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1 **Title: Conservation slows down emission increase from a tropical peatland in**
2 **Indonesia**

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22 **Abstract**

23 **Tropical peatlands are threatened by climate and land-use changes, but there**
24 **remain substantial uncertainties about their present and future role in the global**
25 **carbon cycle due to limited measurements. Here, we present measurements of CO₂**
26 **and CH₄ emissions between mid-2017 and mid-2020 as well as N₂O emissions**
27 **between 2019 and 2020 at two contrasting sites in a coastal peatland in Sumatra,**
28 **Indonesia. We find that greenhouse gas emissions from intact peatlands increased**
29 **significantly due to an extreme drought caused by a positive Indian Ocean Dipole**
30 **phase combined with El Niño. The emission in the degraded site was two times**
31 **greater than at the intact site. The smaller emission in the intact peatland suggests**
32 **that protecting the remaining intact tropical peatlands from degradation offers**
33 **significant climate benefits, avoiding greenhouse gas emissions of 24 ± 5 tCO₂e**
34 **ha⁻¹ yr⁻¹ (average \pm standard deviation) at our study site in Indonesia.**

35 Tropical peatlands have been one of the most important global sinks of atmospheric
36 carbon dioxide (CO₂) over millennia and have accumulated at least 75 Gt carbon under
37 anoxic water-logged conditions¹⁻⁶. However, they are vulnerable to climate change⁷⁻⁹,
38 especially responses to the hydrologic cycle¹⁰⁻¹³. Thus, variability and change in rainfall
39 regime are important factors determining peat carbon accumulation and loss¹¹⁻¹³. Most
40 tropical peatlands have formed since the Last Glacial Maximum¹⁴⁻¹⁶. In Southeast Asia,
41 which holds one of the world's largest tropical peatland areas⁵, the coastal peatlands were
42 initiated following a sea level high-stand, coupled with ample year-round rainfall and low
43 intensity and frequency of droughts^{11,14-16}. In recent times, more frequent and severe El
44 Niño–Southern Oscillation (ENSO)¹⁷ and the linked positive phase of Indian Ocean Dipole

45 (IOD)¹⁸ may change local hydrology in this region^{19,20}. Increasing rainfall seasonality
46 lowers dry season groundwater level (GWL) and permits oxidation of previously stored
47 carbon¹⁰⁻¹³. Yet almost no high-quality contemporary CO₂ flux measurements exist for the
48 remaining intact tropical peatlands in Southeast Asia. Previous research in Borneo^{21,22}
49 reported flux measurements from peatlands disturbed by historical forest cover loss due
50 to selective logging and these data do not necessarily represent an intact reference.
51 Given the role of intact tropical peatlands in long-term climate mitigation through carbon
52 sequestration, an improved understanding of their fate under current and future climate
53 is a prerequisite for assessing the significance of peatland conservation as a climate
54 mitigation strategy^{23,24}.

55 Tropical peatlands are among the world's most threatened ecosystems due to land-cover
56 changes driven by transmigration, population growth and ongoing economic
57 development^{25,26}. Large-scale artificial drainage exposes previously accumulated peat
58 carbon to oxygen and promotes aerobic decomposition, resulting in CO₂ emissions and
59 subsidence of the peatland surface^{27,28}.

60 In addition, tropical peatlands also emit methane (CH₄)²⁹ and nitrous oxide (N₂O)³⁰, potent
61 greenhouse gases (GHGs)³¹ which may be modified by land-cover change^{29,30}. However,
62 assessments of full GHGs budgets are scarce and CH₄ and N₂O contributions remain
63 highly uncertain. Tropical peatlands are reported to be GHG sources²³ in the order of 1.48
64 GtCO₂e yr⁻¹, but with a wide uncertainty range of 0.04-2.79 GtCO₂e yr⁻¹, reflecting a lack
65 of comprehensive quantitative understanding of emissions from these ecosystems. Given
66 the sensitivity of peat GHG emissions to changes in hydrology and/or vegetation, we
67 cannot adequately capture their global climate impact, manage peatlands responsibly or

68 optimize mitigation measures if we do not know the scale of GHG emissions from tropical
69 peatlands^{32,33}.

70 Here, we report on a paired of eddy covariance (EC) study that measured continuous net
71 ecosystem CO₂, CH₄ and H₂O (evapotranspiration) exchanges for seven site-years over
72 two land-cover types within the same peat landscape on the Kampar Peninsula in
73 Sumatra, Indonesia: (1) an intact peatland and (2) a degraded peatland with disturbed
74 hydrology and forest cover loss (Figure 1, Extended Data Table 1). Such comparative
75 studies are rare for peatlands globally (see Methods for a detailed description). Soil N₂O
76 fluxes from manual soil flux chamber measurements were also incorporated to obtain
77 comprehensive GHG budgets. In addition, we also measured peat subsidence around
78 the EC towers. Finally, we linked measured rainfall, evapotranspiration and GWL data to
79 the GHG budget.

80 **Significant CO₂ emissions from an intact peatland**

81 During the study period, June 2017-May 2020, the average GWL was -0.27 ± 0.23 m
82 (average \pm standard deviation, negative sign indicates that the GWL was below the hollow
83 peat surface). The intact peatland emitted 15.5 ± 8.8 tCO₂ ha⁻¹ yr⁻¹ and subsided $3.3 \pm$
84 0.7 cm yr⁻¹ (Table 1, Extended Data Table 2).

85 Net ecosystem CO₂ exchange showed a clear seasonal pattern corresponding to GWL
86 fluctuation (Figure 2). GWL rose up to 0.19 m above the hollow peat surface and remained
87 above the surface for only 16% of the study period. The intact peatland showed a monthly
88 CO₂ uptake of 1.3 ± 0.4 tCO₂ ha⁻¹ when the peat surface was inundated in January 2018
89 but a monthly CO₂ emission of 3.8 ± 0.7 tCO₂ ha⁻¹ in October 2019 when GWL fell to 0.78

90 m below the hollow peat surface. GWL drawdown coincided with an extended dry season
91 when a major positive IOD combined with an El Niño event. Over a 3-month period (July–
92 September 2019), there was only 89 mm of rainfall (Figure 2a, Extended Data Figure 1).
93 Such deep GWL was also observed in an undrained tropical peatland in Kalimantan
94 during the 2015 El Niño event¹³. The results indicate that during the measurement period
95 (which included years with and without El Niño and positive IOD, Extended Data Figure
96 1), large CO₂ emissions during the dry seasons were not entirely offset by relatively small
97 CO₂ uptake during the wet seasons. The climate sensitivity of the ecosystem was also
98 demonstrated by interannual variations in the strength of dry and wet seasons, which led
99 to corresponding variations in the duration and rate of wet season CO₂ uptake and dry
100 season CO₂ emissions (Figure 2c). Notably, during the extended dry season from July
101 2019 through February 2020, rainfall was less than half compared to the same period in
102 previous years (676 vs. 1385 mm) but evapotranspiration remained relatively constant
103 (1014 vs. 1046 mm), leading to almost five times lower GWL as compared to the previous
104 years (-0.51 vs. -0.11 m). This comparatively low GWL resulted in four times more CO₂
105 emissions than in the previous years during the same months (19.9 vs. 5.7 tCO₂ ha⁻¹)
106 (Figure 2c).

107 Observed net CO₂ emissions is driven by changes in ecosystem respiration (R_{eco}). R_{eco}
108 had a negative relationship with GWL (a proxy that captures both oxygen and water
109 availability controls on peat respiration) (Extended Data Figure 2) as also reported in
110 previous studies²¹. The gross primary production (GPP), an indicator of photosynthetic
111 capacity of the vegetation community, also appeared to decrease with GWL drawdown
112 (Extended Data Figure 2c). Notably, the intact peatland showed higher light-use efficiency

113 when GWL was above -0.2 m as compared to deeper GWL (Extended Data Figure 2d).
114 In an intact tropical peatland³⁴, significantly higher root biomass in elevated mounds and
115 shallow lateral rooting mean that approximately 83% of the total live root biomass lies
116 within the top 0.25 m of peat layer. Thus, we surmise that when GWL falls below a critical
117 depth, there is inadequate vertical recharge of near-surface peat layers through capillary
118 rise owing to high air-filled porosity. This effect likely causes near-surface peat
119 desiccation and reduces stomatal conductance³⁵, thereby reducing CO₂ uptake and
120 slowing down the translocation of photosynthates between above- and belowground
121 biomass²¹. This sequence of events suggests that peat swamp forests are not adapted
122 to low GWL³⁶.

123 The results suggest that changes in rainfall regime play a central role in shaping the
124 seasonal and interannual variability of intact tropical peatland hydrology (Figure 2a), and
125 therefore the CO₂ budget (Figure 2c). The observed CO₂ emissions due to GWL
126 drawdown in this study are consistent with previous studies in tropical peatlands where
127 the carbon loss^{11,13}, reduction in peat accumulation rate^{17,37} and a hiatus in peat genesis¹⁶
128 have been reported in response to droughts driven by ENSO activity.

129 The relationship between net ecosystem CO₂ exchange and GWL suggests that the intact
130 site would approach CO₂ neutrality if the peat hollow surfaces were continuously flooded
131 (Figure 3a). Rainfall, evapotranspiration and GWL measurements clearly demonstrated
132 that during the dry period the ecosystem's demand for evapotranspiration exceeded the
133 rainfall in this ombrotrophic environment (Figure 2a). This resulted in GWL drawdown,
134 even though the study area is hydrologically connected to a large peat dome. The
135 condition was exacerbated during prolonged drought periods induced by climate

136 extremes, for instance during the strong positive IOD together with El Niño event which
137 occurred during 2019. This significantly low GWL driven by rainfall deficit can lead to tree
138 mortality³⁶, CO₂ emissions¹² and subsidence in these ecosystems. The results indicate
139 that the long-term rate of carbon accumulation of 2.8 tCO₂ ha⁻¹ yr⁻¹ in coastal peatlands
140 of Indonesia¹ may no longer occur under current rainfall regime. This is further indicated
141 by observed subsidence (3.1 cm yr⁻¹) at the sampling location far away from the forest
142 edge (>5 km), that should have minor effects (if any) of regional land-cover change³⁸.

143 Observed net CO₂ emissions during years without El Niño and positive IOD events are
144 more surprising, since an intact peatland would be expected to act as a long-term CO₂
145 sink. This may raise the possibility that chronic long-term changes in rainfall seasonality
146 and quantity⁸⁻¹⁰, could be promoting a large-scale shift of the tropical peatland carbon
147 balance, exacerbated by superimposed and/or lagged responses to periodic El Niño or
148 positive IOD events. Longer term measurements from this and other sites are therefore
149 needed to establish whether our observations are occurring across other remaining areas
150 of intact peatland and to provide reliable and robust understanding of long-term climate
151 responses.

152 **Degradation enhances CO₂ emissions**

153 During the study period, October 2016–September 2020, GWL in the degraded peatland
154 ranged between -0.21 to -1.18 m with an average of -0.66 ± 0.21 m, indicating that a
155 significant part of the peat profile was aerated throughout the year (Figure 2b). The
156 degraded peatland emitted 39.8 ± 2.9 tCO₂ ha⁻¹ yr⁻¹ and subsided 4.2 ± 1.3 cm yr⁻¹ (Table
157 1, Extended Data Table 2).

158 We attribute observed high R_{eco} at the degraded site to four interlinked ecosystem
159 responses to degradation. Firstly, greater peat aeration due to consistently deeper GWL
160 enhances heterotrophic respiration rates. Secondly, increased soil temperatures due to
161 both canopy cover loss and GWL drawdown further boosts microbial activities and
162 heterotrophic respiration (Extended Data Table 2). Thirdly, the area was noted to be
163 continuously losing large trees from 'edge-effect exposure', where both wind exposure
164 and deeper GWL are probably involved. Big trees are a dominant repository of above-
165 ground carbon stock in tropical peatlands and even a modest rate of loss may exceed the
166 biomass gain by any number of small trees. The contribution of CO_2 emissions from peat
167 decomposition is experienced immediately after GWL drawdown, but the release of CO_2
168 from coarse woody debris may be delayed as dead trees do not decompose
169 instantaneously, providing a lagged but sustained contribution to R_{eco} ³⁹. Finally, R_{eco} may
170 be enhanced by higher autotrophic respiration as a result of higher GPP from
171 regenerating trees and shrubs.

172 In contrast to the intact peatland, net ecosystem CO_2 exchange at the degraded peatland
173 appeared to be less sensitive to GWL changes (Figure 3a). Net ecosystem CO_2 exchange
174 did not change with GWL from -0.4 to -0.8 m at the degraded site as the increase in R_{eco}
175 was fully offset by increased GPP (Extended Data Figure 2b,c). The adaptation of tree
176 species to a low GWL environment has been reported in a drained tropical peatland²¹.
177 Tree roots grow into aerated deeper peat layers and hence these trees do not experience
178 water stress even with a GWL of -0.9 m²¹. However, the increase in GPP with deeper
179 GWL may not sustain with further drawdown in the later part of a prolonged dry season
180 when GWL falls below a critical level²¹.

181 **Emissions of CH₄, N₂O and fluvial carbon export**

182 The intact peatland emitted 73 ± 31 kgCH₄ ha⁻¹ yr⁻¹ and 0.2 ± 0.2 kgN₂O ha⁻¹ yr⁻¹ (Table
183 1). A proportion of ecosystem net primary production is also exported in the forms of
184 dissolved inorganic/organic and particulate organic carbon via lateral export. A fluvial
185 carbon export of 1.1 ± 0.2 tC ha⁻¹ yr⁻¹ was reported in the intact peatland within the same
186 landscape⁴⁰. We conservatively assume that all fluvial carbon is ultimately emitted as
187 CO₂⁴¹. Similarly, the degraded peatland emitted 43 ± 11 kgCH₄ ha⁻¹ yr⁻¹ and 1.1 ± 0.6
188 kgN₂O ha⁻¹ yr⁻¹. A fluvial carbon export of 1.8 ± 0.4 tC ha⁻¹ yr⁻¹ can further occur from a
189 drained tropical peatland⁴⁰.

190 GWL is a key driver of CH₄ emissions from tropical peatlands²⁹ (Figure 3b). The lower
191 GWL reduces CH₄ emissions as aerobic conditions are unfavorable to methanogens and
192 promote methanotrophy. In addition, GWL drawdown below the root zone will limit plant-
193 mediated transport of CH₄ from the anaerobic zone to the atmosphere²⁹. Higher N₂O
194 emissions at the degraded peatland as compared to the intact site can be due to
195 accelerated mineralization of the peat under aerobic conditions, releasing mineral
196 nitrogen as ammonium (Extended Data Table 1) and producing N₂O as a by-product of
197 the nitrification process⁴².

198 **Climate benefits from tropical peatland conservation**

199 Our study indicates that peatland degradation resulted in a significant increase in net CO₂
200 emissions, decrease in CH₄ emissions and increase in N₂O emissions (Table 1). We
201 computed GHG balance of both sites using sustained-flux global warming potential of 1,
202 45 and 270 for CO₂, CH₄ and N₂O over a 100-year time period⁴³. The GHG balance for

203 CO₂, CH₄ and N₂O increased from 20.0 ± 4.5 tCO₂e ha⁻¹ yr⁻¹ at the intact site to 43.8 ±
204 1.5 tCO₂e ha⁻¹ yr⁻¹ at the degraded site. Although the measurements indicate that both
205 intact and degraded peatlands in this study are warming the atmosphere (Table 1), it
206 remains clear that protection of the remaining intact tropical peatlands in Indonesia offers
207 a viable way to avoid substantial GHG emissions from this globally important ecosystem.
208 The results suggest that protecting intact tropical peatland in Sumatra from degradation
209 effectively avoid GHG emissions of 23.8 ± 4.7 tCO₂e ha⁻¹ yr⁻¹ (Table 1). The results
210 highlight that conserving all remaining intact peat swamp forests in Sumatra and
211 Kalimantan (5.97 Mha)⁴⁴ will avoid GHG emissions of around 0.14 GtCO₂e yr⁻¹. This
212 equates to ~10% of Indonesia's GHG emissions in 2016. This estimate is conservative.
213 If some remaining intact peatlands are continuing to sequester CO₂, the avoided
214 emissions will be correspondingly higher. The Paris Agreement target to keep global
215 warming to below 2.0°C above pre-industrial times reinforces REDD+ (Reducing
216 Emissions from Deforestation and Forest Degradation). The role of conservation in the
217 responsible management of peatland landscapes³³ in developing countries is a key
218 component of future global climate change mitigation efforts. Apart from protecting carbon
219 stock and avoiding GHG emissions, intact tropical peatlands deliver ecosystem services
220 such as water storage, flood protection, nature-based recreation and biodiversity support.
221 Given the rapid observed changes in rainfall regime and land-use, conservation
222 measures implemented now will be more effective than those implemented in future,
223 because any delay will likely lead to continued forest loss and ecosystem degradation,
224 requiring greater restoration effort which will cost more⁴⁵ and be more challenging⁴⁶.

225 Therefore, peatland management approaches³³ that integrate evidence-based, long-term
226 conservation commitment and finance should be encouraged.

227 The estimated effect of peatland degradation on GHG emissions related to land-cover
228 change that we present here may vary in time and space. In addition to variations related
229 to natural hydrology driven by rainfall, the effect is also likely to vary with degradation
230 level²¹ including forest cover loss and drainage intensity. Furthermore, results presented
231 here are specific for coastal Indonesian peatlands, thus, should not be extrapolated to
232 other tropical peatlands, such as those of the Amazon and Congo basins, since they have
233 different rainfall regimes, vegetation and peat formation history. Therefore, more
234 ecosystem-scale measurements from the peatlands within Amazon and Congo basins
235 are needed since they are also of global significance in terms of future potential
236 deforestation and degradation driven by both land-use and climatic changes^{9,26,47}.

237 A strong negative relationship between CO₂ emissions and GWL for peatlands across
238 Southeast Asia (Extended Data Figure 3) suggests that our results may be regionally
239 applicable. Over coming decades, it is likely that tropical peatland ecosystems will
240 become increasingly exposed to changes in rainfall regime, such as those induced by
241 more frequent and severe El Niño and positive IOD events^{17,18}. The strong response of
242 CO₂ emissions to extreme drought in the intact peatland site indicates the potential
243 significance of climate regime in determining the future GHG budget of these ecosystems,
244 highlighting their vulnerability and global consequences. Given the large GHG emissions,
245 a continuing use of a long-term rate of carbon accumulation for future projection^{7,24} is
246 incompatible with ongoing efforts to better understand and manage remaining intact
247 tropical peatlands in a changing climate and thus should be considered with caution. The

248 results should help to reduce the uncertainty in the estimation of GHG budgets from a
249 globally important ecosystem, develop science-based peatland management practices to
250 reduce carbon emissions and improve the understanding of their role under current and
251 future climate.

252 **Reference**

- 253 1. Dommain, R., Couwenberg, J. & Joosten, H. Development and carbon sequestration
254 of tropical peat domes in south-east Asia: links to post-glacial sea-level changes and
255 Holocene climate variability. *Quat. Sci. Rev.* **30**, 999–1010 (2011).
- 256 2. Dargie, G. C. et al. Age, extent and carbon storage of the central Congo Basin
257 peatland complex. *Nature* **542**, 86–90 (2017).
- 258 3. Lähteenoja, O. et al. The large Amazonian peatland carbon sink in the subsiding
259 Pastaza-Maranon foreland basin, Peru. *Glob. Change Biol.* **18**, 164-178 (2012).
- 260 4. Warren, M., Hergoualc'h, K., Kauffman, J. B., Murdiyarso, D., & Kolka, R. An
261 appraisal of Indonesia's immense peat carbon stock using national peatland maps:
262 uncertainties and potential losses from conversion. *Carbon Balance Manag.* **12(1)**,
263 12 (2017).
- 264 5. Gumbricht, T. et al. An expert system model for mapping tropical wetlands and
265 peatlands reveals South America as the largest contributor. *Glob. Change Biol.* **36**,
266 335 (2017).
- 267 6. Xu, J., Morris, P. J., Liu, J., & Holden, J. PEATMAP: Refining estimates of global
268 peatland distribution based on a meta-analysis. *Catena* **160**, 134–140 (2018).
- 269 7. Gallego-Sala, A. V. et al. Latitudinal limits to the predicted increase of the peatland
270 carbon sink with warming. *Nat. Clim. Chan.* **8**, 907–913 (2018).

- 271 8. Loisel, J., et al. Expert assessment of future vulnerability of the global peatland
272 carbon sink. *Nat. Clim. Chan.* **11**, 70–77 (2021).
- 273 9. Wang, S., Zhuang, Q., Lahteenoja, O., Draper, F. C. & Cadillo-Quiroz, H. Potential
274 shift from a carbon sink to a source in Amazonian peatlands under a changing
275 climate. *Proc. Natl. Acad. Sci.* **115**, 12407–12412 (2018).
- 276 10. Li, W. et al. Future precipitation changes and their implications for tropical peatlands.
277 *Geophys. Res. Lett.* **34**, L01403 (2007).
- 278 11. Dommain, R., Couwenberg, J., Glaser, P. H., Joosten, H., & Suryadiputra, I. N. N.
279 Carbon storage and release in Indonesian peatlands since the last deglaciation.
280 *Quat. Sci. Rev.* **97**, 1–32 (2014).
- 281 12. Cobb, A. R. et al. How temporal patterns in rainfall determine the geomorphology
282 and carbon fluxes of tropical peatlands. *Proc. Natl. Acad. Sci. USA* **114** (26) E5187-
283 E5196 (2017).
- 284 13. Swails, E. et al. The response of soil respiration to climatic drivers in undrained
285 forest and drained oil palm plantations in an Indonesian peatland. *Biogeochemistry*
286 **142**, 37–51 (2019).
- 287 14. Yu, Z., Loisel, J., Brosseau, D. P., Beilman, D. W., & Hunt, S. J. Global peatland
288 dynamics since the last glacial maximum. *Geophys. Res. Lett.* **37**, L13402 (2010).
- 289 15. Morris, P. J. et al. Global peatland initiation driven by regionally asynchronous
290 warming. *Proc. Natl. Acad. Sci. USA* **115**, 4851– 4856 (2018).
- 291 16. Ruwaimana, M. Anshari, G. Z. Silva, L. C. R. & Gavin, D. G. The oldest extant
292 tropical peatland in the world: a major carbon reservoir for at least 47,000 years.
293 *Environ. Res. Lett.* **15** (2020).

- 294 17. Cai, W. et al. Increasing frequency of extreme El Niño events due to greenhouse
295 warming. *Nat. Clim. Chan.* **4**, 111–116 (2014).
- 296 18. Cai, W. et al. Increased frequency of extreme Indian Ocean Dipole events due to
297 greenhouse warming. *Nature* **510**, 254–258 (2014).
- 298 19. Saji, N.H., Goswami, B.N., Vinayachandran, P.N., & Yamagata, T. A dipole mode in
299 the tropical Indian Ocean. *Nature* **401**, 360-363 (1999).
- 300 20. Alsepan, G., & Minobe, S. Relations between interannual variability of regional-scale
301 Indonesian precipitation and large-scale climate modes during 1960–2007. *J. Climate*
302 **33**, 5271–5291 (2020).
- 303 21. Hirano, T., et al. Effects of disturbances on the carbon balance of tropical peat
304 swamp forests. *Glob. Change Biol.* **18**, 3410–3422 (2012).
- 305 22. Tang, A. C., et al. A Bornean peat swamp forest is a net source of carbon dioxide to
306 the atmosphere. *Glob Change Biol.* **26(12)**, 6931–6944 (2020).
- 307 23. Leifeld, J. & Menichetti, L. The underappreciated potential of peatlands in global
308 climate change mitigation strategies. *Nat. Commun.* **9**, 1–8 (2018).
- 309 24. Leifeld, J., Wüst-Galley, C., & Page, S. Intact and managed peatland soils as a
310 source and sink of GHGs from 1850 to 2100. *Nat. Clim. Chan.* **9**, 1–3 (2019).
- 311 25. Miettinen, J., Shi, C., & Liew, S.C. Land-cover distribution in the peatlands of
312 Peninsular Malaysia, Sumatra and Borneo in 2015 with changes since 1990. *Glob.*
313 *Ecol. Conserv.* **6**, 67–78 (2016).
- 314 26. Roucoux, K. H. et al. Threats to intact tropical peatlands and opportunities for their
315 conservation. *Conserv. Biol.* **31(6)**:1283–1292 (2017).

- 316 27. Hooijer, A. et al. Subsidence and carbon loss in drained tropical peatlands.
317 *Biogeosciences* **9**, 1053–71 (2012).
- 318 28. Hoyt, A. M., Chaussard, E., Seppäläinen, S. S. & Harvey C. F. Widespread
319 subsidence and carbon emissions across Southeast Asian peatlands. *Nat. Geosci.*
320 **13**, 435–440 (2020).
- 321 29. Deshmukh, C. S. et al. Impact of forest plantation on methane emissions from
322 tropical peatland. *Glob. Change Biol.* **26**, 2477–2495 (2020).
- 323 30. Prananto, J. A., Minasny, B., Comeau, L-P., Rudiyanto, R., & Grace, P. Drainage
324 increases CO₂ and N₂O emissions from tropical peat soils. *Glob Change Biol.* **00**:1–
325 18 (2020).
- 326 31. Myhre, G. et al. In *Climate Change 2013* (eds Stocker, T. F. et al.) 659–740
327 Cambridge University Press, Cambridge, UK and New York, USA, (2013).
- 328 32. Murdiyarso, D., Lilleskov, E. & Kolka, R. Tropical peatlands under siege: the need
329 for evidence-based policies and strategies. *Mitig. Adapt. Strateg. Glob. Change* **24**,
330 493–505 (2019).
- 331 33. Clarke, D. & Rieley, J. O. *Strategy for responsible peatland management*. 6 Edition,
332 International Peatland Society, Jyväskylä, Finland (2019).
- 333 34. Sulistiyanto, Y. Nutrient dynamics in different sub-types of peat swamp forest in
334 Central Kalimantan, Indonesia. Doctoral dissertation, University of Nottingham.
335 225p. (2004).
- 336 35. Hirano, T., Kusin, K. Limin, S. & Osaki M. Evapotranspiration of tropical peat swamp
337 forests. *Glob. Change Biol.* **21**, 1914–1927 (2015).

- 338 36. Nishimura, T. et al. Mortality and growth of trees in peat-swamp and heath forests in
339 Central Kalimantan after severe drought. *Plant Ecol.* **188**, 165–177 (2007).
- 340 37. Kelly, T. J. et al. The vegetation history of an Amazonian domed peatland.
341 *Palaeogeogr. Palaeoclimatol. Palaeoecol.* **468**, 129–141 (2017).
- 342 38. Cobb, A. R., Dommain, R. R., Tan, F., Heng, N. H. E., & Harvey, C. F. Carbon
343 storage capacity of tropical peatlands in natural and artificial drainage networks.
344 *Environ. Res. Lett.* **15**, 23–25 (2020).
- 345 39. Kiew, F. et al. Carbon dioxide balance of an oil palm plantation established on
346 tropical peat. *Agric. For. Meteorol.* **295**, 1–8 (2020).
- 347 40. Yupi, H. M., Inoue, T., Bathgate, J., & Putra, R. Concentrations, loads and yields of
348 organic carbon from two tropical peat swamp forest streams in Riau Province,
349 Sumatra, Indonesia. *Mires and Peat*, **18**, 1–15 (2016).
- 350 41. Evans, C. D., Renou-Wilson, F. & Strack, M. The role of waterborne carbon in the
351 greenhouse gas balance of drained and re-wetted peatlands. *Aquat Sci.* **78**, 573–
352 590 (2016).
- 353 42. Jauhainen, J. et al. Nitrous oxide fluxes from tropical peat with different disturbance
354 history and management. *Biogeosciences* **9**, 1337–1350 (2012).
- 355 43. Neubauer, S. C., & Megonigal J. P. Moving beyond global warming potentials to
356 quantify the climatic role of ecosystems. *Ecosystems* **18**, 1000–1013 (2015).
- 357 44. Wijedasa, L. S. et al. Carbon emissions from South-East Asian peatlands will
358 increase despite emission reduction schemes. *Glob. Change Biol.* **00**, 1–16 (2018).
- 359 45. Hansson, A. & Dargusch, P. An estimate of the financial cost of peatland restoration
360 in Indonesia. *Case Stud. Environ.* **2(1)** 1–8 (2017).

- 361 46. Ward, C. et al. Wading through the swamp: what does tropical peatland restoration
362 mean to national-level stakeholders in Indonesia? *Restor. Ecol.* **28**, 817-827 (2020).
- 363 47. Lilleskov, E. et al. Is Indonesian peatland loss a cautionary tale for Peru? A two-
364 country comparison of the magnitude and causes of tropical peatland degradation.
365 *Mitig. Adapt. Strateg. Glob. Change* **24**, 591–623 (2019).

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370 **AUTHORS' CONTRIBUTIONS**

371 C.S.D., C.D.E. and S.E.P. conceived of the study. C.S.D., D.J., A.P.S, Y.W.S, and A.R.D.
372 completed eddy covariance data processing. D.J., Nardi, A.P.S., Nurholis, M.H., C.S.D.,
373 S.K., Y.S. and A.A. performed data collection, eddy covariance instruments calibration
374 and maintenance. C.S.D. conceived the paper and wrote the initial draft, to which all
375 authors provided critical contributions and approved submission.

376 **CONFLICT OF INTEREST**

377 C.D.E., S.E.P., S.S., V.G., F.A., and D.A. contributed to this paper as part of their
378 contribution to the Independent Peat Expert Working Group (IPEWG), which was set up
379 by Asia Pacific Resources International Ltd. (APRIL) to provide objective science-based
380 advice on peatland management. ARD's contribution was also supported by APRIL to
381 provide technical guidance on the Eddy Covariance data processing including quality

382 controls and gap-filling protocols. C.S.D., D.J., Nardi, A.P.S., Nurholis, M.H., S.K., Y.S.,
383 A.A. and Y.W.S. are employed by APRIL to conduct data collection, instrument
384 maintenance and calibration. The funders had no role in the interpretation of data, in the
385 writing of the manuscript, or in the decision to publish the results. The authors declare
386 that all views expressed are their own.

387 **DATA AVAILABILITY STATEMENT**

388 All data that support the findings of this study are archived on
389 <http://doi.org/10.5281/zenodo.4835696>.

390 **Methods**

391 **Study area**

392 The study area, Kampar Peninsula, is an ombrotrophic coastal tropical peatland of around
393 700,000 ha and largely formed within the past 8,000 years¹. The peat thickens from
394 approximately 3 m deep near the river boundaries to over 11 m in the center of the
395 approximately 60 km wide dome, with an average depth of 8 m. The peninsula
396 experiences a humid tropical climate (warm year-round) with average monthly air
397 temperature ranging from 26 to 28°C (Extended Data Table 2). The interannual variability
398 of rainfall is influenced by ENSO and IOD^{19,20} (Extended Data Figure 1). The average
399 annual rainfall for the last 6 years (2014–2019, with 2015 El Niño, 2017 La Niña, and
400 2019 with a major positive IOD combined with an El Niño event) is ~1900 mm and
401 characterized by a high seasonal variability with two rainfall peaks, one in November-
402 December and one in March-April. The peninsula land-cover is characterized by a large
403 central intact forest area surrounded by a mosaic of degraded peat swamp forest,

404 industrial fiber wood plantation (largely *Acacia crassicarpa*), smallholder agriculture
405 (largely oil palm, *Elaeis guineensis*) and degraded, shrub and undeveloped open land²⁵
406 (Figure 1).

407 The site selection for the intact peatland were carefully under taken to minimize any local
408 land-use effect on measured fluxes. The intact peatland site is characterized as peat
409 swamp forest^{25,44} and comprises one of the largest remaining peat swamp forest in
410 Southeast Asia. The forest structure was determined from vegetation survey in
411 permanent sampling plots (20 x 125 m) around the tower. It was identified as mixed forest
412 with uneven canopy (the tallest canopy is in a range of 28–35 m). Tree density with
413 diameter at breast height >5 cm was 1,343 trees per hectare. The dominant tree species
414 of the overstory are *Shorea uliginosa*, *Calophyllum ferrugineum* and *Syzygium spp.*;
415 together they represent around 75% of the overstory vegetation (Extended Data Table
416 1). The understory is dominated by *Pandanus spp.*, *Cyrtostachys renda* and *Nepenthes*
417 *spp.* The forest floor is uneven with a hummock-hollow microtopography, and covered
418 with tree debris, root mat and leaf litter. Hollow surfaces are often 20–40 cm lower than
419 hummock tops. The average area ratio of hollow to hummock was 3:1 around the tower.
420 The surface peat type is fibric and the average peat thickness is $\sim 9.0 \pm 1.0$ m in the area
421 surrounding the tower measured using the Eijkelkamp Peat Auger. The peat is composed
422 almost entirely of organic matter with soil organic carbon of $50 \pm 2\%$ determined using
423 the Loss on Ignition method and a dry bulk density of 0.08 ± 0.03 g cm⁻³ measured using
424 the gravimetric method. These carbon concentration and dry bulk density values are
425 consistent with other measurements in the region⁴⁸ (bulk density of 0.08 g cm⁻³ (0.07–
426 0.09 g cm⁻³) and carbon concentrations of 55% (53–57%)). The GWL fluctuates

427 seasonally following the rainfall variation due to the ombrotrophic nature of the area²⁹. An
428 integrated climatologic footprint analysis⁴⁹ indicated that approximately 80% of fluxes
429 were derived within one km in the upwind direction (Figure 1), and thus originated within
430 intact peat swamp forest^{25,44}. The area outside the flux footprint on the southeastern side
431 of the tower site was disturbed somewhat by historic logging activity within the wider peat
432 landscape in the 1990s, which included logging track and canal construction, or to the
433 effects of plantation operation around the edge of the Kampar peat dome. While we
434 cannot exclude some long-term hydrological effect of land use activities beyond the flux
435 tower footprint on the study area, any such effects appear minor, and there is no indication
436 that it has affected the functioning of the forest^{25,44}. The closest logging track was more
437 than 1 km from the flux tower site in the southeast direction and outside the flux
438 measurement footprint. There was neither logging nor canal construction activity within
439 the footprint and it retained an intact forest. The closest area of active plantation was 3.5
440 km to the southeast, well outside the flux footprint. To avoid any possible boundary effect
441 and associated bias, the measurements between 78 and 191° wind direction were further
442 excluded in this study (Figure 1).

443 The second tower is located on the boundary of plantation and degraded forest (Figure
444 1). To represent only the degraded area, the measurements between 270 and 90° wind
445 direction were considered in this study. The degraded peatland was selectively logged
446 and drained in the late 1990s and early 2000s, while some parts were burnt in 2014. The
447 average canopy height was about 19 m (Extended Data Table 1). Tree densities with
448 diameter at breast height >5 cm were 663 trees per hectare. Compared to the intact
449 peatland site, more than 50% of the large trees have been logged or fallen and many of

450 those remaining are now leaning. The dominant tree species of the over-story are
451 *Syzygium claviflorum*, *Shorea teysmanniana*, *Stemonurus secundiflorus*, *Blumeodendron*
452 *kurzii*, *Horsfieldia crassifolia* and *Macaranga pruinosa*; together they represent around
453 60% of the overstory vegetation. The understory is dominated by *Dicranopteris linearis*,
454 *Nepenthes spp.*, *Pandanus sp.* and *Cyrtostachys renda*. Formerly burned areas comprise
455 of modest biomass in shrub, sedge and fern (*Nephrolepis sp.*, *Stenochlaena sp.*,
456 *Blechnum sp.*), with increasing cover of shrubs and small trees of *Macaranga pruinosa*
457 with time since fire. The surface peat type is fibric and the average peat thickness is 8.2
458 \pm 0.8 m. The forest floor in the degraded part was uneven with a hummock-hollow
459 microtopography. Hummocks were relatively small in size (typically \sim 30 cm in height).
460 The average area ratio of hollow to hummock was 3:1. There were no hummocks on the
461 ground in the ex-burnt area. The surface peat is composed almost entirely of organic
462 matter with soil organic carbon of $55 \pm 0.5\%$ and a dry bulk density of $0.09 \pm 0.03 \text{ g cm}^{-3}$.
463 The integrated climatologic footprint analysis⁴⁹ indicated that approximately 80% of fluxes
464 originated within one km in the upwind direction and the ex-burnt area only represented
465 5% of the flux footprint (Figure 1). The average footprint can be considered as
466 representative of the majority of unmanaged degraded peatlands in Southeast Asia²⁵.

467 The terrain around the towers is flat (slope $<0.05\%$), ensuring a good fetch regardless of
468 wind direction. The relatively close proximity of the intact peatland and the degraded
469 peatland sites (\sim 35 km apart) within the same peat landscape avoids potentially
470 confounding variables such as climatic differences¹, past natural succession⁵⁰ and peat
471 formation⁵¹ (Figure 1, Extended Data Table 1). Thus, although it is inherently difficult and
472 expensive to replicate flux measurements using the eddy covariance (EC) technique, the

473 sites should provide a robust and unbiased basis for evaluating the effect of land-cover
474 change (from intact peatland to degraded peatland) on CO₂ and CH₄ exchanges.

475 **Eddy covariance and environmental variables measurements**

476 The EC measurements were conducted for seven site-years (October 2016-September
477 2020 over the degraded site and June 2017-May 2020 over the intact site). The EC
478 system consisted of an enclosed path CO₂/H₂O analyzer (LI-7200, LI-COR Inc.) to
479 measure the atmospheric CO₂ and H₂O, an open path CH₄ analyzer (LI-7700, LI-COR
480 Inc.) to measure CH₄ and a three-dimensional sonic anemometer (WindMaster Pro3-Axis
481 Anemometer, Gill Instruments Limited) to measure orthogonal components of wind speed
482 fluctuations. The sensors were mounted at the top of the tower to ensure complete
483 exposure in all directions (Figure 1b,c). The raw turbulence EC data were recorded at 10
484 Hz using an analyzer interface unit (LI-7550, LI-COR Inc.) and stored on a removable
485 flash disk (APRO, Industrial Grade USB Flash Disk). The filters of the CO₂ analyzer were
486 manually cleaned either at biweekly or if the flow drive (indicating filter clogging condition)
487 increased above 80% (LI-7200, LI-COR Inc.). The mirrors of the CH₄ analyzer were self-
488 cleaned either at 5:00 (local time) every day or if the received signal strength indicator
489 (RSSI) dropped below 20%. Furthermore, the upper and lower mirrors of the CH₄ analyzer
490 were manually cleaned on a biweekly basis²⁹.

491 Quantum sensors (LI-190SL-50, LI-COR Inc.) were mounted at the top of the towers to
492 measure the incoming photosynthetic photon flux density (PPFD, $\mu\text{mol m}^{-2} \text{s}^{-1}$). The
493 vertical profiles of relative humidity (%) and air temperature (°C) were measured using air
494 temperature and humidity probes (Vaisala HMP155 Humidity Temperature Probe,

495 Vaisala Inc.), which were installed inside a ventilated radiation shield at five heights of 3,
496 7, 14, 21 and 40 m for the degraded site and 4, 11, 20, 29 and 48 m for the intact site.
497 The vertical profiles of CO₂ concentrations were measured at four heights of 3, 14, 21
498 and 40 m for the degraded site and 4, 11, 29 and 48 m for the intact site to calculate flux
499 storage term below the measurement height²¹ using a closed-path CO₂ analyzer (LI-8100,
500 LI-COR Inc.). The sampling lines were changed in rotation every 90 seconds and CO₂
501 concentrations were measured for last 10 seconds of each 90 seconds sampling time at
502 each sampling height and recorded with a data logger (LI-8100, LI-COR Inc.); therefore,
503 one rotation of measurements took six minutes in every 30 minutes. Both enclosed-path
504 and closed-path CO₂ analyzers were calibrated every three months using ultra high-purity
505 nitrogen as the zero-point gas for CO₂ and reference gases with concentration of 396
506 ppm CO₂ in air (certified grade ± 1 ppm). Soil temperature ($^{\circ}\text{C}$) was measured at 0.15 m
507 below the hollow peat surface using temperature probe (Stevens Hydra Probe II, Stevens
508 Water Monitoring Systems, Inc.) with three replicates at the intact peatland. The soil
509 temperature were not measured at the degraded site due to the site logistic issues.

510 All meteorological sensors took measurements every second and were recorded as one
511 minute average with a data logger (Sutron Model 9210 XLITE, Sutron Corporation). All
512 measuring systems were powered using solar panels along with a rechargeable battery
513 system (65 Watt Solar Package, SunWize Power & Battery).

514 Daily rainfall (mm d^{-1}) were measured using three and two manual bucket rain gauges
515 within 11 km distance from the tower location in the intact site and the degraded site,
516 respectively. Rain gauges were installed 1.5 m above the ground, in an open area so that
517 rainfall was not intercepted by the tree canopy.

518 Peat subsidence was measured at eight locations in the intact site and four locations in
519 the degraded site (Figure 1f,g), with hollow, perforated 5 cm diameter hollow polyvinyl
520 chloride (PVC) poles, inserted vertically into the peat and anchored into underlying
521 mineral subsoil following the approach described in ref⁵². Annual average subsidence
522 rates were derived from measurements between December 2016-September 2020 and
523 December 2017-June 2020 in the intact and the degraded peatland respectively.

524 GWLs (m) were monitored as the water elevation relative to the ground surface, taking
525 the base of the hollows as a datum²⁹, every 30 min using GWL logger (Solinst Levellogger
526 Model 3001). The three and two GWL loggers respectively at the intact and the degraded
527 peatland were placed in a perforated PVC tubes that were inserted vertically into the peat
528 around the towers (Figure 1f,g). The GWL logger also recorded temperature at 1.5 m
529 below the peat surface. Additional GWLs were recorded biweekly at two locations per
530 tower site and on a quarterly basis at additional four locations in transect in the intact
531 peatland (Figure 1). Average GWLs were derived from nine and four monitoring locations
532 in the intact peatland and the degraded peatland respectively.

533 Measured environmental variables are summarized in Extended Data Table 2 and
534 presented in Extended Data Figure 5-6.

535 **Eddy covariance data processing**

536 Eddy covariance CO₂, CH₄ and H₂O (evapotranspiration) fluxes were computed from the
537 10 Hz concentration and vertical wind velocity data using EddyPro software (version
538 6.2.0, LI-COR Inc.) at a standard averaging interval of half-hour period⁵³. A de-spiking
539 procedure was applied to detect and eliminate individual out-of-range values for vertical

540 wind velocity and concentrations⁵⁴. De-trending was carried out using the block averaging
541 method. A coordinate correction was applied to force the average vertical wind velocity
542 to zero by the planar fit method⁵⁵. Frequency response loss corrections were applied to
543 compensate the flux losses at low and high frequencies⁵⁶. The Webb-Pearman-Leuning
544 correction⁵⁷ for air density fluctuations induced by temperature (thermal expansion) and
545 water vapor (dilution) was applied. Differences between deployment specific variables,
546 that is, sensor separation distance and instrument placement, were considered while
547 processing the data. The half-hourly CO₂ storage below the flux measurement height was
548 calculated from the four-point vertical profiles of CO₂ concentration, relative humidity and
549 air temperature by temporal interpolation²¹. Finally, the net ecosystem CO₂ exchange was
550 calculated as the sum of storage flux and eddy covariance flux. Owing to the large power
551 requirement and cost of a separate CH₄ analyzer, we could not conduct CH₄ profile
552 measurements to calculate CH₄ storage²⁹. In theory, accumulated CH₄ below the canopy
553 during nighttime is likely to be released and measured by the EC system following the
554 onset of turbulence after sunrise and the bias on annual sums should be negligible⁵⁸. We
555 adopted the standard meteorological notation whereby a negative value denotes a net
556 uptake of atmospheric CO₂ or CH₄ by the ecosystem, while a positive value indicates a
557 net release of CO₂ or CH₄ from the ecosystem to the atmosphere⁵³.

558 After a set of quality controls⁵⁹⁻⁶¹ and system malfunctions due to lightning strikes and
559 power supply failure, the numbers of high-quality measurements during the course of the
560 study were 41% and 38% for CO₂, 32% and 26% for CH₄ and 34% and 33% for
561 evapotranspiration in the intact and the degraded site, respectively. The similar range of
562 25-50% were reported for other tropical forested peatlands^{21,29,35,39}. We considered a total

563 half-hourly measurements of 12,926 and 10,694 for CO₂, 11,761 and 9,019 for CH₄ and
564 9,954 and 10,538 for evapotranspiration that met all quality criteria for the intact (between
565 191-78° wind direction) and the degraded (between 270-90° wind direction) site,
566 respectively. We gap-filled both low-quality and missing data, as is commonly done in
567 eddy covariance studies^{21,29,35,39,62-67}.

568 For CO₂, we applied three gap-filling approaches (a) marginal distribution sampling
569 (MDS)^{21,63}, (b) artificial neural network (ANN)⁶⁵; and (c) random forest (RF)⁶⁶. In addition,
570 we applied principal component analysis as an input to the algorithms to address
571 multidriver dependency of CO₂ exchange and reduce the internal complexity of the
572 algorithmic structures for the MDS approach⁶⁷. Following other regional eddy covariance
573 studies in peat swamp forests²¹, we performed MDS gap-filling using the REddyProc
574 package on a half-hourly basis⁶⁸ separately for the daytime (0600-1800 hr) and the
575 nighttime (1800-0600 hr) data. ANN and RF procedures were iterated 20 times. Average
576 of the 20 models were used to fill the gap and standard deviation was used to quantify
577 uncertainty due to gap-filling. Nighttime CO₂ exchanges were considered equivalent to
578 ecosystem respiration (R_{eco})⁶⁹. PPFD, VPD, T_{air} , GWL and friction velocity were used for
579 the daytime, and PPFD and VPD were excluded in the nighttime. GWL is reported as the
580 main controlling factor of R_{eco} from tropical peatlands²¹. Therefore, we used GWL as
581 environmental factors for the look-up table to derive daytime R_{eco} using the MDS gap-
582 filling algorithm⁶⁸. Then, GPP was calculated as the difference between CO₂ exchange
583 and R_{eco} on a half-hourly basis⁶⁹. Negative GPP represent gross CO₂ uptake by the
584 ecosystem whereas positive R_{eco} represent the gross CO₂ release from the ecosystem to
585 the atmosphere⁵³. We applied the same above gap-filling approaches for

586 evapotranspiration, using net radiation instead of PPFD during daytime, while adding net
587 radiation and VPD during nighttime. To provide a conservative estimate of CO₂ and
588 evapotranspiration, we used the average of the three approaches (Extended Data Table
589 3). After gap-filling, we corrected daily evapotranspiration for the energy imbalance using
590 net radiation, sensible heat and latent heat as described in ref³⁵. We applied two gap-
591 filling approaches, mean diurnal course (MDC) and MDS, for CH₄ as described in ref²⁹.
592 To provide a conservative estimate of CH₄, we used the average of the MDC and MDS
593 approaches (Extended Data Table 3).

594 Flux random uncertainty was calculated following ref⁷⁰. The standard deviation of three
595 different flux values derived from friction velocity thresholds of 5th, 50th and 95th
596 percentiles were applied as an uncertainty due to friction velocity threshold using the
597 REddyProc package⁶⁸. The total uncertainty in CO₂, CH₄ and evapotranspiration
598 (Extended Data Table 3) was calculated with the law of propagation of errors⁷¹.

599 **Soil N₂O flux measurements**

600 Soil N₂O fluxes were measured using the manual soil flux chamber⁷² within 2 km from the
601 EC towers. Three and four plots were selected in the intact and degraded peatland,
602 respectively (Figure 1f,g). At each plots, two stainless steel rectangular collars (surface
603 area of 0.08 m²) on hummocks and four in the adjacent hollow (randomly around 50-100
604 m apart) were inserted permanently 5 cm into the soil five months before the flux sampling
605 was started. Within 45 min, four air samples (in duplicates) from the chamber were
606 collected with a syringe and needle at 15 min intervals starting from the initial time when
607 chambers were placed onto the collars. The air within the chambers was gently mixed

608 prior to sample extraction using the syringe and needle. Air samples were transferred into
609 pre-evacuated 20 ml glass vials capped with butyl stoppers and aluminum seals. Analysis
610 of N₂O concentration was performed by gas chromatography (SRI® 8610C gas
611 chromatograph, USA) equipped with an electron capture detector within 48 hours.
612 Commercial gas standards with concentration of 350 ppbv N₂O in N₂ (uncertainties less
613 than 10%) were injected after an analysis of every 10 samples for calibration. Nitrous
614 oxide fluxes were calculated applying the ideal gas law to the slope of the linear
615 regression of gas concentration in the chamber versus time. The soil N₂O flux
616 measurements were made during the dry and wet season to represent variation in the
617 GWL and soil temperature. We averaged the soil N₂O fluxes of all replicates from each
618 sampling month, then we averaged all monthly values (June 2019-December 2020 for
619 the intact and July 2019-December 2020 for the degraded peatland) to estimate annual
620 N₂O emissions. Thus our annual estimates cover both spatial and temporal variability.
621 We didn't measure emissions from tree stems, this inclusion may increase total annual
622 N₂O emissions by 11-38%⁷³. Notably, given a minor (<1%) contribution of N₂O to total
623 GHG balance, our reported GHG emissions should be considered representative.

624 **References**

- 625 48. Couwenberg, J. & Hooijer, A. Towards robust subsidence-based soil carbon
626 emission factors for peat soils in south-east Asia, with special reference to oil palm
627 plantations. *Mires Peat* **12**, 1 (2013).
- 628 49. Kljun, N., Calanca, P., Rotach, M. W., & Schmid, H. P. A simple two-dimensional
629 parameterisation for Flux Footprint Prediction (FFP). *Geosci. Model Dev.* **8**, 3695–
630 3713 (2015).

- 631 50. Cole, L. E. S., Bhagwat, S. A., & Willis, K. J. Long-term disturbance dynamics and
632 resilience of tropical peat swamp forests. *J. Ecol.* **103**, 16–30 (2015)
- 633 51. Hapsari, K. A. et al. Environmental dynamics and carbon accumulation rate of a
634 tropical peatland in Central Sumatra, Indonesia. *Quat. Sci. Rev.* **169**, 173–187
635 (2017).
- 636 52. Evans, C. D. et al. Rates and spatial variability of peat subsidence in Acacia
637 plantation and forest landscapes in Sumatra, Indonesia. *Geoderma* **338**, 410–421
638 (2019).
- 639 53. Aubinet, M. et al. Estimates of the annual net carbon and water exchange of
640 European forests: The EUROFLUX methodology. *Adv. Ecol. Res.* **30**, 113–175
641 (2000).
- 642 54. Vickers, D., & Mahrt, L. Quality control and flux sampling problems for tower and
643 aircraft data. *J. Atmos. Oceanic Technol.* **14**, 512–526 (1997).
- 644 55. Wilczak, J. M., Oncley, S. P., & Stage, S. A. Sonic anemometer tilt correction
645 algorithms. *Bound.-Layer Meteorol.* **99**, 127–150 (2001).
- 646 56. Massman, W. J. A simple method for estimating frequency response corrections for
647 eddy covariance systems. *Agric. For. Meteorol.* **104(3)**, 185–198 (2000).
- 648 57. Webb, E. K., Pearman, G. I., & Leuning, R. Correction of flux measurements for
649 density effects due to heat and water vapour transfer. *Q. J. R. Meteorol. Soc.* **106**,
650 85–100 (1980).
- 651 58. Xu, K. E., et al. The eddy-covariance storage term in air: Consistent community
652 resources improve flux measurement reliability. *Agric. For. Meteorol.* **279**, 107734
653 (2019).

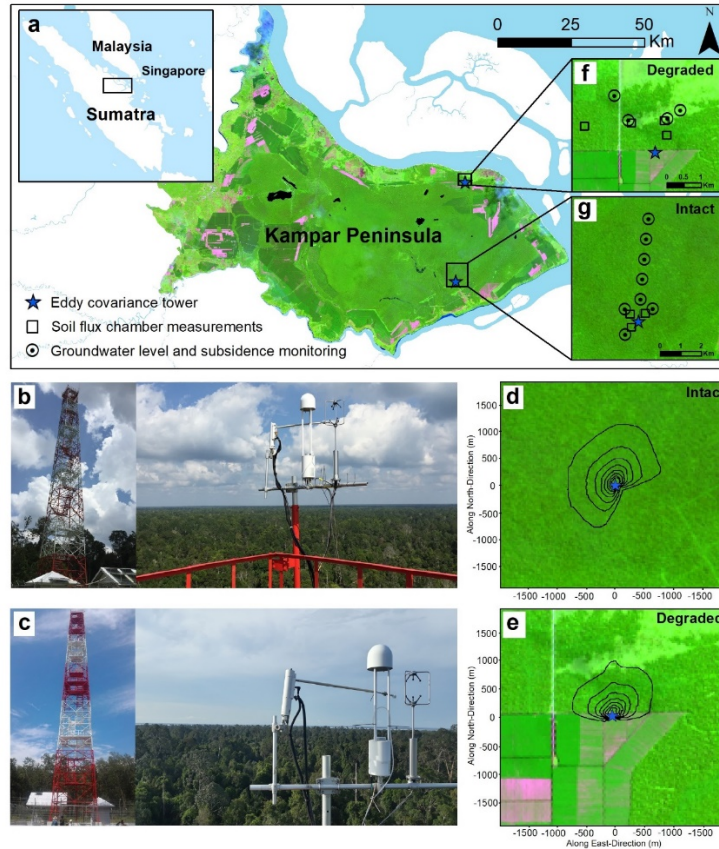
- 654 59. Foken, T., & Wichura B. Tools for quality assessment of surface-based flux
655 measurements. *Agric. For. Meteorol.* **78**, 83–105 (1996).
- 656 60. Mauder, M. et al. A strategy for quality and uncertainty assessment of long-term
657 eddy-covariance measurements. *Agric. For. Meteorol.* **169**, 122–135 (2013).
- 658 61. Papale, D. et al. Towards a standardized processing of net ecosystem exchange
659 measured with eddy covariance technique: algorithms and uncertainty estimation.
660 *Biogeosciences* **3**, 571–583 (2006).
- 661 62. Falge, E. et al. Gap-filling strategies for defensible annual sums of net ecosystem
662 exchange. *Agric. For. Meteorol.* **107(1)**, 43–69 (2001).
- 663 63. Kiew, F., et al. CO₂ balance of a secondary tropical peat swamp forest in Sarawak,
664 Malaysia. *Agr. For. Meteorol.* **248**, 494–501 (2018).
- 665 64. Moffat, A. M. et al. Comprehensive comparison of gap-filling techniques for eddy
666 covariance net carbon fluxes. *Agric. For. Meteorol.* **147(3)**, 209–232 (2007)
- 667 65. Papale, D., & Valentini, R. A new assessment of European forests carbon
668 exchanges by eddy fluxes and artificial neural network spatialization. *Glob. Change*
669 *Biol.* **9(4)**, 525–535 (2003).
- 670 66. Xu, T. et al. Evaluating different machine learning methods for upscaling
671 evapotranspiration from flux towers to the regional scale. *J. Geophys. Res. Atmos.*
672 **123(16)**, 8674–8690 (2018).
- 673 67. Kim, Y. et al. Gap-filling approaches for eddy covariance methane fluxes: A
674 comparison of three machine learning algorithms and a traditional method with
675 principal component analysis. *Glob. Change Biol.* **00**:1–20 (2019).

- 676 68. Wutzler, T. et al. Basic and extensible post-processing of eddy covariance flux data
677 with REdyProc. *Biogeosciences* **15**, 5015–5030 (2018).
- 678 69. Reichstein, M. et al. On the separation of net ecosystem exchange into assimilation
679 and ecosystem respiration: review and improved algorithm. *Glob. Change Biol.* **11**,
680 1424–1439 (2005).
- 681 70. Finkelstein, P. L., & Sims, P. F. Sampling error in eddy correlation flux
682 measurements. *J. Geophys. Res.* **106(D4)**, 3503–3509 (2001).
- 683 71. Deventer, M. J. et al. Error characterization of methane fluxes and budgets derived
684 from a long-term comparison of open- and closed-path eddy covariance systems.
685 *Agric. For. Meteorol.* **278**, 107638 (2019).
- 686 72. Serça, D., Delmas, R., Jambert, C. & Labroue, L. Emissions of nitrogen oxides from
687 equatorial rain forest in central Africa: origin and regulation of NO emission from
688 soils. *Tellus* **46B**, 243-254 (1994).
- 689 73. Iddris, N. A.-A., Corre, M. D., Yemefack, M., van Straaten, O., & Veldkamp, E. Stem
690 and soil nitrous oxide fluxes from rainforest and cacao agroforest on highly
691 weathered soils in the Congo Basin. *Biogeosciences*, **17**, 5377–5397 (2020).

692 Table 1 | Greenhouse gas balance (average \pm standard deviation) at the intact peatland
 693 and the degraded peatland in Sumatra, Indonesia. To quantify GHG balance in CO₂-
 694 equivalent, we used sustained-flux global warming potential of 1, 45 and 270 for CO₂,
 695 CH₄ and N₂O over a 100-year time period, respectively⁴³. We assumed that all fluvial
 696 carbon export is ultimately converted to CO₂⁴¹. Change in GHG balance due to peatland
 697 degradation in Sumatra is given in lowermost line.

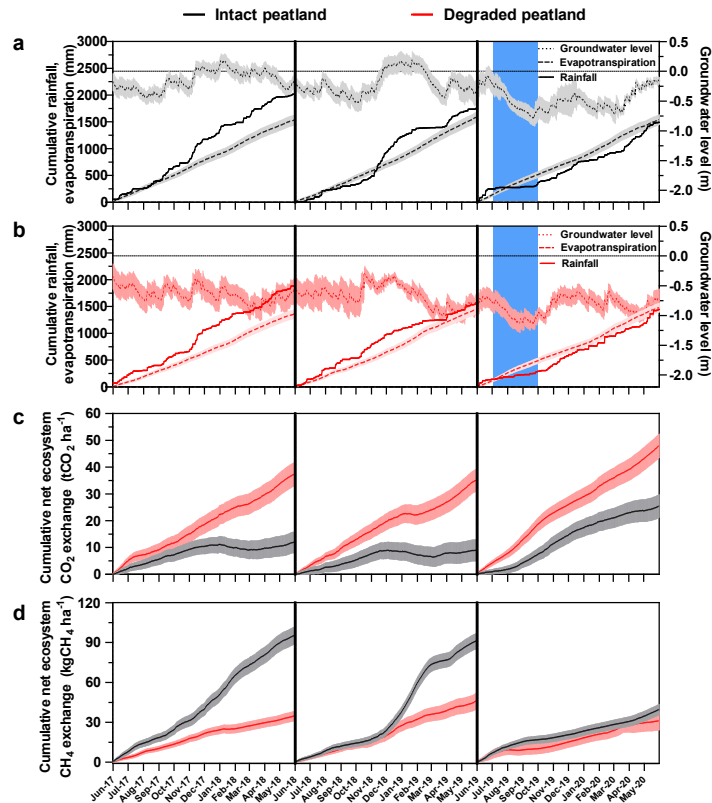
	Intact peatland	Degraded peatland
Net ecosystem CO ₂ exchange (tCO ₂ ha ⁻¹ yr ⁻¹)	15.5 \pm 8.8	39.8 \pm 2.9
Net ecosystem CH ₄ exchange (kgCH ₄ ha ⁻¹ yr ⁻¹)	73 \pm 31	43 \pm 11
Soil N ₂ O flux (kgN ₂ O ha ⁻¹ yr ⁻¹)	0.2 \pm 0.2	1.1 \pm 0.6
Fluvial carbon export (tC ha ⁻¹ yr ⁻¹)	0.3 \pm 0.1	0.5 \pm 0.1
<hr/>		
GHG balance (tCO ₂ e ha ⁻¹ yr ⁻¹)		
CO ₂	15.5 \pm 8.8	39.8 \pm 2.9
CH ₄	3.3 \pm 1.4	1.9 \pm 0.5
N ₂ O	0.1 \pm 0.1	0.3 \pm 0.2
Fluvial carbon export	1.1 \pm 0.2	1.8 \pm 0.4
Total	20.0 \pm 4.5	43.8 \pm 1.5
<hr/>		
Change in GHG balance due to peatland degradation (tCO ₂ e ha ⁻¹ yr ⁻¹)		23.8 \pm 4.7

698



699

700 Figure 1 | Location of study area, Kampar Peninsula, Sumatra, Indonesia. (a) The
 701 location of research eddy covariance tower sites with satellite image taken from Landsat
 702 8 (Source: <https://earthexplorer.usgs.gov/>). Photographs of the eddy covariance
 703 instruments installed at the top of the tower at (b) the intact peatland and (c) the
 704 degraded peatland. Integrated eddy covariance footprint contour lines from 10% to 80%
 705 in 10% intervals over (d) the intact peatland for June 2017–May 2020 and (e) the
 706 degraded peatland for October 2016–September 2020. Groundwater level, peat
 707 subsidence and soil nitrous oxide flux measurement locations at (f) the degraded
 708 peatland and (g) the intact peatland. An integrated climatologic footprint analysis
 709 indicated that approximately 80% of fluxes were derived within one km in the upwind
 710 direction.



711

712 Figure 2 | Intact and degraded tropical peatland in Sumatra, Indonesia are emitting CO₂

713 and CH₄ to the atmosphere. Cumulative rainfall (solid line), evapotranspiration (dashed

714 line, with cumulative uncertainty) on the left vertical axis and daily groundwater level

715 (dotted line, with standard deviation) on the right vertical-axis at (a) the intact peatland

716 (black) and (b) the degraded peatland (red) for three years (June 2017–May 2020). The

717 blue shaded area shows a convergence of El-Niño and positive Indian Ocean Dipole.

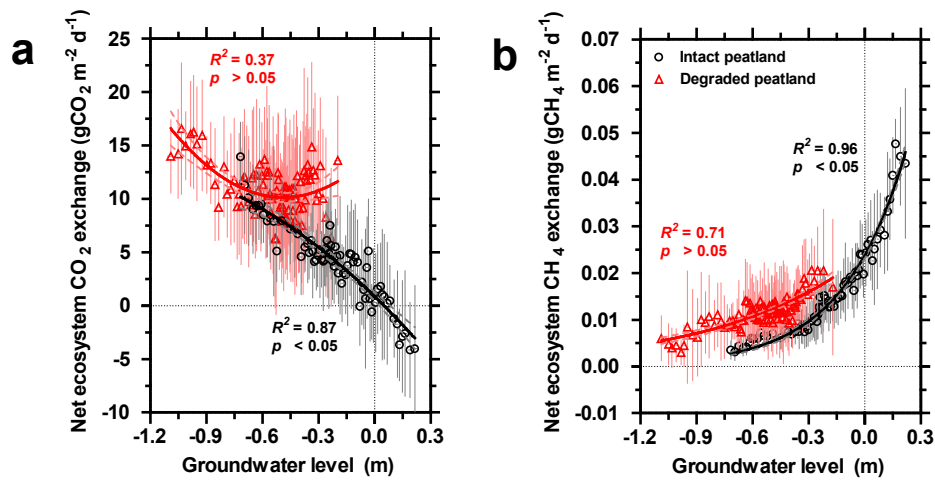
718 Hourly groundwater level were averaged from two locations around the degraded tower

719 site and three locations in the intact peatland. Positive and negative value of

720 groundwater level indicates water level above and below the peat surface, respectively.

721 Rainfall variation controls groundwater level fluctuation. Annual cumulative net

722 ecosystem (c) CO₂ and (d) CH₄ exchanges with cumulative flux uncertainty.



723

724 Figure 3 | The groundwater level is a key driver of both net ecosystem exchange of CO₂

725 and CH₄. Response of gap-filled daily (a) net ecosystem CO₂ exchanges and (b) net

726 ecosystem CH₄ exchanges to the groundwater level at the intact (black) and degraded

727 peatland (red). Data were binned by subgroups of 15 days values of independent

728 variable and corresponding groundwater level and then averaged for the subgroup. The

729 vertical bars represent the standard deviation for the subgroup. The statistical test used

730 a significance level of 5%.

731 Extended Data Table 1 | Characteristics of the intact and the degraded peatland sites in
 732 Sumatra, Indonesia. Value represents average with standard deviation.

Parameter	Intact peatland	Degraded peatland	Method
Tower location	Latitude: 0° 23' 42.735" N Longitude: 102° 45' 52.382" E	Latitude: 0° 41' 58.169" N Longitude: 102° 47' 35.898" E	
Tower height (m)	48	40	
Average Canopy height (m)	32 ± 6	19 ± 6	Permanent sampling plot
Dominant understory species	<i>Nepenthes spp.</i> , <i>Pandanus spp.</i> , <i>Cyrtostachys renda</i>	<i>Dicranopteris linearis</i> , <i>Nepenthes spp.</i> , <i>Pandanus spp.</i> , <i>Cyrtostachys renda</i> , <i>Nephrolepis sp.</i> , <i>Stenochlaena sp.</i> , <i>Blechnum sp.</i>	Permanent sampling plot
Dominant overstory species	<i>Shorea uliginosa</i> , <i>Calophyllum ferrugineum</i> , <i>Syzygium spp.</i> , <i>Camposperma macrophylla</i> , <i>Tetramerista glabra</i> , <i>Palaquium burckii</i>	<i>Syzygium claviflorum</i> , <i>Shorea teysmanniana</i> , <i>Stemonurus secundiflorus</i> , <i>Blumeodendron kurzii</i> , <i>Horsfieldia crassifolia</i> , <i>Macaranga pruinosa</i>	Permanent sampling plot
Surface peat type (0 – 0.5 m)	Fibric	Fibric	Von Post's scale for peat humification
Peat depth (m)	9.0 ± 1.0	8.4 ± 0.6	Manual subsidence pole
Surface peat bulk density (0 – 0.5 m) (g cm ⁻³)	0.08 ± 0.03	0.09 ± 0.03	Gravimetric method
Carbon concentration (%) (0 – 0.5 m)	50.2 ± 2.0	55.4 ± 0.5	Loss on Ignition method
Total Nitrogen concentration (%) (0 – 0.5 m)	1.3 ± 0.2	1.5 ± 0.3	Kjeldahl method
Nitrate (ppm) (0 – 0.5 m)	1167 ± 509	1588 ± 725	Titration method
Ammonium (ppm) (0 – 0.5 m)	475 ± 130	654 ± 184	Titration method

733

734 Extended Data Table 2 | Annual average with standard deviation of environmental variables at the intact and the
 735 degraded peatland in Sumatra, Indonesia. Groundwater level were averaged from four locations around the degraded
 736 tower site and nine locations in the intact peatland. Annual peat subsidence were derived from eight locations in the intact
 737 peatland from December 2017-June 2020 and four locations around the degraded tower site from December 2016-
 738 September 2020.

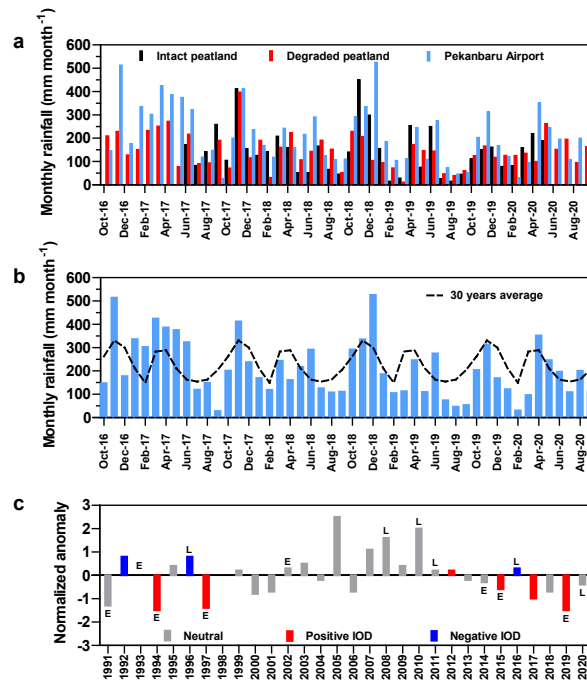
	Intact peatland			Degraded peatland			
	June 2017 - May 2018	June 2018 - May 2019	June 2019 - May 2020	Oct 2016 - Sep 2017	Oct 2017 - Sep 2018	Oct 2018 - Sep 2019	Oct 2019 - Sep 2020
Photosynthetic photon flux density ($\mu\text{mol m}^{-2} \text{s}^{-1}$)	680 \pm 177	667 \pm 170	644 \pm 164	748 \pm 200	727 \pm 191	717 \pm 183	729 \pm 194
Air temperature ($^{\circ}\text{C}$)	26.6 \pm 0.9	26.9 \pm 0.9	27.0 \pm 0.9	26.8 \pm 0.9	26.7 \pm 0.9	27.2 \pm 0.9	28.1 \pm 1.3
Vapor pressure deficit (hPa)	7.3 \pm 2.8	7.4 \pm 2.8	7.2 \pm 2.8	7.4 \pm 2.9	7.2 \pm 2.8	7.5 \pm 3.0	8.0 \pm 3.1
Soil temperature at 0.15 m below peat surface ($^{\circ}\text{C}$)	27.2 \pm 0.5	27.4 \pm 0.1	28.1 \pm 0.1				
Soil temperature at 1.5 m below peat surface ($^{\circ}\text{C}$)	26.2 \pm 0.2	25.7 \pm 0.1	25.7 \pm 0.1	28.4 \pm 0.3	27.7 \pm 0.2	27.2 \pm 0.1	27.5 \pm 0.1
Cumulative rainfall (mm)	2020	1756	1496	2142	1837	1325	1763
Cumulative evapotranspiration (mm)	1575	1617	1575	1372	1371	1539	1496
Groundwater level (m)	-0.15 \pm 0.16	-0.21 \pm 0.20	-0.47 \pm 0.18	-0.51 \pm 0.13	-0.67 \pm 0.10	-0.74 \pm 0.23	-0.73 \pm 0.14
Subsidence rate (cm yr^{-1})		3.3 \pm 0.7			4.2 \pm 1.3		

739

740 Extended Data Table 3 | Annual net ecosystem CO₂, CH₄ and H₂O (evapotranspiration) exchanges at the intact and the
 741 degraded peatland in Sumatra, Indonesia. Table shows annual estimates from different gap-filling approaches with
 742 uncertainties associated with random error in measurements, friction velocity quality-control criteria and gap-filling
 743 approach. The bold values represent the average with standard deviation from different approaches.

	Intact peatland			Degraded peatland			
	June 2017-May 2018	June 2018-May 2019	June 2019-May 2020	Oct 2016-Sep 2017	Oct 2017-Sep 2018	Oct 2018-Sep 2019	Oct 2019-Sep 2020
Net ecosystem CO ₂ exchange (tCO ₂ ha ⁻¹ yr ⁻¹)							
Marginal distribution sampling	14.9 ± 5.5	8.8 ± 5.5	25.8 ± 6.2	42.2 ± 6.5	42.3 ± 6.4	46.1 ± 6.4	48.3 ± 7.0
Random forest	10.8 ± 2.2	9.2 ± 2.3	25.5 ± 2.3	35.4 ± 2.5	38.7 ± 2.1	40.4 ± 2.0	41.8 ± 1.7
Artificial neural network	10.0 ± 2.4	9.3 ± 2.4	25.6 ± 2.6	31.2 ± 2.3	35.1 ± 2.4	36.3 ± 2.5	39.3 ± 2.2
Average	11.9 ± 3.7	9.1 ± 3.7	25.6 ± 4.1	36.3 ± 4.2	38.7 ± 4.1	40.9 ± 4.1	43.1 ± 4.4
Net ecosystem CH ₄ exchange (kg CH ₄ ha ⁻¹ yr ⁻¹)							
Marginal distribution sampling	89.7 ± 6.7	86.1 ± 5.8	35.1 ± 3.7	55.1 ± 4.0	34.2 ± 3.6	43.8 ± 6.6	29.5 ± 6.6
Mean Diurnal Course	95.3 ± 6.4	91.4 ± 5.6	39.9 ± 3.8	57.3 ± 3.8	40.2 ± 3.6	45.2 ± 6.6	34.6 ± 6.6
Average	92.5 ± 6.5	88.8 ± 5.7	37.5 ± 3.8	56.2 ± 3.9	37.2 ± 3.6	44.5 ± 6.6	32.1 ± 6.6
Evapotranspiration (mm yr ⁻¹)							
Marginal distribution sampling	1520 ± 87	1537 ± 87	1522 ± 86	1263 ± 75	1439 ± 80	1485 ± 86	1398 ± 77
Random forest	1524 ± 91	1630 ± 95	1557 ± 93	1439 ± 78	1482 ± 79	1614 ± 87	1471 ± 76
Artificial neural network	1568 ± 91	1610 ± 91	1563 ± 88	1414 ± 84	1191 ± 73	1519 ± 88	1619 ± 85
Average	1537 ± 90	1592 ± 91	1547 ± 89	1372 ± 79	1371 ± 77	1539 ± 87	1496 ± 80

744



745

746 Extended Data Figure 1 | Rainfall and climate indices in the study area. (a) Comparison

747 between monthly rainfall in the intact peatland, the degraded peatland, and Pekanbaru

748 Airport (Riau, Sumatra) which represent the closest available long-term rainfall

749 measurements in the region, (b) comparison between rainfall in Pekanbaru Airport

750 during study period and long-term average of 30 years from 1991-2020 and (c) the

751 relationship between rainfall in Pekanbaru Airport and climate indices represented with

752 normalized anomaly of rainfall amount in July-August-September; “E” and “L” indicate El

753 Niño and La Niña years, respectively; red and blue bars indicate positive and negative

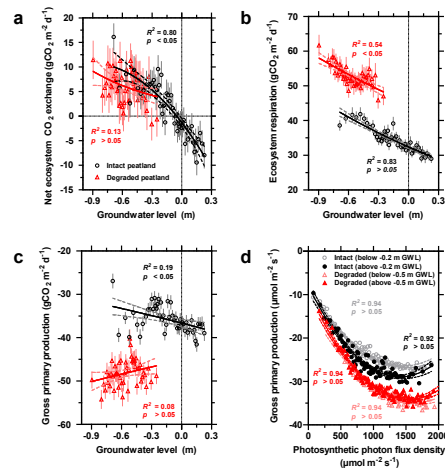
754 IOD years, respectively; grey bar shows neutral years. The Dipole Mode Index data

755 were obtained from <http://psl.noaa.gov/>. Southern Oscillation Index were obtained from

756 <https://www.cpc.ncep.noaa.gov/data/indices/soi>. The extended dry period in July-

757 September 2019 coincided with a convergence of El-Niño and positive Indian Ocean

758 Dipole.



759

760 Extended Data Figure 2 | Groundwater level controls net ecosystem CO₂ exchanges.

761 Response of (a) measured half-hourly net ecosystem CO₂ exchanges, (b) ecosystem

762 respiration and (c) gross primary production to the groundwater level at the intact

763 peatland (black) and the degraded peatland (red). Measured half-hourly data were

764 grouped into daytime (06:00-18:00 hr) and nighttime (18:00-06:00 hr). Then each

765 dataset were sorted into 50 classes of groundwater level and averaged for each class.

766 Subsequently, the daily value was computed from the average of daytime and nighttime

767 values. The vertical bars represent the standard deviation for each class. (d) Response

768 of the 09:00-15:00 hr gross primary production to photosynthetic photon flux density at

769 two different groundwater level classes based on median value. The selection of time

770 window may have created biases in actual response curves of both ecosystems, but

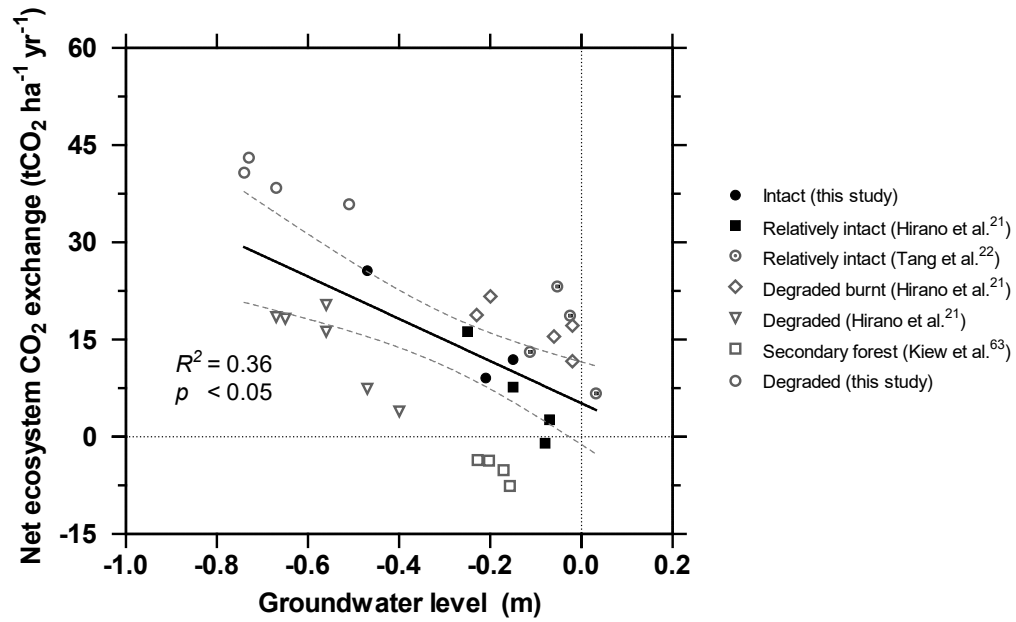
771 this bias would not change the interpretation. Data were binned by subgroups of 100

772 half-hourly values of gross primary production and corresponding photosynthetic photon

773 flux density and then averaged for each class. All statistical tests used a significance

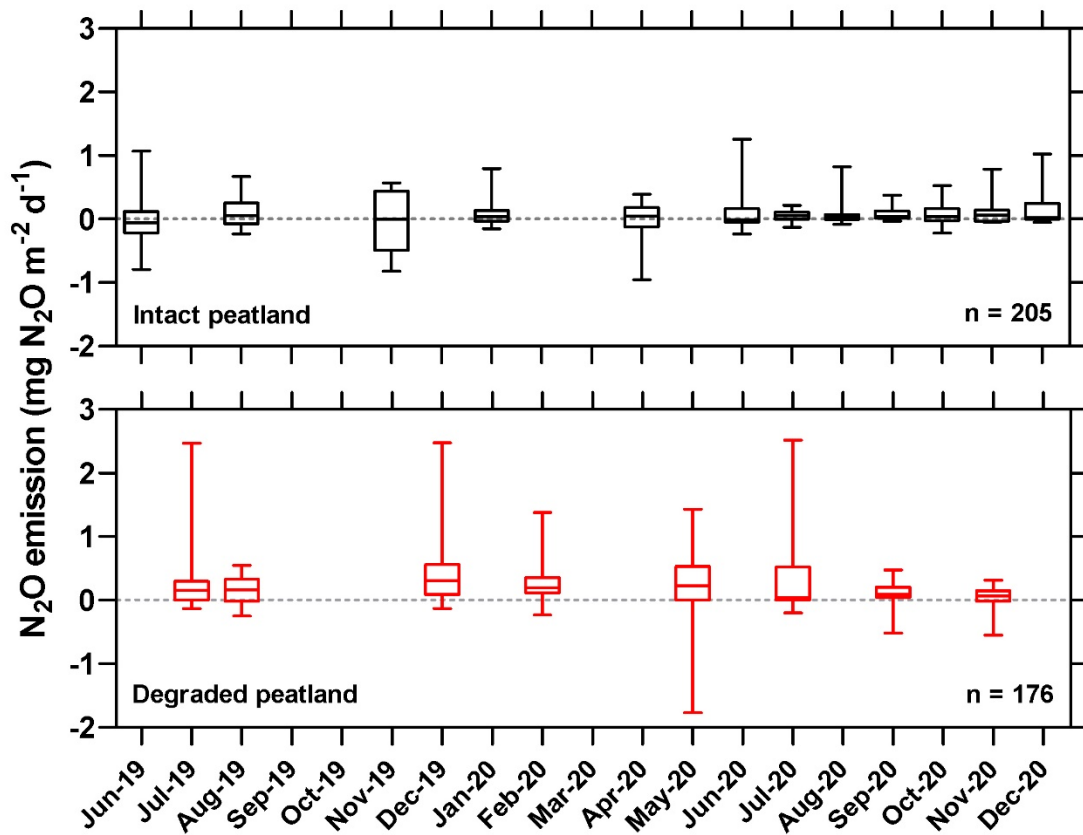
774 level of 5%. Intact peatland shows higher light-use efficiency when groundwater level is

775 shallow.

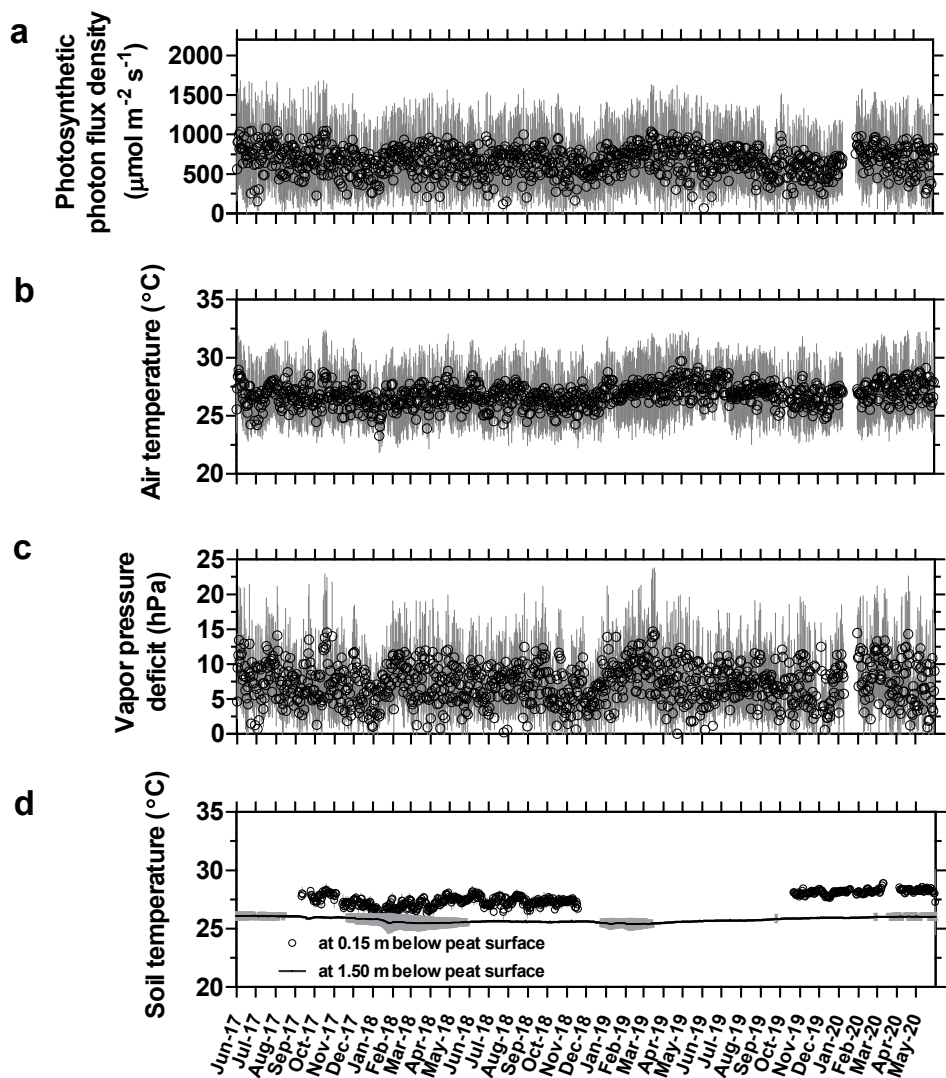


776

777 Extended Data Figure 3 | Relationship between net ecosystem CO₂ exchange and
 778 groundwater level. Data are derived from 30 site-years of the eddy covariance
 779 measurements in tropical peatlands in Southeast Asia. Solid line indicates linear relation
 780 with dashed lines for 95% confidence interval. The statistical test used a significance
 781 level of 5%. Carbon dioxide emissions increase with lower groundwater level.

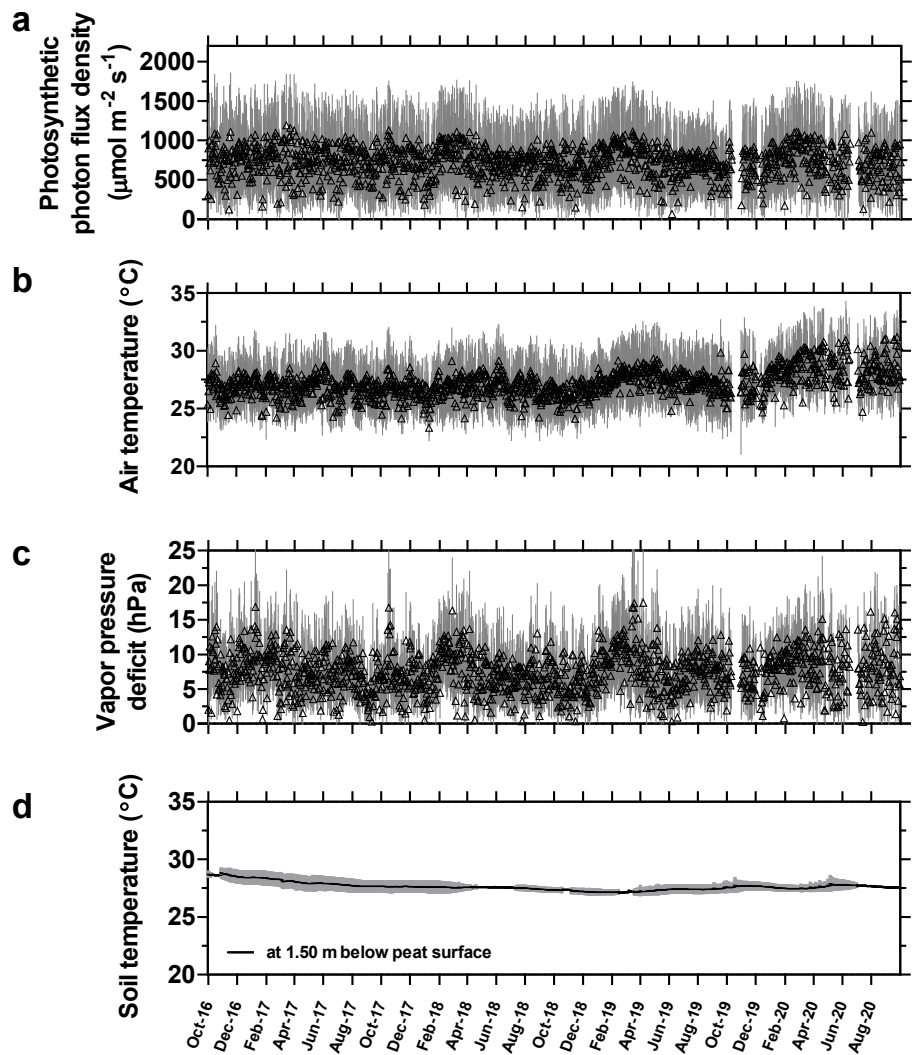


782
 783 Extended Data Figure 4 | The intact peatland emits lower soil N₂O emissions than the
 784 degraded peatland. Measurements of N₂O emissions at the intact peatland (black) and
 785 the degraded peatland (red). The boxes show the median value and the interquartile
 786 range, and whiskers denote the full range of all chambers. The n values represent total
 787 number of soil N₂O flux measurements.



788

789 Extended Data Figure 5 | The environmental variables at the intact peatland during the
 790 study period. Variations in daily (a) photosynthetic photon flux density, (b) air
 791 temperature, (c) vapor pressure deficit and (d) soil temperature at the intact peatland.
 792 The vertical bar represents standard deviation.



793

794 Extended Data Figure 6 | The environmental variables at the degraded peatland during
 795 the study period. Variations in daily (a) photosynthetic photon flux density, (b) air
 796 temperature, (c) vapor pressure deficit and (d) soil temperature at the degraded
 797 peatland. The vertical bar represents standard deviation.