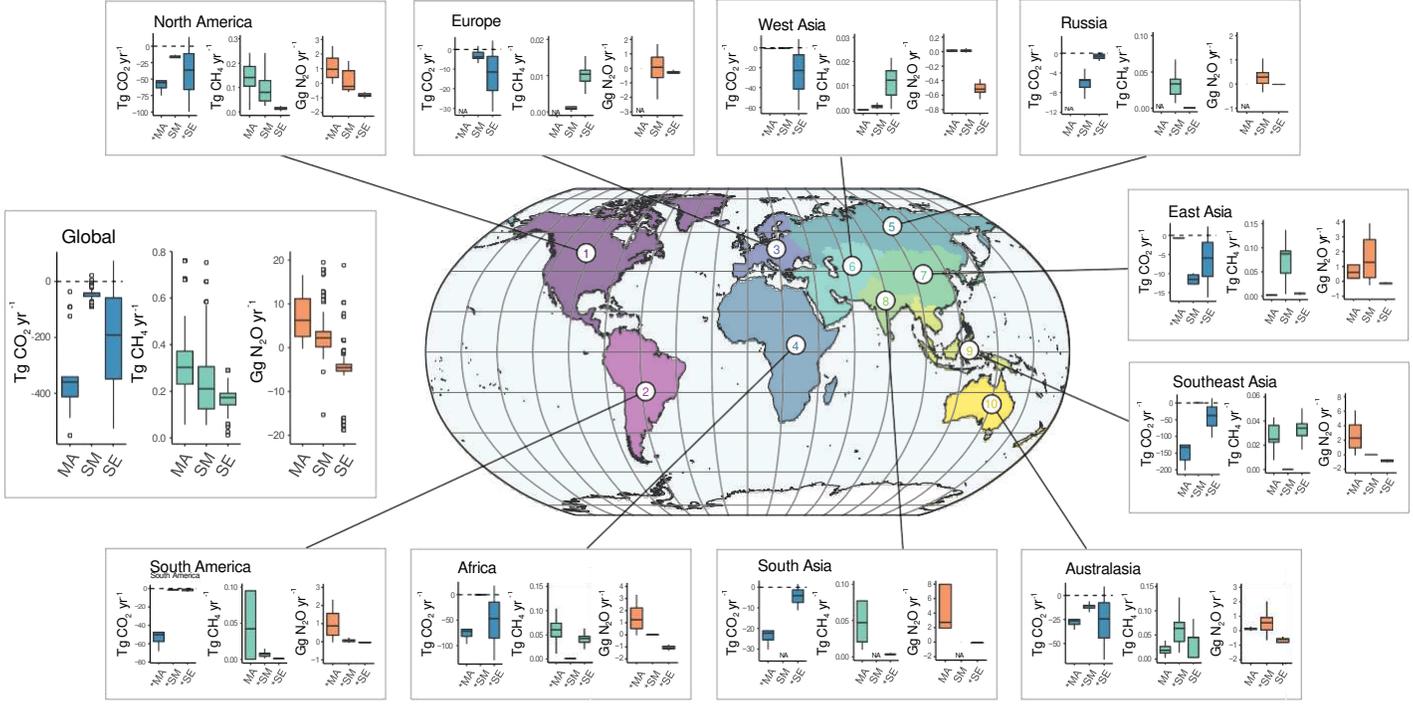
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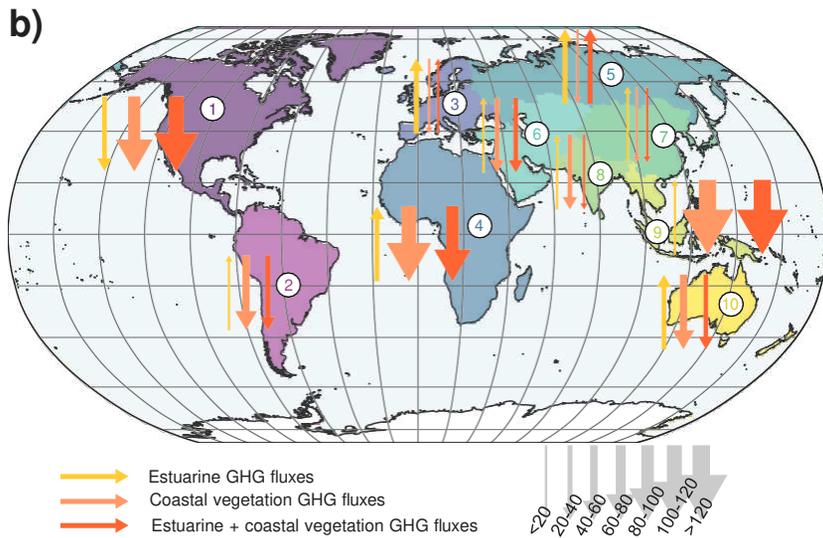
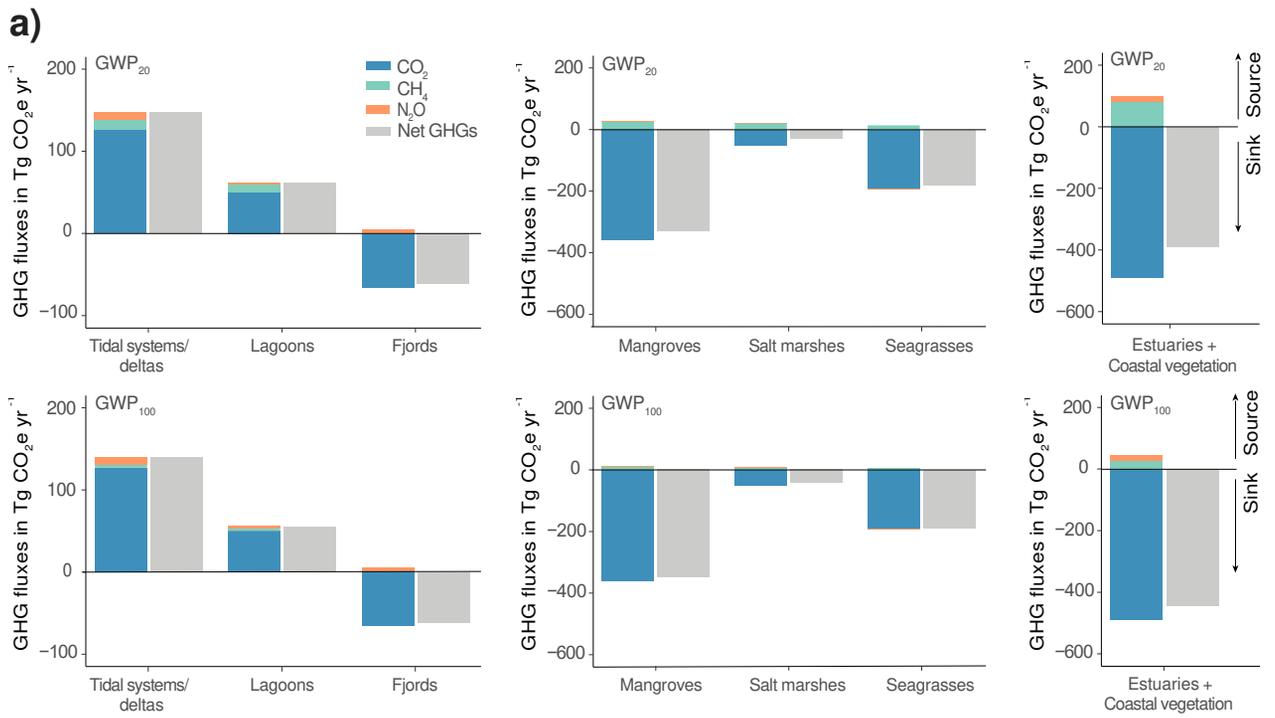
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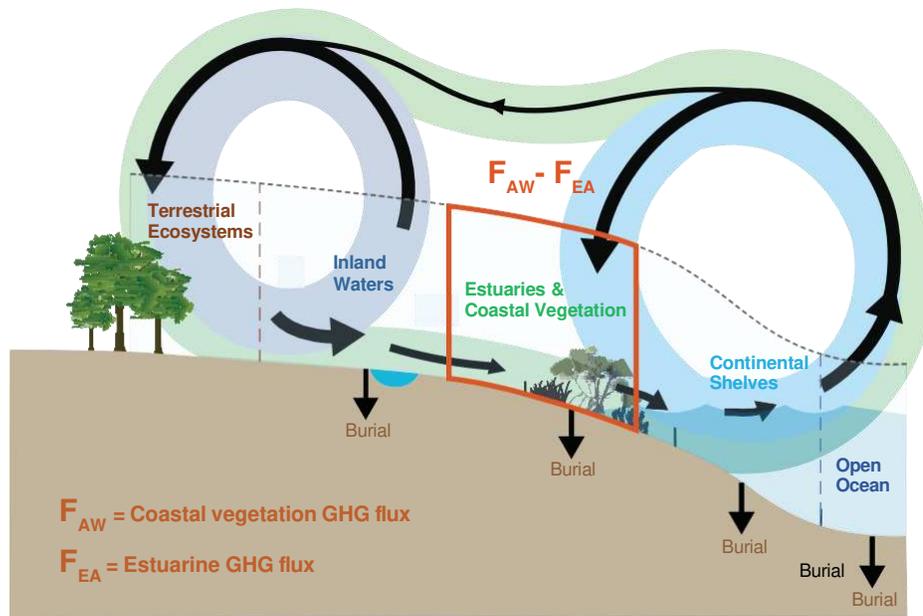
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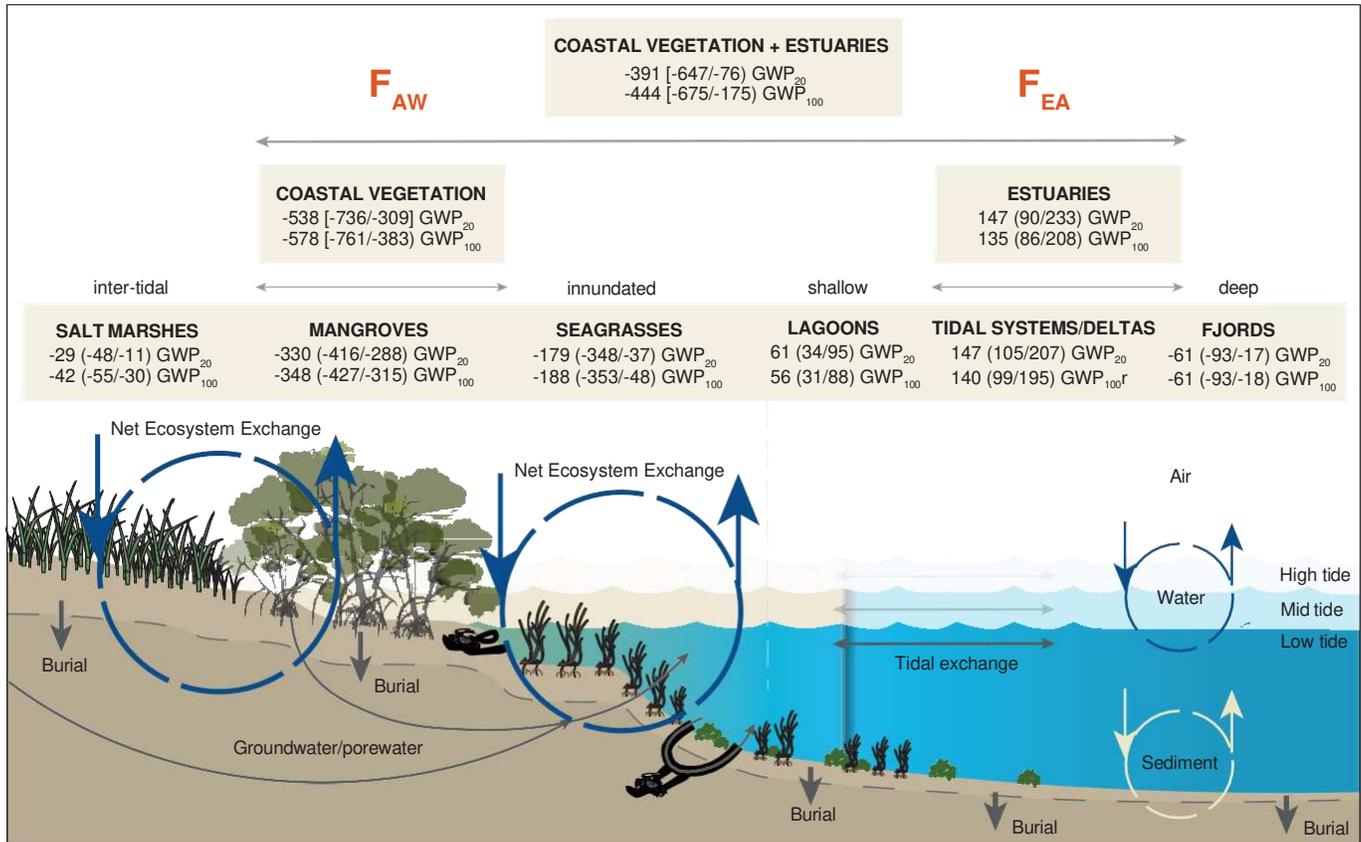


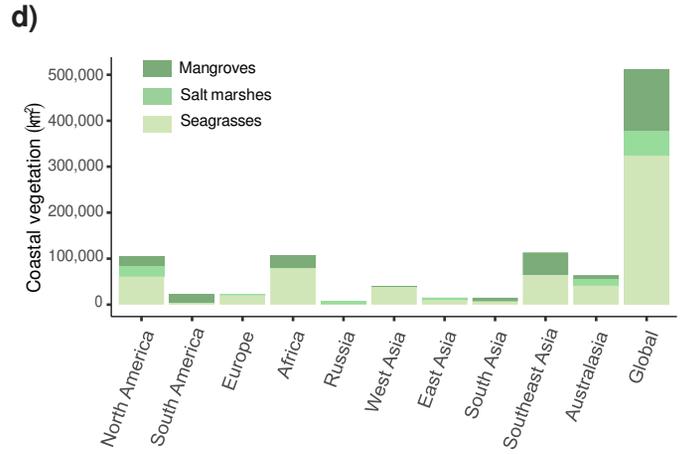
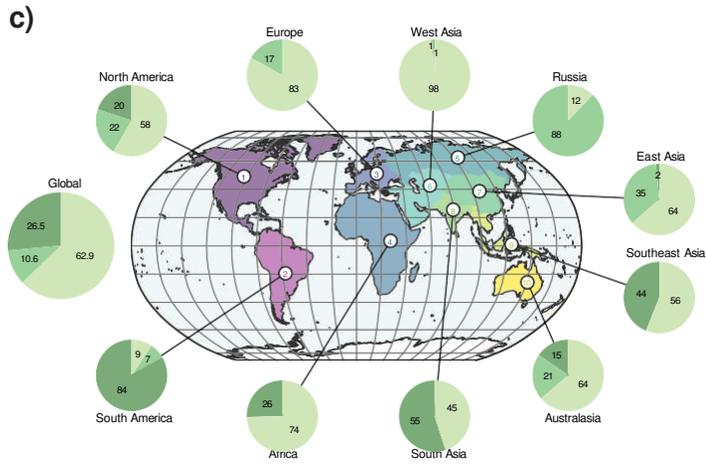
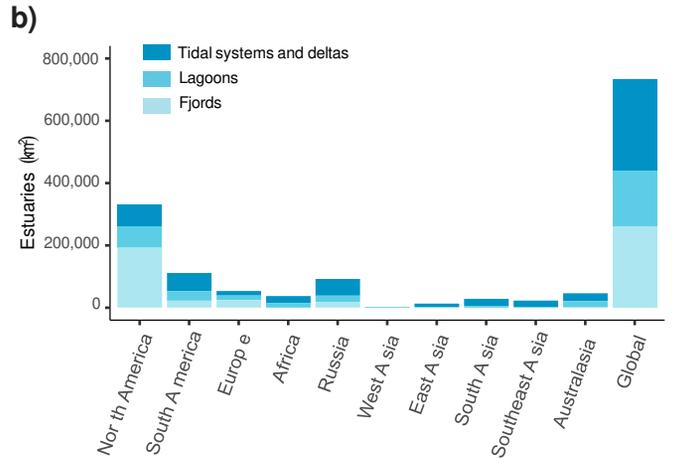
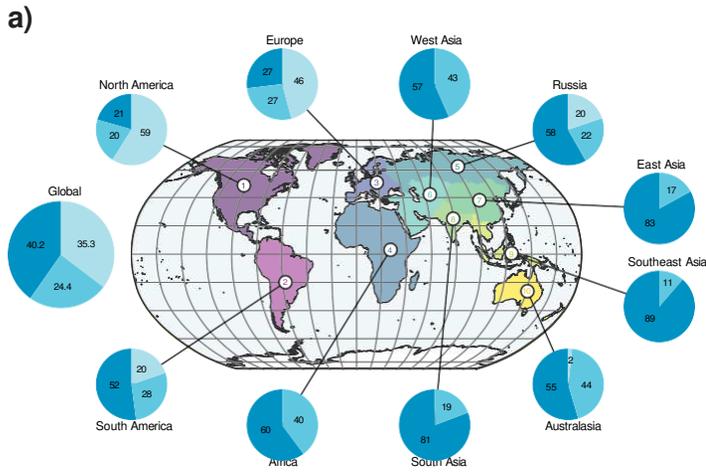


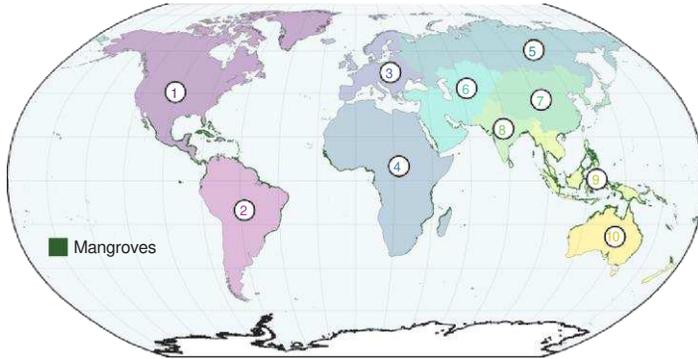
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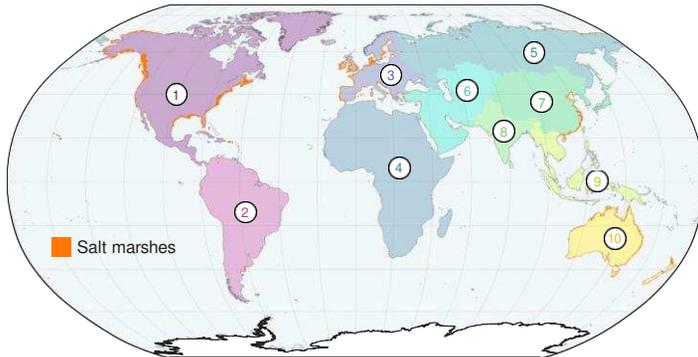






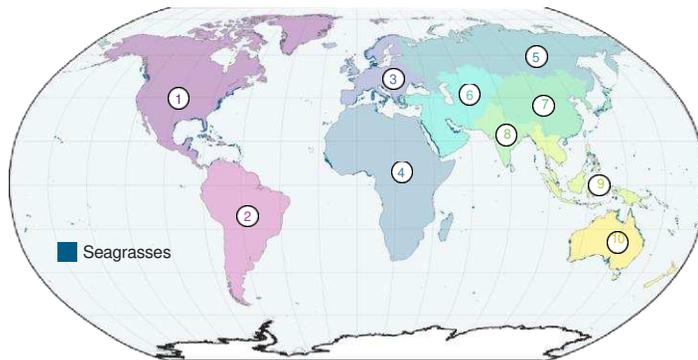
Mangrove forest (km²)

1 North America	= 20,910
2 South America	= 18,990
3 Europe	= 0
4 Africa	= 27,086
5 Russia	= 0
6 West Asia	= 267
7 East Asia	= 243
8 South Asia	= 8,402
9 Southeast Asia	= 49,967
10 Australasia	= 9,948



Salt marshes (km²)

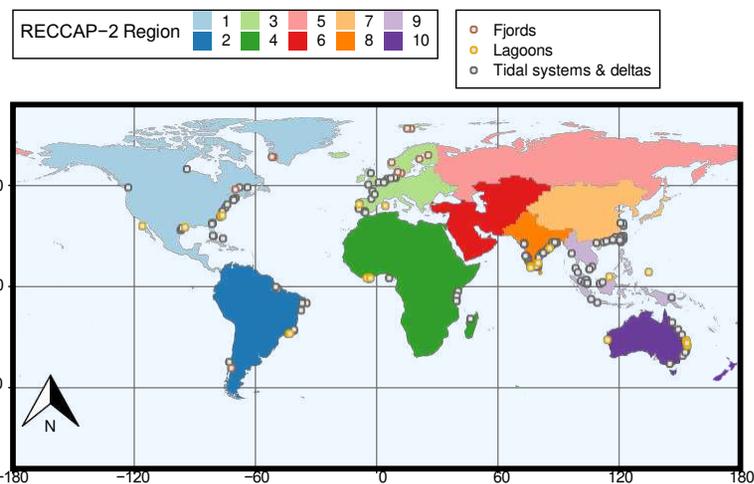
1 North America	= 22,672
2 South America	= 1,564
3 Europe	= 3,975
4 Africa	= 119
5 Russia	= 7,044
6 West Asia	= 318
7 East Asia	= 5,475
8 South Asia	= 0
9 Southeast Asia	= 0.1
10 Australasia	= 13,383



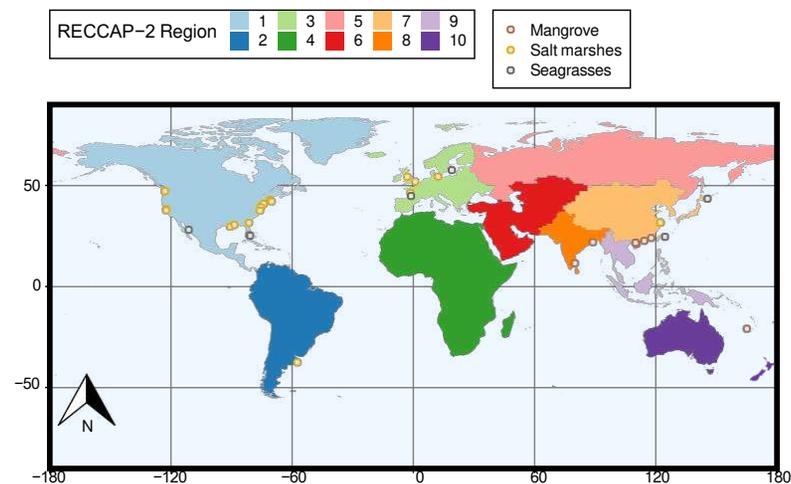
Seagrass meadows (km²)

1 North America	= 61,110
2 South America	= 2,136
3 Europe	= 19,365
4 Africa	= 79,201
5 Russia	= 961
6 West Asia	= 38,352
7 East Asia	= 9,984
8 South Asia	= 6,873
9 Southeast Asia	= 63,631
10 Australasia	= 41,009

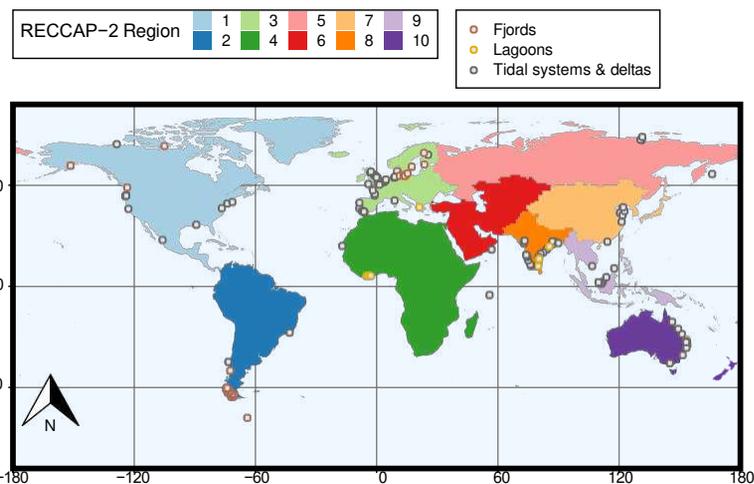
Estuary CO₂ Sites



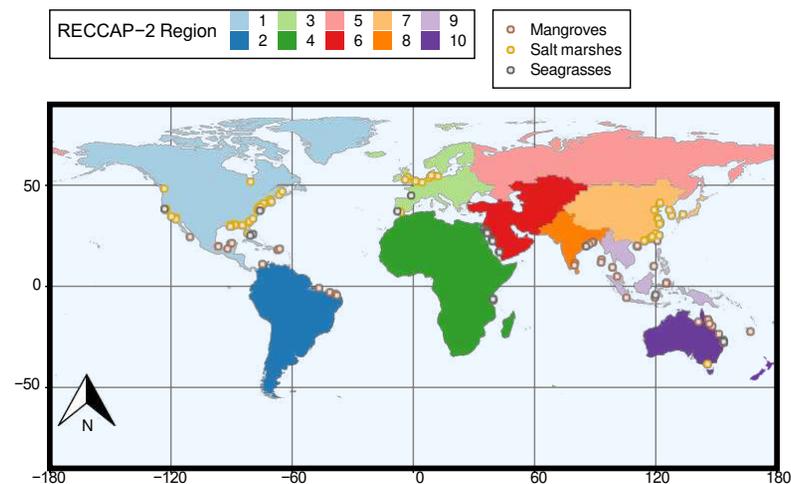
Coastal Vegetation CO₂ Sites



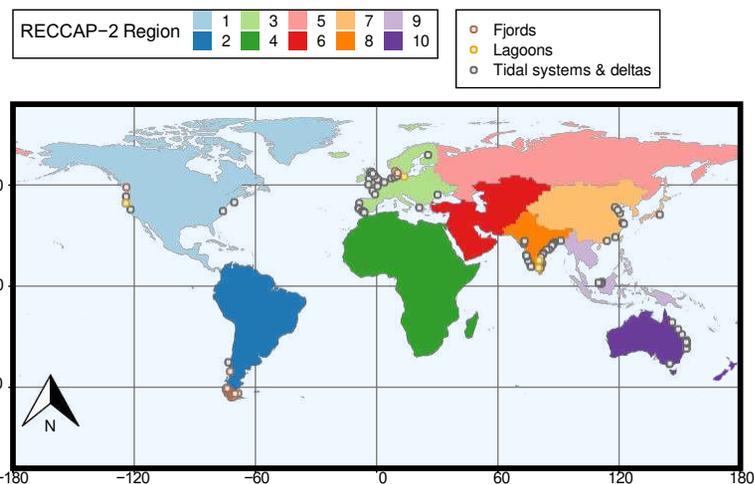
Estuary CH₄ Sites



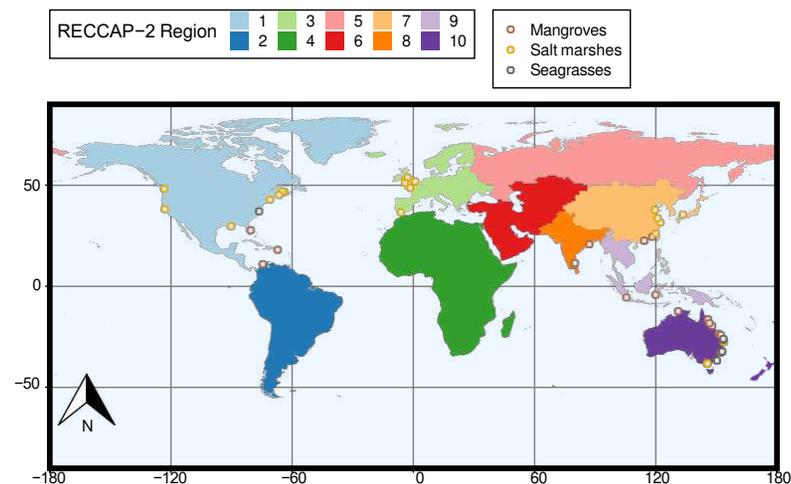
Coastal Vegetation CH₄ Sites

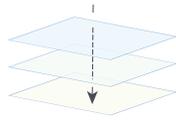


Estuary N₂O Sites



Coastal Vegetation N₂O Sites





Level 1 - Greenhouse gas (CO₂, CH₄, N₂O) x3

Level 2 - Estuary type (tidal systems and deltas, lagoons, fjords) x3 OR
Coastal vegetation type (mangroves, salt marshes, seagrasses) x3

Level 3 - RECCAP2 region x10