New insights into diel to interannual variation in carbon dioxide emissions from lakes and reservoirs

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Abstract

Accounting for temporal changes in carbon dioxide (CO_2) emissions from freshwaters remains a challenge for global and regional carbon budgets. Here, we synthesize 171 site-months of eddy covariance flux measurements of CO_2 from 13 lakes and reservoirs in the Northern Hemisphere (NH) and quantify the magnitude and dynamics at multiple temporal scales. We found pronounced diel and sub-monthly oscillatory variations in CO_2 flux at all sites. Diel variation converted sites to daily net sinks of CO_2 in only 11% of site-months. Upscaled annual emissions had an average of 25% (range 3-58%) interannual variation. Given temporal variation remains under-represented in inventories of CO_2 emissions from lakes and reservoirs, revisions in CO_2 flux are needed using a better representation of sub-daily to interannual variability. Constraining short- and long-term variability is necessary to improve detection of temporal changes of CO_2 fluxes in response to natural and anthropogenic drivers. New insights into diel to interannual

² variation in carbon dioxide emissions from

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6 Abstract

- 7 Accounting for temporal changes in carbon dioxide (CO₂) emissions from freshwaters remains a
- 8 challenge for global and regional carbon budgets. Here, we synthesize 171 site-months of eddy
- 9 covariance flux measurements of CO₂ from 13 lakes and reservoirs in the Northern Hemisphere
- 10 (NH) and quantify the magnitude and dynamics at multiple temporal scales. We found
- 11 pronounced diel and sub-monthly oscillatory variations in CO_2 flux at all sites. Diel variation
- 12 converted sites to daily net sinks of CO_2 in only 11% of site-months. Upscaled annual emissions
- had an average of 25% (range 3-58%) interannual variation. Given temporal variation remains
- 14 under-represented in inventories of CO_2 emissions from lakes and reservoirs, revisions in CO_2
- 15 flux are needed using a better representation of sub-daily to interannual variability. Constraining
- 16 short- and long-term variability is necessary to improve detection of temporal changes of CO_2
- 17 fluxes in response to natural and anthropogenic drivers.

18 Plain Language Summary

19 Lakes and reservoirs around the world are likely a major component of the global carbon cycle.

- 20 Recent syntheses of measurements find their contributions to be on the order of 2-6% of total
- 21 global fossil fuel emissions. However, these estimates are primarily derived from compilations
- with low frequency of sampling, from a few times a year up to weekly, often restricted to a
 single season, and with limited regard to year-to-year variations. Here, we conduct the first
- single season, and with limited regard to year-to-year variations. Here, we conduct the first
 analysis of a globally distributed network of sub-hourly, multi-year lake and reservoir carbon
- 25 dioxide emissions. These measurements were made using eddy-covariance flux towers, which
- 26 continuously sample these emissions year-round. Across our 13 study sites, we found nighttime
- 27 emissions regularly exceeding daytime emissions and persistent sub-monthly oscillations
- 28 regardless of lake size or nutrient status. For sites with multiple years of data, we found an
- 29 average 25% variation in estimated annual emissions depending on the year chosen. Together,
- 30 these results point to a need for improved, systematic sub-weekly sampling of freshwater
- 31 systems to better understand dynamics of freshwater ecosystems, reduce uncertainty in landscape
- 32 to global carbon budgets, and project changes to atmospheric greenhouse gas burdens in a
- 33 warming climate.
- 34 Index terms (5): 0428 Carbon cycling, 0426 Biosphere/atmosphere interactions, 0438 Diel,
- 35 seasonal, and annual cycles, 0434 Data sets, 0458 Limnology
- 36 Keywords (6): eddy covariance; freshwater systems; lakes; reservoirs; carbon flux; synthesis
- 37 Key Points:
- First synthesis of high-frequency aquatic freshwater carbon dioxide flux observations
 reveals large diel, sub-annual, and interannual variation
- 40 At all sites, nighttime emissions are larger than daytime, sub-monthly oscillations are
 41 present, and year-to-year variation averaged 25%
- 42 Under-sampling of these dynamics leads to potential bias in estimates of contribution of
 43 freshwater systems to the global carbon cycle

44 **1.** Introduction

45 The global carbon budget is rapidly changing in response to human emissions, climatic modes of

46 variability, and global changes (Friedlingstein *et al.*, 2020; Hanson *et al.*, 2006). Prior studies

47 have estimated that 0.14-0.64 Pg C-CO₂ is annually released to the atmosphere through lakes and

48 reservoirs (Aufdenkampe et al., 2011; Ciais et al., 2013; Cole et al., 1994, 2007; Drake et al.,

49 2018; Holgerson *et al.*, 2016; Raymond *et al.*, 2013), offsetting 10-40% of the global terrestrial

50 land sink. However, most of these estimates are made with relatively limited sampling, generally

- 51 constrained to the open-water or summer season during the daytime, and with limited
- 52 consideration of interannual and shorter-scale variation (Butman *et al.*, 2018; Ran *et al.*, 2021).

53 Underrepresentation of temporal CO₂ flux variability in existing CO₂ flux inventories may bias

54 estimates of lake CO₂ emissions (Deemer *et al.*, 2016; Klaus *et al.*, 2019). For example, recent

55 studies have found nighttime emissions exceeding daytime emissions or uptake in reservoirs (Liu

56 *et al.*, 2016) and rivers (Gómez-Gener *et al.*, 2021). A lack of frequent and long-term CO₂

57 observations also limits our ability to differentiate natural CO₂ flux variations from the

58 consequences of anthropogenic perturbations to freshwater biogeochemistry and predict future

59 CO₂ responses to the global change (Hasler *et al.*, 2016). Decadal-scale time series that capture

60 sub-annual variability of the CO₂ flux remain rare (Finlay *et al.*, 2019; Huotari *et al.*, 2011).

61 Traditional in-situ aquatic sampling methods for CO2 concentrations and derived fluxes in

62 natural and artificial freshwaters also come with high uncertainty (Baldocchi et al., 2020; Golub

63 *et al.*, 2017).

64 Advances in the past several decades, however, have enabled more long-term, continuous high-

65 frequency (hourly) measurements in freshwater ecosystems, which are capable of capturing the

66 dynamics of air-water fluxes at time scales of hours to years (Eugster *et al.*, 2003; Huotari *et al.*,

67 2011; Morales-Pineda *et al.*, 2014). At these time scales, CO₂ fluxes have been shown to respond

to variations in wind speed and direction (Podgrasjek *et al.*, 2015), carbonate equilibria (Atilla *et*

69 *al.*, 2011), ecosystem metabolism (Provenzale *et al.*, 2018), convective mixing (Eugster *et al.*,

70 2003; Mammarella et al, 2015), internal waves (Heiskanen et al., 2014), ice phenology (Reed et

71 *al.*, 2018), and hydrological and carbon inflows (Rantakari *et al.*, 2005; Weyhenmeyer *et al.*,

- 2015). These sources of variation may be overlooked by low-frequency and season-restricted
 sampling that dominate freshwater science (Desai *et al.*, 2015).
- 74 Many previous studies were conducted using eddy covariance (EC) flux towers, which have
- 75 gained prominence for use in freshwaters (Vesala *et al.*, 2012). The eddy covariance method
- 76 directly measures air-water CO₂ fluxes within an ecosystem-scale footprint (Vesala *et al.*, 2006).
- 77 While its application over lakes has mostly covered short periods of time (e.g., Eugster *et al.*,
- 78 2003; Podgrajsek et al., 2015; Vesala et al., 2006), an increasing number of sites are now
- 79 measuring lake-atmosphere fluxes continuously over multiple years (Franz et al., 2016; Huotari
- 80 *et al.*, 2011; Mammarella *et al*, 2015; Reed *et al.*, 2018).

81 This recent growth of continuous measurements affords an opportunity to investigate the relative 82 magnitude and importance of diel to interannual variation in lake and reservoir exchanges and 83 discuss pathways to incorporating these insights into improving quantification of freshwaters in the global carbon cycle. Here, we quantify diel to inter-annual dynamics of CO2 fluxes, directly 84 85 measured by eddy covariance from 13 lakes and reservoirs representing a broad nutrient-humic 86 spectrum of sites in the Northern Hemisphere. Our main aim was to identify modes of CO₂ flux 87 variability missed by infrequent sampling that may lead to biases in estimates of annual CO₂ flux 88 from lakes and reservoirs.

89 2. Materials and Methods

90 2.1 Study sites

91 Data on air-water CO_2 exchange and meteorological drivers were acquired from study sites 92 across the Northern Hemisphere with at least one season of observations between 2005-2015, of 93 which 13 were retained here for analysis (Table 1 and S1). The remaining submitted sites were 94 withheld for challenges in meeting uncertainty and gap filling criteria (see Supplemental 95 Methods). This analysis represents the largest synthesis of lake and reservoir eddy-covariance 96 CO₂ flux observations to-date. These sites were collected based on organization of a workshop 97 (Desai et al., 2015) and an open call through listservs. Selected sites included 9 lakes and 4 98 reservoirs, mostly located between 40-68°N latitude, coinciding with the largest area of Earth's

99 covered with lakes. Most sites had data available over multiple seasons, but only a few also had

100 measurements during winter ice cover. Lake area ranged from 0.036 km^2 to 623 km^2 (median:

101 15.2 km²), with median mean depth of 6 m (range: 0.6 to 11 m); most developed a seasonal

102 thermocline and were dimictic or monomictic (Table S1). Two water bodies had a significant

103 fraction of submerged and emergent macrophytes (SE-Tam and DE-Zrk) within the footprint of

104 the flux tower.

105 2.2 Measurements

106 The eddy covariance technique directly measures the exchange of momentum, heat and matter 107 (water vapor, CO₂, or other trace gases) at the air-water interface and is considered the most 108 direct method of measuring surface exchanges with the atmosphere (Vesala *et al.*, 2006). The 109 measured fluxes are integrated across the EC flux footprint (i.e., the upwind area "seen" by the 110 tower), capturing all sources and sinks of turbulent exchanges. The flux towers were located on 111 floating platforms, lake shoals or islands, or on shore depending on the site (Table S1). The high 112 frequency (10 or 20 Hz) measurements were made with open-path or closed-path infrared gas 113 analyzers and processed into half-hourly average fluxes by site PIs according to standard 114 methods (Aubinet et al, 2012). The towers were additionally equipped with instruments 115 providing half-hourly to hourly measurements of biophysical variables (e.g. net radiation, air 116 temperature and humidity, photosynthetically-active radiation (PAR), 2-D wind direction and 117 speed, water temperature, aquatic CO₂ or O₂ concentration, water level), although data 118 availability and frequency varied among the sites. Data were harmonized to uniform formats and 119 units, screened for fetch, and de-spiked using a common flux post-processing standard 120 (Pastorello et al., 2020) to reduce cross-site flux uncertainty due to methodological differences. 121 All data were submitted to the Environmental Data Initiative repository (Golub et al., 2021).

122 2.3 Flux data processing

After despiking and quality control, the half-hourly averages of CO_2 fluxes retained 3-90% of observations during measurement periods (Table S1). A larger fraction of gaps relative to

125 terrestrial systems were caused by the exclusion of out-of-lake and mixed tower footprints and

- 126 flagging by quality control algorithms applied to flux computation. Despite these gaps, the
- 127 available data provide an unprecedented number of direct CO₂ flux observations (171 site-
- 128 months and 3,832 site-hours in total) that captured flux variability at multiple time scales (i.e.
- diel and seasonal) that are usually poorly represented in traditional limnological studies. For
- 130 further analysis, all quality-controlled flux observations were identified, with missing
- 131 observations gap-filled, uncertainty assessed for each time point, and continuous flux time series
- 132 aggregated for diel, seasonal (sub-annual), and annual estimates.
- 133 Flux data were gap-filled using marginal distribution sampling (MDS) (Reichstein *et al.*, 2005)
- 134 within REddyProc (Wutzler *et al.*, 2018). The MDS approach, which applies both a moving
- 135 window and look-up table multiple imputation approach, used observations of shortwave
- 136 incoming radiation, air temperature, and vapor pressure deficit (VPD) to fill data gaps.
- 137 The CO₂ uncertainty measured with the EC approach exhibits a variety of systematic and random
- 138 errors, several of which can be quantified (Richardson *et al.*, 2012; Rannik *et al*, 2016). Random
- errors in half-hourly averages of CO_2 flux over lakes can range from 26% to 40%, with
- 140 uncertainty more pronounced in eutrophic systems (Jammet et al., 2017; Mammarella et al.,
- 141 2015). The uncertainty in half-hourly flux averages was estimated as the standard deviation of
- 142 observations used for gap-filling with the MDS algorithm.
- 143 The mean diel change of CO_2 and associated uncertainty was calculated by deriving the average
- and one standard deviation from all observations within the same month for each half-hour of the
- 145 day. Lake-months with <15 observations per half-hourly average were discarded to avoid
- 146 influences of unreliable means on calculated statistics. The monthly-averaged flux amplitudes
- 147 were calculated as a difference between 95th percentile of nighttime observations (when
- 148 shortwave incoming radiation was <10 W m⁻²) and 5th percentile of daytime observations (>10
- 149 W m⁻²).
- 150 Daily CO₂ flux values were calculated by averaging the 48 half-hourly fluxes. Daily CO₂ flux
- 151 observations were binned according to Sturge's formula (i.e. log2(N)+1 where N represents the
- 152 number of observations per lake) to accurately represent frequencies of flux distribution. To
- 153 avoid the influence of extreme outliers on bin resolution, the bins were scaled to $>1^{st}$ and $<99^{th}$
- 154 percentiles. Annual CO₂ flux sums were calculated for the duration of ice-free season by

- summing daily fluxes. The ice-free period was determined from observational ice-on and ice-off
- 156 data or predicted from 0.5°x0.5° gridded mean monthly air temperature (2000-2010) at a given
- 157 latitude (Wei *et al.*, 2014). All but two sites (US-RBa and LA-NT2) had seasonal ice cover.

158 2.4 Data analysis

159 We analyzed the half-hourly CO_2 fluxes and three major groups of biophysical covariates. The 160 first group included variables related to wind forcing acting on the water surface (i.e. friction 161 velocity, wind speed, momentum flux). The second group encompassed the variables related to 162 temperature cycles and proxies of energy in the system (i.e. air temperature, water temperature, 163 ΔT (T_{water}-T_{air}), sensible and latent heat fluxes). The last group included the variables associated 164 with solar radiation -- proxies for primary productivity (i.e. $\Delta p CO_2 (\Delta p CO_{2water} - \Delta p CO_{2air})$, 165 PAR). Variables were included if they were measured at the site (Table S1). The robust linear 166 least-squares second-order polynomial model with bisquare weighting method was used to 167 compare bivariate relationships across lakes. To estimate confidence intervals around estimated 168 parameters and curves, we bootstrapped residuals with 1,000 iterations. The analyses were 169 performed with MATLAB ver. R2018a using Curve Fitting Toolbox. To determine the 170 standardized difference between two means with repeated unpaired measurements and 171 imbalanced population sizes, we used the Cohen's d test. The mean difference between the mean 172 daily CO₂ fluxes was divided by the pooled variance. A coefficient d of 0.20, 0.50, 0.80 173 indicates small, medium, and large differences, respectively. One macrophyte-covered reservoir 174 with distinct fluxes is provided in the Supplemental Materials.

175 **3. Results**

176 3.1 Magnitude of CO₂ fluxes from lakes and reservoirs

177 Study sites represented a wide range of nutrient-color statutes and physical characteristics of

178 water bodies, and as a result spanned a range of daily CO₂ fluxes, though with some common

elements (Fig. 1). The mean daily CO₂ flux across all sites was 0.43 ± 0.34 µmol m⁻² s⁻¹ (range: -

- 180 0.075 to 1.25 μ mol m⁻² s⁻¹) with only 6% of observations indicating neutral fluxes or net CO₂
- 181 uptake. The spread of time-resolved fluxes varied 102-798 % of the site-specific daily mean (Fig.
- 182 1). Reservoirs had smaller but more variable fluxes relative to the lakes $(0.32\pm0.71 \text{ vs}, 0.41\pm0.31)$
- $183 \mu mol m^{-2} s^{-1}$), though the reservoir sample size is smaller and more geographically restricted.
- 184 Two thirds of sites had at least 66% of daily fluxes within the cross-site flux mean ± 1 SD
- 185 (Cohen's d: $0.02 \le d \le 0.76$).

186 3.2 Temporal variability of CO₂ fluxes from lakes and reservoirs

187 Averaged diel CO₂ curves had regular patterns of daytime minima and nighttime maxima across

all sites in most months (Fig. 2a). Daytime hourly fluxes were on average 35% (range: 7-60%)

189 lower than nighttime fluxes, though in 94% of site days, those were still net positive emissions.

190 Despite the commonly observed daytime CO₂ flux dip, the flux decrease was large enough to

- 191 convert our sites to daily net sinks of CO₂ in only 11% of site-months (Fig. 2a, Table 1). The
- 192 mean uncertainty of diel CO₂ was strongly influenced by extreme observations, with 192% mean
- 193 uncertainty, but only 79% median uncertainty (Fig. 2b).

194 Maximum diel flux amplitudes typically occurred in July and August and ranged 0.24-1.09 µmol

195 $m^{-2} s^{-1}$. Relative to the summer amplitudes, shoulder season CO₂ flux amplitudes were on

average 44-49% smaller in May and September and 26-37% in April and October. Diel variation

197 was negligible at both ends of the ice-free season (Fig. 2a).

198 Monthly to seasonal CO₂ flux variability was nearly twofold compared to diel flux variation

199 (Table 1). Surprisingly, we found frequent sub-monthly (20-30-day) oscillations across all water

200 bodies, regardless of the system's physical or biogeochemical conditions (Fig. 3a). While most

201 site-level oscillations fluctuated around the CO₂ flux averages, for some, amplitudes scaled with

- 202 flux minima and maxima (Fig. S1).
- 203 Sites with multi-year data had relatively consistent sub-annual patterns across years, although the

204 timing and amplitudes of sub-monthly oscillations varied among lake-years. When integrated

- 205 over time-resolved daily CO₂ fluxes, both sub-monthly and sub-annual modes of variability
- accounted for two thirds of the site-level daily CO₂ flux variability (range 10-190%). Mean and
- 207 median uncertainty were 167% and 67% of mean daily CO₂ flux, respectively (Fig. 3b).

- 208 Once scaled to ice-free season annual emissions, and assuming zero fluxes during ice cover, we
- 209 found all water bodies were net sources of CO₂, despite missing any ice off/on related fluxes
- 210 (Table 1). The cross-site mean and standard deviation of 23 site-years was 95 ± 49 gC m⁻² yr⁻¹
- 211 (range: 14-224 gC m⁻² yr⁻¹). Inter-annual variability (IAV) was calculated as a standard deviation
- of annual CO₂ flux for each site with multi-year data. The mean cross-site IAV was 22 gC m⁻² yr⁻
- 213 1 (25%) and ranged 4-44 gC m⁻² yr⁻¹ (3-58%).

3.3 Drivers of CO₂ fluxes from lakes and reservoirs

215 While the continuous data allowed capturing CO₂ fluxes variability at different temporal scales, 216 we still had a limited capacity to attribute which factors and processes governed the observed 217 patterns of CO₂ flux. We found small standardized differences between CO₂ fluxes among site 218 groups belonging to the three humic states (d<0.01), medium differences between oligotrophic 219 and eutrophic states (d=0.24), and large CO₂ differences between mesotrophic and oligotrophic 220 states (d=0.66), and between mesotrophic and eutrophic states (d=0.72) (Fig. 4). Commonly 221 observed biophysical covariates explained an average of 32% of variance in half-hourly CO₂ 222 fluxes (Fig. 4g). Wind-related variables were identified as key to explaining CO₂ flux variability 223 in eight out of 13 sites. Biophysical variables related to exchanges of heat at the air-water 224 interface, particularly air-water temperature difference (ΔT) and turbulent energy exchange 225 (latent and sensible heat fluxes) correlated with CO₂ flux. The fitted regressions were non-linear 226 and highly variable across sites, owing to ecosystem differences and presence of confounding 227 factors (e.g. differential responses to co-dependent covariates).

4. Discussion

4.1 Unresolved temporal variation in CO2 fluxes

230 We demonstrated that CO₂ fluxes from lakes and reservoirs exhibited significant and consistent

- 231 variability at diel to (inter)-annual scales, which could comprise unresolved sources of
- uncertainty or bias in current estimates of annual CO₂ fluxes from infrequent and season-

- 233 restricted sampling. Though our study lakes were not randomly selected and cannot be directly
- used to upscale (Stanley *et al.*, 2019), they were broadly reflective of common mid-latitude
- 235 freshwater systems in a broad range of humic-status and mixing regimes. Additional
- 236 considerations for sampling across lake size and catchment area (Hanson et al., 2007; Holgerson
- *et al.*, 2016) and hydrological setting (Jones *et al.*, 2018) would be required to design a
- 238 representative estimate for global upscaling.
- 239 Instead, we were able to investigate the role of temporal variation on a range of systems that
- 240 broadly reflect many lakes and reservoirs. Our reported continuous daily fluxes corresponded
- 241 with the upper end (88th percentile) of previously published flux magnitudes (Tables S2). The
- 242 observed temporal variation suggests that infrequent and time of day or year scheduling
- 243 restricted sampling may add a significant source of underestimation bias in existing inventories
- of CO₂ fluxes from lakes and reservoirs of similar type and size (Klaus *et al.*, 2019).
- 245 In particular, we note significant diel variation found in all study sites, with routinely higher
- emissions at night, consistent with a recent study over rivers (Gómez-Gener et al., 2021). The
- 247 diel reduction of dissolved CO₂ concentrations and fluxes are often associated with ecosystem
- 248 metabolism (Hanson *et al.*, 2003) and was supported by negative correlations with PAR (Fig.
- 249 4g). Water temperature (Provenzale et al., 2018), carbonate equilibria fluctuations (Atilla et al.,
- 250 2011), water-side convection (Mammarella *et al*, 2015; Podgrajsek *et al.*, 2015), and internal
- 251 waves (Heiskanen *et al.*, 2014) can additionally govern diel CO₂ dynamics. Our observed diel
- amplitudes were within 21-43% of sub-hourly flux amplitudes derived from dissolved CO₂
- 253 concentrations (Hanson et al., 2003; Morales-Pindea et al., 2014) or previously published EC-
- 254 measured fluxes (Liu *et al.*, 2016; Vesala *et al.*, 2006). Our results support the notion that
- existing global lake carbon budgets are underestimates of net emissions.
- 256 We also found common sub-monthly modes of CO₂ flux variability across all of our sites.
- 257 Similar oscillations in the continuous observations have been reported for dissolved CO₂ (Atilla
- *et al.*, 2011; Huotari *et al.*, 2009; Morales-Pineda *et al.*, 2014; Vachon and del Giorgio, 2014)
- and CO₂ fluxes (Franz *et al.*, 2016), indicating the prevalence of oscillatory patterns in CO₂ time
- series at both sides of the air-water interface. Oscillations have previously been attributed to the
- 261 interplay of wind forcing (Liu et al., 2016), upwellings of CO₂-rich waters (Morales-Pineda et

262 al., 2014), biologically-driven (metabolic and trophic) changes in carbonate equilibria (Atilla et

263 *al.*, 2011), convective mixing (Huotari *et al.*, 2009) and water temperature (Atilla *et al.*, 2011).

However, this is the first study to find a consistent pattern in a wide range of systems, regardless

- 265 of size. We also observed changes to the prevalence of underlying sub-monthly CO₂ flux
- 266 oscillations through the year at several sites, likely reflecting seasonal ecosystem changes, such
- as spring/fall turnover (Baehr et al., 2004), radiative and heat exchanges (Heiskanen et al., 2014),
- and hydrological inflows (Vachon *et al.*, 2017).

4.2 Implications for the global carbon budget

270 After our daily fluxes were scaled to annual totals, our estimates of annual CO₂ emissions were 271 in the upper end reported for lakes and reservoirs (Table S2). All systems were sources of CO_2 in 272 most years, though there have been sites that reported significant carbon sinks (e.g., Shao et al. 273 2015; Reed et al., 2018) and additional propagation of uncertainty from data gap filling and 274 filtering (e.g., of nighttime uptake) can push some of our study sites toward sinks, though 275 weakly. While our lakes are not fully representative for all lakes on Earth, we postulate that 276 improved temporal resolution of site-level CO₂ fluxes is one of the sources of differences 277 between this study and published annual fluxes (Table S2). The results also imply that a 278 proposed recommended number of samples per year (4-8) (Klaus et al., 2019; Natchimuthu et 279 al., 2017) is likely insufficient to constrain annual CO₂ fluxes from lakes and reservoirs. Rather, 280 approaches to increase nighttime samples and open-water season weekly or higher-frequency 281 sampling would increase the accuracy of annual estimates, given our observed diel and sub-282 monthly variations.

283 Additionally, sites with multiple years of data all showed non-trivial interannual variation. The 284 estimate of average interannual variability of CO_2 fluxes (25%) is modest compared to that 285 (88%) observed in terrestrial ecosystems (Baldocchi et al., 2018), probably reflecting the lower 286 number and diversity of ecosystems with multi-year measurements or more buffering against 287 climate extremes by large water bodies. However, given that CO₂ flux from freshwaters 288 positively scales with the productivity of terrestrial ecosystems at shorter time-scales (Butman et 289 al., 2016; Hastie et al., 2018; Walter et al., 2021), it is possible that the interannual variation of 290 carbon displaced from land will propagate onto CO₂ outgassed through freshwaters (Drake *et al.*, 2018; McDonald *et al.*, 2013), providing a possible pathway to constrain freshwater interannual
variability. Neglecting this variation is an additional source of bias in our current view on global
CO₂ emissions from lakes and reservoirs (Fig. 4).

294 Given that EC CO₂ fluxes are affected at both sides of the air-water interface (Wanninkhof et al., 295 2009), a better constraint of the contribution of lakes to the global carbon cycle will also require 296 reporting and synthesis of additional continuous waterside data (e.g. temperature, dissolved CO₂ 297 and O₂), site-level ecosystem characteristics (e.g. nutrient-color legacies, ecosystem metabolism, and aquatic vegetation such as algae) and sampling an increased site diversity within climatic 298 299 zones (Lehner and Döll, 2004). With more frequent air and aquatic observations, we will better 300 constrain CO_2 fluxes at different time scales, assess the prevalence of temporal patterns in CO_2 301 fluxes, and reduce uncertainty in eddy flux measurements over freshwaters (e.g., Ejarque et al., 302 2021). Such work will be needed to quantify and evaluate landscape (Buffam et al., 2011; Zwart 303 et al., 2018) to global (DelSontro et al., 2018) carbon budget components from lakes and 304 reservoirs.

305 5. Conclusions

306 Across 13 study sites with EC flux observations, on average all lakes and reservoirs were net 307 annual sources of CO₂ to the atmosphere. However, the time series unraveled large diel to (sub)-308 monthly oscillatory CO₂ patterns across sites, among a broad range of biogeochemical and 309 physical site characteristics. These modes of variability accounted for two thirds of daily and a 310 quarter of annual CO₂ flux variation, with sub-annual variability dominating over diel and inter-311 annual flux variabilities. After integrating these modes of variability into time-resolved fluxes, 312 the CO₂ flux estimates were at the upper end of published CO₂ emissions for lakes and 313 reservoirs. Our results support the idea that long-term, continuous, sub-weekly or sub-daily 314 measurements of carbon dynamics in freshwater aquatic systems are necessary to detect long-315 term trends of lake carbon fluxes. Long-term, frequent measurements of lake fluxes are also 316 needed to attribute natural and anthropogenic drivers to ecosystem changes that influence the 317 global carbon cycle and its future projections. We advocate for establishing and maintaining a 318 long-term observation network that combines EC flux measurements with highly detailed site319 specific carbon budget studies over key lake and reservoir ecosystems representing broader320 geographical gradients.

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341 Open Research

- 342 We have deposited all eddy covariance lake observations and gap-filled values in the
- 343 Environmental Data Initiative repository at: <u>https://environmentaldatainitiative.org/</u> and a DOI
- 344 will be provided once available, prior to acceptance. In the interim, reviewers can access all
- 345 original and gap-filled flux data are available at: <u>http://co2.aos.wisc.edu/data/lakeflux/synthesis/</u>

- 346 or this staging site: <u>https://portal-</u>
- 347 <u>s.edirepository.org/nis/mapbrowse?scope=edi&identifier=835&revision=1</u>. Several sites are
- also accessible from Fluxnet affiliated archives as noted in Table S2.

349 Author Contribution Statement

- 350 M.G. designed experimental protocol and conducted the data syntheses. A.R.D and M.G. wrote
- the manuscript. T.V., I.M., G.B., and G.W. supervised research, contributed observations, and
- 352 edited the manuscript. All other authors contributed with flux observations and commented on
- the manuscript.

354 Competing Financial Interests

355 The authors declare no competing financial interests.

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518 Figures

- 519 Fig. 1. Normalized histograms of daily CO₂ fluxes over ice-free season in nine lakes and four
- 520 reservoirs, showing that all studied ecosystems emitted CO₂ to atmosphere in the majority of site-
- 521 days. Vertical solid lines and their numerical representation indicate mean daily CO₂ flux.
- 522 Shaded areas show observations with negative CO₂ flux, which by convention, indicate net CO₂
- 523 uptake.



Fig. 2. a) Mean diel course of CO₂ fluxes across site-months (May-October) and site-years and
b) associated uncertainty (one standard deviation of monthly-averaged half-hourly observations
show consistent patterns of daytime flux reduction. Lines represent 6-hour moving average.
Negative fluxes indicate net CO₂ uptake. Note separate flux scale for the macrophyte reservoir.





- 531 Fig. 3. a) Seasonal evolution of daily CO₂ fluxes across site-years and b) associated uncertainty
- 532 (one standard deviation of daily values) indicate frequent sub-monthly oscillations across water
- 533 bodies. Negative fluxes indicate net CO₂ uptake. Note separate flux scale for two reservoirs.









544 Tables

- 545 **Table 1.** Comparison of ice-free CO₂ flux at temporal (i.e. annual, seasonal, diurnal and
- 546 nocturnal) scales derived from high-frequency eddy covariance measurements over lakes and
- 547 reservoirs. One standard deviation of the mean represents uncertainty of sub-annual CO₂ fluxes.
- 548 The numbers in brackets represent the number of observations integrated at a given time scale.

Lake ID	Name	Year	Air-water CO2 fluxes			
			Annual Totals	Seasonal daily mean	Daytime flux	Nighttime flux
			[gC m-2 yr-1]	[mgC m-2 d-1]	[mgC m-2 hr-1]	[mgC m-2 hr-1]
CA-Dar	Daring	2006	13.6 (n=153)	89±157 (n=95)	0.8±10.7 (n=1685)	12.2±7.5 (n=497)
	Lake					
	Fastersia	0000	104.4 (= 01.4)	501 (000 (m. 014)		00.0.00.4 (= 0117)
CA-ESI	Eastmain	2008	124.4 (n=214)	581±398 (n=214)	22.4±27.5 (n=2790)	26.2±23.4 (n=2117)
	Reservoir	2009	130.4 (n=214)	610±433 (n=214)	21.9±24.2 (n=2786)	30.1±25.1 (n=2127)
		2011	92.3 (n=214)	431±335 (n=214)	18±20.8 (n=2804)	17.9±19.5 (n=2108)
		2012	78.6 (n=214)	367±272 (n=173)	15.2±18.8 (n=2399)	15.7±15.7 (n=1568)
DE-Zrk	Zarnekow	2013	17.3 (n=214)	81± 880 (n=170)	-78.6±111.6 (n=2240)	103.1±47.5 (n=1678)
	Polder	2014	-53 6 (n=214)	-250+ 835 (n=214)	-86+104 1 (n=2817)	81 4+42 (n=2098)
	Reservoir	2015	84.7 (n-214)	396+1148 (n-214)	-41.2+101.2(n-2791)	84.2+54.6 (n-2139)
	110001101	2010	04.7 (11-214)	00011140 (11-214)	41.21101.2 (II=2701)	04.2104.0 (11-2100)
FI-Kui	Kuivajarvi	2010	137.6 (n=214)	643±140 (n= 58)	22.8±13.9 (n= 670)	30.2±13.8 (n= 656)
	Lake	2011	224.0 (n=214)	1047±304 (n=153)	39.7±17.1 (n=2075)	48.3±21.2 (n=1455)
		2012	164.7 (n=241)	684±274 (n=169)	24.4±16.5 (n=1981)	32.4±18.4 (n=1893)
FI-Pal	Pallasjärvi	2013	52.6 (n=173)	304±154 (n=93)	8.8±9.8 (n=1201)	17.2±9.9 (n=939)
	Lake					
EL VK a	Valkoa	2002	112.6 (n=200)	544+155(p-208)	22+7 (p=2285)	22.4 ± 9.9 (n=1949)
r i- v i la	Valkea-	2003	107.5 (n. 200)	544±155 (II=208)	22 ± 7 (II=2000)	23.4 ± 0.0 (II=1040)
	Kounen	2004	107.5 (1=239)	450±261 (n=238)	10.5±10.4 (II=2980)	21 ± 13.0 (1=2404)
	Lake	2005	87.1 (n=227)	384±215 (n=226)	11.4±15.3 (n=2940)	22.6±9.1 (n=2103)
		2006	119.9 (n=254)	472±263 (n=253)	15.8±13 (n=2983)	23.4±13.4 (n=2824)
		2007	119.6 (n=222)	539±232 (n=221)	20.8±11.3 (n=3033)	24.5±13.5 (n=2038)
FI-Van	Vänajavesi	2016	108.2 (n=237)	457±334 (n=237)	17.6±18.7 (n=2943)	20.8±17.8 (n=2505)
	Lake					
LA-NT2	NamTheun	2009	na	1762±186 (n=10)	61±17.8 (n=125)	87±39.2 (n=106)
	2	2010	na	1623±345 (n=15)	73.5±28.2 (n=146)	63.2±29.7 (n=200)
	Reservoir	2011	na	861±183 (n= 4)	36±16.3 (n= 47)	35.3±13.5 (n= 46)
			1			

SE-Mer	Merasjärvi	2005	23.9 (n=165)	145±149 (n=117)	4.7±9.4 (n=1877)	8.6±9.3 (n=835)
	Lake					
SE-	Tamnaren	2010	40.8 (n=216)	189±125 (n= 49)	6.9± 9.9 (n= 493)	8.6±11.4 (n= 628)
Tam	Lake	2011	35.9 (n=291)	124±161 (n=290)	4.9±11.2 (n=3619)	5.3±10 (n=3027)
		2012	82.7 (n=214)	386±176 (n=105)	10.8±12.3 (n=1663)	27.4±16.2 (n= 743)
US-	Douglas	2013	118.7 (n=275)	432±318 (n=102)	10.5±25.9 (n=1374)	28.5±39 (n= 965)
UM3	Lake	2014	113.4 (n=275)	412±313 (n=142)	9.6±24.9 (n=1889)	27.7±38.7 (n=1380)
US-	Ross	2007	59.0 (n=365)	162±308 (n=129)	5±23.5 (n=1324)	8.4±27.4 (n=1659)
RBa	Barnett					
	Reservoir					
US-Too	Toolik	2012	46.0 (n=153)	304±130 (n=62)	8±13 (n=1120)	28±27.5 (n=308)
	Lake					

560 Supplemental Material

561 S1. Supplemental Methods Text

562 Accounting for ice-cover

Current observations of CO_2 fluxes over lakes are mostly limited to an open water season due to wintertime measurement challenges and lack of consistent observations for all sites during ice covered season. These measurement challenges lead to a persistent under-sampling of icecovered seasons and periods around ice-on/ice-off. Transitions to/from open-water are often accompanied with large CO_2 efflux (Anderson *et al.*, 1999) and in some cases, comprise a significant proportion of annual CO_2 budget (Denfeld *et al.*, 2018). Therefore, the annual CO_2 flux estimates in Table 1 are conservative.

570 When the observational CO_2 data extended beyond the predicted ice-free season, the length of

571 ice-free days was adjusted to the first and/or last day of flux measurements. When the

572 observational fluxes were missing data near ice-on/ice-off dates, the seasonal mean daily CO_2

573 flux for a given lake was imputed to derive annual open water emissions. We assumed negligible

574 CO₂ transfer at the air-lake interface during the ice-covered season.

575 Footprint screening

576 Six out of 13 flux towers were placed on the lakeshore, shoals or islands (Table S1) to avoid

577 problems with power supply, wave and ice exposure, or because of the original research question

578 studied (e.g. CO₂ flux in heterogeneous landscape). This introduced an additional problem with

579 CO₂ advection from catchments and flux contamination. While well-selected tower locations

580 minimize the advection term to <3% of CO₂ flux (Morin *et al.*, 2018), the towers located in the

- 581 middle of the lake can also be affected by CO₂ advection, particularly small lakes surrounded by
- 582 forest (Esters *et al.*, 2021; Kenny *et al.*, 2017). The contribution of advected air to annual CO₂
- 583 lake budgets in this project is unknown and might be substantial. Tower height also influences
- 584 footprint area and likelihood of encountering secondary circulations (Kenney et al., 2017). To

account for these at some level, data from sites and time periods with suspected significant
contribution of mixed footprint were removed from this analysis. PI applied wind directional
screening is also applied to avoid land contributions and noted in Table S1.

588 Gap-filling of missing observations

589 Observations gap-filled were the climatic (i.e. air temperature, incoming solar radiation, 590 photosynthetically active radiation, horizontal wind speed, friction velocity, relative humidity, 591 barometric pressure, net radiation, vapor pressure deficit) and the lake fluxes (i.e. sensible, latent, 592 and CO_2 turbulent flux), and the in-water (i.e. surface water temperature, CO_2 concentration) 593 variables. However, there is no consistent method of flux gap-filling existing for freshwater 594 waterbodies. Here, we tested two approaches to gap-filling, the artificial neural network (ANN) 595 (Morin et al., 2014) and marginal distribution sampling (MDS) (Wutzler et al., 2018). The MDS 596 approach resulted in a smaller number of end-gaps, always used the same variables for gap-fill 597 and was computationally efficient relative to the ANN approach. Since a standardized gap-filling 598 protocol significantly reduces the uncertainty of compared NEE sums in multi-site syntheses 599 (Moffat et al., 2007), we therefore used the MDS approach for computing filled fluxes and 600 biophysical variables.

601 Uncertainty analysis

To reflect uncertainty, we calculated the standard error of the mean (i.e. square root of summed

603 variances normalized by square root of number of observations, SEM) for daytime and nighttime

half-hourly averages (Table S1). SEM for daytime observations varied from 0.196 μ mol m⁻² s⁻¹

605 in FI-VKa to 1.82 μ mol m⁻² s⁻¹ in DE-Zrk, whereas SEM for nighttime observations ranged

- from 0.200 μ mol m⁻² s⁻¹ in FI-VKa to 1.38 μ mol m⁻² s⁻¹ in US-UM3. The average nighttime CO₂
- 607 uncertainty was higher than daytime uncertainty in seven lakes.
- 608 The open-path (OP) gas analyzer measurements were on average one third more uncertain than
- the closed-path (CP) measurements (Table S1). The daytime SEM in OP ranged from 0.228
- 610 μ mol m⁻² s⁻¹ to 0.932 μ mol m⁻² s⁻¹ (mean: 0.565 μ mol m⁻² s⁻¹), while the daytime SEM in CP
- 611 ranged from 0.196 μ mol m⁻² s⁻¹ to 0.558 μ mol m⁻² s⁻¹ (mean: 0.382 μ mol m⁻² s⁻¹). The CO₂ flux
- 612 uncertainty in DE-Zrk measured with CP was higher (mean: 1.76 µmol m⁻² s⁻¹) compared to

613 uncertainties in lakes because a large proportion of emergent macrophytes within the flux tower614 footprint contributed to much stronger signal and flux magnitudes comparable to wetlands.

615 CO₂ fluxes over freshwater systems are small relative to fluxes to terrestrial systems, show low

616 signal-to-noise ratio, and require the Webb-Pearson-Leuning (Webb *et al.*, 1980) and Burba

617 corrections (Burba *et al.*, 2008) for covarying fluctuations of water vapor flux and temperature.

618 The corrections terms, especially for OP measurements, can be larger than measured CO₂

619 quantities, leading to biased CO₂ flux especially when carbon flux is small and corresponding

heat flux is large (Helbig *et al.*, 2016) and result in physiologically unreasonable net CO₂ flux,

621 such as nighttime uptake in eutrophic water bodies (Lee *et al.*, 2014; Potes *et al*, 2017). The sites

622 consistently showing such a nighttime uptake were excluded from this meta-analysis.

623 Intercomparison with other methods

624 We assume with sufficient sampling period, the continuous EC flux measurements are

625 representative for ecosystems with similar biotic and abiotic conditions. The inter-comparison

626 with other methods of estimating CO₂ flux from lakes (i.e. floating chambers, surface renewal

627 model, and boundary layer models) showed varying degrees of agreement. Relative to CO₂ flux

628 estimates, the simultaneous measurements with other methods typically agreed within 20%,

though periods with large departures up to 2-3 times larger or smaller do occur (e.g., Anderson et

630 *al.*, 1999; Baldocchi *et al.*, 2020; Eugster *et al.*, 2003; Erkkilä *et al.*, 2018; Jonsson *et al.*, 2008;

631 Podgrasjek et al., 2014; Vesala et al., 2006). The agreement varied on level of stratification,

632 overlap in timing of measurements, season, and the selection of piston velocity models. There is

633 good reason to believe that flux tower approaches can be a viable method for estimating

634 lake/reservoir CO₂ fluxes, though studies that found greater discrepancies among independent

635 methods and models require reconciliation.

636 Interannual variability calculation

637 One standard deviation of annual CO₂ fluxes was calculated to determine the inter-annual

638 variation (IAV) of fluxes for sites with multi-year measurements. We acknowledge that

639 calculating the standard deviation from a limited number of site-years (i.e. <5 years) can lead to

640 uncertain estimates of IAV, however, it cannot be further constrained with this study dataset.

With more multi-year time series of continuous measurements, we will be able to determine the
5-year time threshold is sufficient to capture inter-annual CO₂ flux variability in freshwater
ecosystems.

644 Regression analysis

We tested several models (e.g. quadratic, linear, exponential, gaussian, etc) available in the library of models in the Curve Fitting Toolbox in MATLAB with the variable number of models' parameters to select the most robust model. We selected the robust linear least-squares secondorder polynomial model with bisquare weighting method. This statistical model maximized the goodness-of-fit (e.g. r2 and rmse), required less parameters to estimate and dealt with nonlinearities. To avoid influences of outliers on fitted curves, values beyond 1st and 99th percentile of each variable were removed before curve fitting.

652 Fluxes over a macrophyte reservoir (DE-Zrk)

A significant fraction of emergent macrophytes within a flux tower footprint of a shallow reservoir DE-Zrk increased the flux temporal dynamics by an order of magnitude relative to fluxes measured over open water lake surfaces (Table S2). Since the emergent macrophyte stands are common in shallow lakes and reservoirs, the unique CO_2 flux over such systems is worth describing separately.

The mean and standard deviation of daily CO₂ were 0.072 ± 0.970 µmol m⁻² s⁻¹ (range: -0.858-

 1.352μ mol m⁻² s⁻¹, Fig. 1, Table 1). Daytime to nighttime hourly fluxes were on average 250%

660 lower, indicating a strong mid-day photosynthetic CO₂ fixation of macrophytes, roughly seven

- times higher than daytime CO_2 drawdown observed in open-water systems (Fig. 2). The negative
- 662 correlation with PAR additionally confirmed a strong control of macrophyte photosynthetic
- 663 activity over sub-daily CO₂ flux variation (Fig. 3). The maximum uptake of the monthly-
- averaged daily flux amplitudes typically occurred in July ranging 6.15-8.31 μ M m⁻² s⁻¹ and
- declined towards both ends of the ice-free season. At annual timescale, CO₂ fluxes indicated a
- 666 net source of C in two lake-years (Table 1). The mean interannual variability (IAV) of annual
- 667 CO₂ flux 428%. Overall, the mean and median uncertainty of daily CO₂ flux were 162% and
- 668 421% for DE-Zrk (Fig. 3b). The flux values were several-fold underestimated relative to the

- 669 published CO₂ flux estimates for this site (Franz *et al.*, 2016) The CO₂ fluxes contributing to the
- 670 tower footprint of DE-Zrk were distinct between the open water and the emergent vegetation
- 671 categories. The lack of footprint heterogeneity was not considered in the uniform method of flux
- 672 computation applied in this study but was the most significant source of discrepancies in
- 673 estimated fluxes between these two studies.

674 S2. Supplemental references

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868 S3. Supplemental Tables

- 869 **Supplemental Table S1**. Additional site characteristics of the eddy flux tower sites synthesized
- 870 in this study.
- 871 Supplemental Table S2. Summary statistics of the literature-compiled mean daily, annual and
- 872 inter-annual CO₂ fluxes derived from at least four samples per year. Within-study flux variation
- 873 is expressed as one standard deviation of the (published or calculated) mean. Numbers in
- 874 brackets indicate the minimum and maximum flux values at given time scale

Site Descriptive Variable	CA-Dar	CA-Est	DE-Zrk	Fl-Kui	FI-Pal	FI-Van	FI-Vka	LA-NT2	SE-Mer	SE-Tam	US-UM3	US-RBa	JS-Tol
Site Name	Daring	Eastmain	Zarnekow Polder	Kuivajarvi	Pallasjärvi	Vanajavesi	Valkea-Kotinen	Nam Theun 2	Merasjärvi	Tamnaren	Douglas	Ross Barnett	Foolik
Type (1)	_	ж	R	_	L	-	<u> </u>	К	L	L	Ţ	R	
Latitude, Longitude [dec dea]	64.86, -112.59	52.13, -75.9	53.88, 12.89	61.83, 24.28	68.01, 24.21	61.13, 24.26	61.24, 25.06	17.69, 105.3	67.55, 21.96	60.16, 17.32	45.57, -84.7	32.44, -90.0	38.63, -150.0
Climate (2)	Dfc	Dfc	Cfb	Dfb	Dfc	Dfb	Dfc	Am	Dfc	Cfb	Dfb	Csa	Ofc
Mean Annual Air Temperature	-9.7	-1.6	9.8	5.1	-1.7	4.1	4.1	23.2	-0.6	6.1	6.3	18.1	-8.8
Mean Annual Precipitation [mm yr-1]	246	738	3 557	550	444	534	533	2109	418	577	727	1385	135
Surface Area [km2]	14.8	623	3 7.5	0.63	17.3	103	0.036	489	3.8	38	15.2	134	1.5
Mean, Maximum Depth [m]	Na, Na	11, 15	5 0.6, 1.2	6.4, 13	9, 36	7, 25	25, 6.5	7.8, na	5.1, 17	1.3, 2	5.5, 14	3.6, 10.7	7, 25
Volume [km3]	Na	6.94	1 0.05	0.004	0.155	0.721	0.0000	3.9	0.019	0.42	0.083	0.482	0.011
Catchment to lake area ratio [km2]	Na	Na	Na	œ	6.08	27	8.1	5.8	17.11	18.9	3.72	58.96	43.33
Impoundment Year [yr]	NA	2006	3 2004/2005	NA	NA	NA	NA	2008	NA	NA	NA	1963	٩٨
TP [µg L-1]	Na	18.1	235	15	5.5	28	16	45	6.6	79.92	6	06	2
TN [µg L-1]	Na	233	8 Na	400	132	Na	465	917	113	1336	454	1500	234.9
Chlorophyll a [µg L-1]	Na	29	9 Na	3.75	2.11	19	15.8	8.5	1.9	Na	0.4	10	1.4
pH [pH units]	6.75	9	8.2	6.5	7.06	Na	5.2	6.67	2	8.97	7.98	8	7.1
DOC [mg L-1]	2.7	6.5	22	12.9	2.72	9.6	11	2.66	6.2	23.16	99-01	i Na	6.9
Macrophytes within footprint (3)	Na	Na	Cd, Cg, Lm, Pa, Sp, Tl	Na	Na	Na	Na	Na	Na	Bs, Es, Ms, Pau, Sa	Na	Na	Va
Trophic Status (4)	от	MT	ET	МТ	от	МТ	MT	ET	oT	ET	от	ET	DT
Color Status (5)	н	НМ	Na	НМ	но	НМ	НМ	но	НМ	На	НМ	Na	Ŧ
Depth Classification (6)	ο	Q	S	۵	Q	٩	s	Q	Q	S	Q	۵	0
Mixis Type Classification (7)	Na	Na	Pm	Dm	Dm	FMm; ESm	FMm; EDm; SSm_V	WMm	Dm	Pm	Dm; EMm	mMW	m
Measurement Height [m]	4.1	15	2.63	1.7	2.5	2.75	1.5	e	1.7-2.6(±.1)	4.3	1.5	4	Va
Sensor Type (8)	PO	ОР	СР	СР	СР	сР	CP	Р	Ð	Р	ОР	Ð	CP (?)
Fetch Screen ing	excl. 10-270	0-40; 220-360	1-360 (mixed footprint)	135-170; 290-345	180-330; 350-50	excl. 100-185	100-170; 290-350	1-360	excl. 155-210	excl. 30-90	excl. 180-270	1-360	1-360
Tower Location	island	island	platform	platform	shoal near shoreline	on shore	platform	platform	platform	island	shoal near shoreline	platform	olatform
Source	Humphreys, unpubl.; Google Earth	Teodoru et al., 2011; Vachon et al., 2013 , Wang et al., 2018	Zak et al., 2008; Franz et al., 2016	Dinsmore et al., 2013; Heiskannen et al., 2015; Mammarella et al., 2015; Miettinen et al., 2015	Lchilla et al., 2015	Provenzale, unpubl.	Vesala et al., 2005; I Arvola et al., 2014 (I	Deshmukh et al., 2014; Chanudet et al., 2016; Martinet et al., 2016	Jonsson et al., 2008; Aberg et al., 2010	Podgrajsek et al., 2014 http://miljodata.slu.se	Kwon et al., 2014; Kemry et al., 2017; Morin et al., 2018 <u>dx.doi.org/10.17190/AMF</u> [1480315	Sobolev et al., 2009, Liu et al., 2016	Eugster et al., 2003; uecke et al., 2014 Tan82;

Supplemental Table S1. Additional site characteristics of the eddy flux tower sites synthesized in this study.

1 Waterbody type: L-lake; R-reservoir

2 Climatic zone based on Koeppen-Geiger climate classification map (5 arc min) following Kotek et al., 2006 re-analyzed by Rude letal., 2017: DE - snow climate, fully humid, cool summer, cod winter; DE - snow climate, fully humid, warm summer; Am - equatorial morecon;

Cfb - warm temperate climate, fully humid, warm summer; Cfa - warm temperate climate, fully humid, hot summer;

3 Submergedficating (SF) and emergent (E) macrophydiers C4-Ceratophydium demensum (SF), Es - Elodeas sp. (SF), Un - Lemma minor (SF), Ma - Myriophyllum spicatum (SF), Pa - Pdygorum an philbium (SF), Sp - Spirobala polyhhizat (SF), Bs - Budomus spp (SF), Cg - Carex gracitilis (E), Pau – Phragmites australis (E), Sa - Schoenoplectus acutus (E), TI - Typha latifolia (E);

4 Traphic status classification: 0T - Oligotrophic (TP-cl0µgL-1); MT - Mesotrophic (10µgL-1): ET - Eutrophic (TP-30µgL-1) 5 Humic status classification: OH - Oligotrumic (DOC<7 mgL-1); MH – Mesotrumic (7<DOC<11 mgL-1); PH – Pdytrumic (DOC>11mgL-1)

6 Depth class filoation: S- Shallow (mean depth <3m); 7 Mixing type class filoation: Dm - dmictic; FMm - fall monomictic; FMm - warm monomictic; ESm - episod cally fall monomictic; EMm - episod cally fall monomictic or spring monomictic 8 Sensor type: OP - open path gas analyzer; CP - closed path gas analyzer.

Supplemental Table S2. Summary statistics of the literature-compiled mean daily, annual and inter-annual CO₂ fluxes derived from at least four samples per year. Within-study flux variation is expressed as one standard deviation of the (published or calculated) mean. Numbers in brackets indicate the minimum and maximum flux values at given time scale

Source	Daily CO2 flux [mgC m	Annual CO2 flux [gC m	Inter-annual CO2 flux	Measurement Freque	Notes
Site-level discrete CO2 flux					
Brothers et al., 2012	1343±72	na	na	22 year-1	Estmain R., Ice-free season; Jun-Sep sampling; Boreal
Casper et al., 2000	480	na	na	13 year-1	Priest Pot; Ice-free season; May-Oct sampling
Chmiel et al., 2013	406	94	na	Two-year 6 years-1	L. Gäddtjärn; lce-free season; Boreal
Cole and Caraco, 1998	109±29 (80-138)	40±11 (30-50)	na	Multi-year weekly	L. Mirror, Two methods of flux estimation;
Demarty et al., 2011	na	81±52 (21-137)	4 (Eastmain R.)	Space-resolved two-ye	One reservoir and two lakes; Ice and ice-free sampling; Boreal
Denfeld et al., 2018	na	30±37 (-11-152)	11 (4-23)	Multi-year 7-9 year-1	10 lakes; Ice-free season; Temperate
Einola et al., 2011	na	50±232 (18-86)	102 (4-19)	1-4 year-1 (5 lakes)2; 1	5 lakes; Ice-free season; Boreal
Eugster et al., 2003	191-102 (114-365)	na	na	8-602 year-1	L. Toolik; Convective and stratified periods; Four methods of flux estimation; Arctic
Finlay et al, 2019	341	61±24	7	Multi-year (37) weekly	Bufallo Pound Lake; Ice-free season; Great Plains;
Jonsson et al., 2008	180±45 (119-225)	na	na	6 summer-1 (2 method	L. Merasjarvi; Three methods of flux estimation; Summer; Boreal
Karlsson et al., 2013	na	9±14	na	7 year-1	Twelve lakes; May-Oct sampling
Kling et al., 1991	258±100 (150-420)	na	na	Multi-year mean 44-62	Two lakes and multi-lake (25) average; Arctic;
Kokic et al., 2015	500±100 (400-600)	na	na	5 year-1	L. Gäddtjärn and headwater lakes, Ice-free season; Jun-Nov sample; Boreal
Miettinen et al., 2015	542±13 (529-555)	84±5 (79-89)	5	14 year-1	Kuivajarvi; Ice-free season; Boreal;
Natchimuthu et al., 2017	549±193 (306-780)	1433	na	Space-resolved 89-129	Three lakes; Ice-free season; Jun-Oct sampling; Hemiboreal
Ojala et al., 2011	261±109 (151-401)	59±16 (41-82)	na	weekly	Two lakes, Two methods of flux estimation; Ice-free season; Boreal
Repo et al., 2007	328±136 (136-437)	na	na	6-7 year-1	Three lakes; Jul-Sep sampling; Subarctic
Riera et al., 1999	245±223 (5-549)	54-49 (1-120)	na	8-32 year-1	Four lakes; Ice-free season; Apr-Nov sampling; Temperate
Sobek et al. 2003	na	10±18	na	4 year-1	29 lakes; Ice-free season; Boreal
Stets et al., 2009	na	26±24 (2-49)	na	daily	Two lakes; Year-round integration; Mass-balance model; Temperate;
Striegl & Michmerhuizen 1998	na	49±47 (1-96)	na	21-25 year-1	Two lakes; Ice and ice-free season; Mar-Aug sampling; Temperate
Site-level, high-frequency, seasonal CO2 flux					
Franz et al., 201529	516	174	na	Year-round continuous	Zarnekow macrophyte reservoir; Footprint heterogeneity incorporated; Temperate; Eddy covariance
Huotari et al., 2009	na	37±7 (30-44)	7	Two-years continuous	L. Valkea-Kotinen; Ice-free season; Boreal
Huotari et al., 2011	na	77±10 (68-97)	10	Five-year continuous	L. Valkea-Kotinen; Ice-free season; Boreal, Eddy covariance
Lundin et al., 2015	na	24±16 (5-54)	na	Continuous	Six lakes; Ice-free season; Subarctic
Morales-Pinnieda et al, 2014	311±202 (96-601)	na	na	Continuous	Two reservoirs; Ice-free annually; May-Oct sampling; Two methods of flux estimation; Mediterranean
Pelletier et al., 2014	450±300	na	na	Two-year continuous	Pool in peatland; Ice-free with a brief under-ice period; May-Oct sampling; Boreal
Vachon et al., 2017	311±39 (272-350)	49±6 (43-56)	na	Continuous	Two lakes; Ice and ice-free season; Boreal
This study	450±354 (-78-1298)	95±49 (14-224)	22 (4-44)	Continuous	Nine lakes and three reservoirs, Ice-free season; Six climatic zones; Eddy covariance
	75±1006 (-890-1402)	16±151 (-39-632)	151	Continuous	Macrophyte reservoir DE-Zar, Ice-free season; Temperate; Eddy covariance
Regional upscaling (Process- or mass balance models) CO2 flux					
Cardille et al, 2009	na	119±18 (99-142)	na	daily time steps	Ice-free season; US Mid-west region; Different precipitation scenarios
Zwart et al., 2018	na	47±5 (41-53)	na	daily time steps	Ice-free season; US Mid-west region
	na	27±3 (24-31)	na	daily time steps	Year-round; US Mid-west region
McDonald et al., 2013	263±180 (-50-610)	96±66 (-18-223)	na	daily	Year-round; Continental US
Regional / global upscaling (Extrapolation-based m	odels or space-resolve	d average) CO2 flux			
Alin and Johnson, 2007	na	52±26 (14-91)	na	variable	Large lakes, Latitudinal gradient, Global
Buffam et al., 2011	na	32 (25-39)	na	1 summer-1 (168 lakes	Ice-free season, Summer sampling; US Mid-west region;
Bogard & delGiorgio., 2016	232 (-187-979)	na	na	1 summer-1 (346 lakes	Boreal region
Deemer et al., 2016	451±86 (330-525)	na	na	variable	Reservoirs, Global summary statistics
Del Sontro et al., 2018	374±114 (242-563)	na	na	variable	Size area bins, Global summary statistics
Hastie et al., 2018	na	139 (54-257)	na	variable	Size area bins, Boreal region
Holgersson and Raymond, 2016	269±77 (138-423)	na	na	variable	Size area bins, Global summary statistics
Kankaala et al., 2013	na	77±43 (41-159)	na	weekly (17 lakes)	17 lakes; Ice-free season; Size area bins, Boreal region
Kortelainen et al., 2006	na	65±24 (37-102)	na	4 year-1	177 random lakes; Ice-free season; Size area bins; Boreal region
Rantakari & Kortelainen, 2005	138±12 (114-149)	24±32 (18-28)	32	3 year-1	37 lakes; Ice-free season; Boreal region
Raymond et al., 2013	na	148±146 (-1-537)	na	variable	Back-calculated from flux yield; Year-round integration; Northern Hemisphere excl. tropical region;
Weyhenmeyer et al., 2015	708 (128-2,620) 448	na	na	1 autumn-1 (5,118 lak	Ice-free season, Two methods of flux estimation; Boreal and hemiboreal region,

875 S4. Supplemental Figures

876 Supplemental Fig. S1. Seasonal patterns of CO₂ flux in three example lakes: a) FI-Van - a

877 boreal lake, mesotrophic, mesohumic, deep, fall monomicitc with episodic summer mixing due

878 to weak stratification, b) FI-VKa - a boreal lake, mesotrophic, mesohumic, shallow, fall

- 879 monomictic, strong summer stratification, and c) DE-Zrk a temperate eutrophic reservoir,
- shallow, cold polymictic with emergent macrophytes.

