

LETTER

Idiosyncratic phenology of greenhouse gas emissions in a Mediterranean reservoir

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Scientific Significance Statement

Reservoirs are important sources of greenhouse gases (GHGs) impacting the global carbon cycle. Global change is increasing surface water temperatures, prolonging stratification, and enhancing eutrophication with uncertain consequences for future GHG emissions. Reducing the uncertainty of current predictive models requires incorporating the phenological variability in GHG emissions across latitude. This need is particularly pertinent in the Mediterranean zone, where extreme seasonal fluctuations characterize the hydrological and thermal regimes. In the Mediterranean reservoir studied, the GHG emissions were more variable than in reservoirs of other latitudes and mainly driven by water temperature, mean depth, and biological productivity. However, the relationship between CO₂ emissions and temperature was more complex. Our results suggest that future scenarios with higher temperatures, drought, and biological productivity will likely increase CH₄ emissions.

Abstract

Extreme hydrological and thermal regimes characterize the Mediterranean zone and can influence the phenology of greenhouse gas (GHG) emissions in reservoirs. Our study examined the seasonal changes in GHG emissions of a shallow, eutrophic, hardwater reservoir in Spain. We observed distinctive seasonal patterns for each gas. CH₄ emissions substantially increased during stratification, influenced predominantly by the increase in water temperature, net ecosystem production, and the decline in reservoir mean depth. N₂O emissions mirrored CH₄'s seasonal trend, significantly correlating to water temperature, wind speed, and gross primary production. Conversely, CO₂ emissions decreased during stratification and displayed a quadratic, rather than a linear relationship with water

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Author Contribution Statement: IR, RM-B, and FJR conceived the study and obtained the funds. ER-V led the fieldwork and the measurements of greenhouse gas emissions with the help of IP-M, AM-G, MG-A, FP, RJG, CLR, RM-B, FJR, and IR. ER-V, IP-M, and CLR performed the data analysis and graphical representation of the results. ER-V and IR wrote the first draft of the manuscript, and all authors contributed significantly to the preparation of the final draft.

Data Availability Statement: Additional Supporting Information can be found in the online version of this article, including an extended version of methods and supplementary tables. The dataset associated with this paper is available from dryad (<https://doi.org/10.5061/dryad.cnp5hqcbz>).

Additional Supporting Information may be found in the online version of this article.

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temperature—an unexpected deviation from CH₄ and N₂O emission patterns—likely associated with photosynthetic uptake of bicarbonate and formation of intracellular calcite that might be exported to sediments. This investigation highlights the imperative of integrating these idiosyncratic patterns into GHG emissions models, enhancing their predictive power.

Inland waters, including reservoirs, are important sources of greenhouse gases (GHGs) such as CO₂, CH₄, and N₂O (Tranvik et al. 2009; Bastviken et al. 2011; Raymond et al. 2013; Soued et al. 2016; León-Palmero et al. 2020a). The global estimations of GHG emissions from reservoirs range from 1.25 to 2.3 Pg CO₂ equivalents per year (Lauerwald et al. 2023). However, these global estimations are still substantially uncertain because of the idiosyncrasy of each gas—with specific external forcing and internal drivers—and the limited availability of temporal data across latitudes. Although there have been some recent attempts to describe the temporal variability of CO₂ emissions using eddy covariance techniques (Golub et al. 2023) and to model CH₄ temporal variability (Johnson et al. 2021; Zhuang et al. 2023), most available data were obtained under fair weather conditions and thus do not include the full range of variability of natural systems (Liu et al. 2016; Ran et al. 2021). Phenology (i.e., seasonal changes) differs from tropical (high and uniform temperature) to boreal (low temperature) latitudes. Therefore, input data in models should incorporate the specific, phenological drivers of GHG emissions across latitudes to reduce uncertainty. This need is especially pertinent in the warm temperate dry climatic zone (including Mediterranean zone), where extreme seasonal fluctuations characterize the hydrological and thermal regimes, and reservoirs are more prevalent than lakes (Lehner and Döll 2004; Lehner et al. 2011).

Previous studies of CO₂ emissions have not shown consistent phenological patterns yet. These emissions can be higher (Yang et al. 2013; Zhao et al. 2013; Podgrajsek et al. 2016) or lower (Shao et al. 2015; Pu et al. 2020; Amanatidou et al. 2023) during the summer than the fall–winter regardless of the reservoir latitude. These incongruent patterns are likely related to local drivers of CO₂ emissions. For instance, Yang et al. (2013) and Golub et al. (2023) found a positive relationship between temperature and CO₂ fluxes in subtropical and temperate reservoirs. It is well known that high water temperatures reduce gas solubility, promoting CO₂ release from surface waters. Moreover, seasonal changes in primary production can also modify the CO₂ exchange with the atmosphere. CO₂ assimilation by photosynthetic microorganisms decreases the dissolved CO₂ in the water and, likely, reduces emissions Pacheco et al. (2014). Zhao et al. (2013) and Pu et al. (2020) also found significant negative correlations between CO₂ fluxes and chlorophyll *a* (Chl *a*). However, this influence appears to be evident only in low-carbonate reservoirs (León-Palmero et al. 2020a). Temperature also affects the balance between primary production and microbial

respiration. All these studies suggest that external forcing (climate) and internal drivers (carbonate–bicarbonate availability, biological productivity) interact to control CO₂ emissions, making scaling relationships challenging.

In the case of CH₄, the maximum emissions are consistently reported during the warm or shoulder seasons (Samiotis et al. 2018; Linkhorst et al. 2020; León-Palmero et al. 2020a; Johnson et al. 2021). The release of CH₄ by direct ebullition from the sediments generally shows this regular seasonal pattern that is related to higher hypolimnetic temperature and lower water level that facilitates ebullition (DeSontro et al. 2010; Linkhorst et al. 2020; León-Palmero et al. 2020a; Waldo et al. 2021b). Other studies correlate Chl *a* concentration with CH₄ emissions (Zhao et al. 2013; Harrison et al. 2017; Beaulieu et al. 2019; Deemer and Holgerson 2021). Particulate organic matter derived from phytoplankton appears to be an essential substrate for methanogenesis (West et al. 2015; Martínez-García et al. 2024). Therefore, climatic forcing, water level and Chl *a* affect CH₄ emissions under a more predictable pattern across latitudes (Johnson et al. 2021).

N₂O emissions also show a consistent temporal pattern with the highest emissions during summer, at least in subtropical reservoirs (Zhu et al. 2013; Musenze et al. 2014; Yang et al. 2023). However, this pattern can be modified by other factors such as dissolved oxygen concentrations, since high dissolved oxygen concentrations can inhibit denitrification and stimulate nitrification (Zhao et al. 2013; Yang et al. 2023). Furthermore, N₂O emissions may also be influenced by the availability of nutrients and organic matter. For instance, León-Palmero et al. (2020a, 2023) found that nitrogen and phosphorus inputs in the reservoirs increase N₂O concentration and emission because they can affect denitrifying bacteria activity (Zhu et al. 2013).

In general, water temperature is a common driver for all GHG emissions. However, the intensity of this factor and the concurrence of other drivers can modulate the net GHG emissions, making such emissions very idiosyncratic. Disentangling these drivers is relevant since recent studies point towards more extended stratification periods (Woolway and Merchant 2019) with widespread deoxygenation in the hypolimnion (Jane et al. 2021), more frequent heatwaves (Woolway et al. 2021), and an increase of eutrophication (Beaulieu et al. 2019).

Here, we determined the seasonal changes of CO₂, CH₄, and N₂O emissions in a Mediterranean reservoir. We hypothesize that the GHG emissions would be higher in summer than in winter due to increasing temperatures. CH₄ ebullition and N₂O emissions will be accentuated in the summertime due

also to reduction of water level and development of anoxic conditions in the hypolimnion.

Materials and methods

Cubillas is a shallow, hardwater, eutrophic reservoir located in the southeast of Spain (37°16'34"N, 3°40'24"W) with a surface of 194.4 ha and a capacity of 13.53 hm³. The reservoir has uniform bathymetry (Supporting Information Fig. S1). The annual water level fluctuated approximately 3 m, with the lowest values during fall. The main reservoir uses are irrigation and recreation.

To describe the GHGs phenology in this reservoir, we monitored the fluxes of CO₂, CH₄ (diffusive and ebullitive), and N₂O weekly from March 2021 to July 2022. We determined CO₂, CH₄, and N₂O fluxes using a Cavity Ring-Down Spectrometer PICARRO G2508 connected to a floating chamber placed on the water surface. On each sampling day, we took 4–6 measurements throughout the daytime at the same location (Supporting Information Fig. S1). We calculated fluxes as in Zhao et al. (2015) (Supporting Information). For the CH₄ fluxes, we also discriminate between diffusion and ebullition using an adaptation of the algorithm proposed by Hoffmann et al. (2017) (Supporting Information). The CO₂ equivalent emissions were calculated by multiplying the mass-based flux (in units of mg CO₂, CH₄, or N₂O m⁻² d⁻¹) by the 100-yr global warming potential of each gas (1 for CO₂, 27 for CH₄, and 273 for N₂O; IPCC 2022).

We recorded the water temperature, oxygen concentration, and water depth using a SeaBird 19 plus profiler. We determined the strength of the thermal stratification (*see* Supporting Information Fig. S2 and the related text). To obtain the wind speed data, we used a Davis® Wind Speed and Direction Smart Sensor and a Campbell Scientific WindSonic2. We determined the nitrate concentrations using ion chromatography. We calculated gross primary production (GPP), respiration (Res), and net ecosystem production (NEP) using the diel oxygen method proposed by Staehr et al. (2010) and data recorded using a PME miniDOT probe (10 min resolution) and a TriOS Dissolved Oxygen Sensor (1 min resolution) located at 1 m depth (Supporting Information).

We explored the potential drivers for each gas using linear mixed-effects models (LMM) including sampling date as a random factor. We selected several variables as potential fixed predictors, including water temperature, wind speed, water level, GPP, NEP, and nitrate concentration. The two-level factor “period” (stratification vs. mixing) was excluded in all models. This factor was strongly correlated with temperature (Supporting Information).

Results and discussion

The study reservoir always acted as a source of CO₂, CH₄, and N₂O (i.e., all fluxes were positive). The CO₂ emissions

ranged from 12.5 to 567.02 mg C m⁻² d⁻¹ (Fig. 1a). We did not find significant differences in the CO₂ fluxes between the stratification and mixing periods ($t = 0.6$, $df = 62$, p -value = 0.55) (Fig. 1b). These values are similar to other temperate reservoirs (Barros et al. 2011; Morales-Pineda et al. 2014; Deemer et al. 2016; León-Palmero et al. 2020a). The highest emissions were observed during spring and fall and immediately after the disruption of Saharan dust deposition (*see* black arrows in Fig. 1a). This reservoir experiences recurrent Saharan dust intrusions that transport phosphorus and organic matter boosting bacterial and primary productivity (Morales-Baquero et al. 2006; Reche et al. 2009; De Vicente et al. 2012). Therefore, these peaks in CO₂ emissions associated with Saharan dust deposition could be related to an enhancement of bacterial metabolism with respect to primary production. The lowest emissions were observed during the summer. This seasonal pattern has been also observed in other Mediterranean reservoirs which even found negative fluxes (i.e., CO₂ uptake) during summertime (Samiotis et al. 2018; Montes-Pérez et al. 2022) and in other subtropical (Yang et al. 2013; Pu et al. 2020) and boreal (Demarty and Tremblay 2019) reservoirs. However, this pattern is less consistent in other Northern-temperate (> 50°N) reservoirs with higher emissions during summertime (Golub et al. 2023).

The total CH₄ emissions ranged from 0.23 to 1204.82 mg C m⁻² d⁻¹ (Fig. 1c, green dots), being significantly higher ($t = 7.9$, $df = 61$, p -value < 0.001) during the stratification than during the mixing period (Fig. 1d). These values are higher than those found in most tropical and Northern-temperate reservoirs (Barros et al. 2011; Deemer et al. 2016) and represent an upper limit in the emissions reported for the Mediterranean reservoirs (Samiotis et al. 2018; León-Palmero et al. 2020a; Montes-Pérez et al. 2022). The diffusive fluxes ranged from 0.23 to 472.09 mg C m⁻² d⁻¹ (Fig. 1c, yellow dots), whereas the ebullitive fluxes ranged from 0 to 1200.89 mg C m⁻² d⁻¹. The ebullition contribution to the total CH₄ emissions was lower than 20% during the mixing period, whereas it was usually higher than 70% during the stratification period accounting for up to 99.8% of the total fluxes for some of the summer sampling dates. These results are similar to those obtained in shallow reservoirs in Northern-temperate, Mediterranean, and tropical latitudes (Keller and Stallard 1994; DelSontro et al. 2010, 2011; Miller et al. 2017; Montes-Pérez et al. 2022) but higher than the emissions measured in deep reservoirs (DelSontro et al. 2011).

The N₂O fluxes ranged from 0 to 670.92 μg N m⁻² d⁻¹ (Fig. 1e). We observed higher emissions during August, but without significant differences between the stratification and the mixing period ($t = 0.7$, $df = 64$, p -value = 0.49) (Fig. 1f). These values are similar to those found in boreal, Northern-temperate, Mediterranean, and subtropical reservoirs (Liu et al. 2011; Musenze et al. 2014; Soued et al. 2016; Descloux et al. 2017; Liang et al. 2019; León-Palmero et al. 2020a).

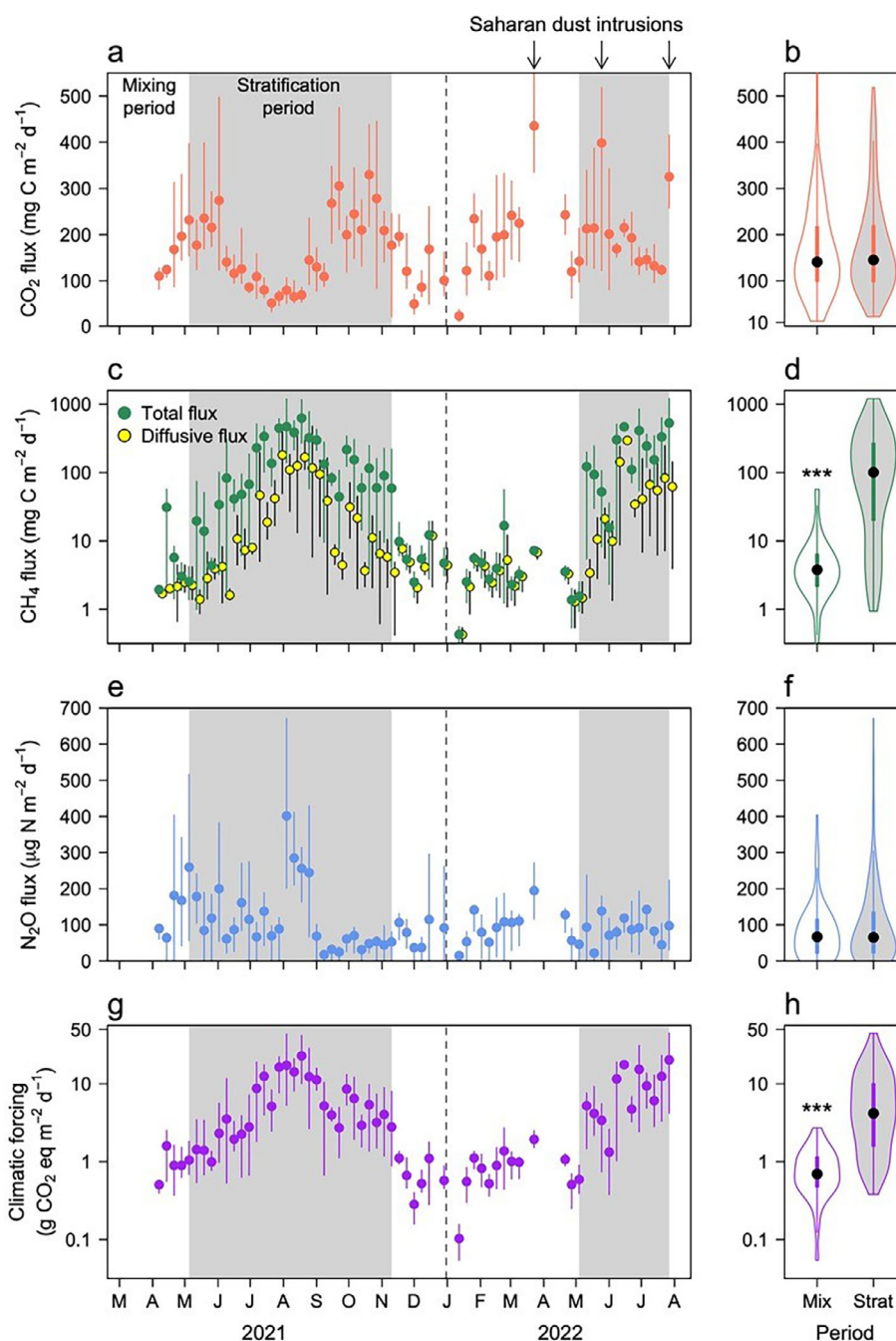


Fig. 1. Seasonal changes in the CO₂ emissions (a) and the violin plots for the distribution of values during the mixing and stratification periods (b), seasonal changes in the CH₄ total (green dots) and diffusive (yellow dots) emissions (c) and the violin plots for the distribution of values during the mixing and stratification periods of total CH₄ emissions (d), seasonal changes in N₂O emissions (e) and the violin plots for the distribution of values during the mixing and stratification periods (f), and the seasonal changes in the climatic forcing due to CO₂, CH₄ and N₂O emissions expressed in CO₂ equivalents in the Cubillas reservoir (g) and the violin plots for the distribution of values during the mixing and stratification periods (h). In the time series, dots are the mean daily values from four to six measurements and the whiskers are the minimum and maximum values. In the violin plots, black dots are the median, bars are the quartile 25–75%, and ****p* < 0.001. Note the logarithmic scale for the CH₄ emissions and the climatic forcing to improve visualization and the units of climatic forcing in g CO₂ equivalent. Dotted vertical lines represent the change from year 2021 to 2022. [Correction added on June 4, 2024, after initial online publication: In Figure 1, the part figure 1d has been replaced in this version.]

The climatic forcing due to emissions of the three GHGs was maximum during summer, reaching more than 10,000 mg CO₂ eq m⁻² d⁻¹ (Fig. 1g) with significantly

more climatic forcing during stratified conditions than when the reservoir was mixed (*t* = 7.8, *df* = 61, *p*-value < 0.001) (Fig. 1h). Overall, we found the maximum

climatic forcing during the summer due to the CH₄ and N₂O emissions.

In the case of CO₂ emissions, the main fixed drivers included surface water temperature and wind speed (details of the LMM in Supporting Information Table S1). The surface water temperature ranged from 8.92°C to 28.52°C (Fig. 2a). The wind speed ranged from 0 to 6.38 m s⁻¹ (Fig. 2b). Unexpectedly, the relationship between CO₂ emissions and water temperature was quadratic (Fig. 2c). The CO₂ emissions increased with the temperature up to approximately 20°C and then decreased at higher temperatures. This result differs from the previous ones reported in the literature, where the relationship between CO₂ emissions and temperature appears linear and positive (Yang et al. 2013; Zhao et al. 2013; Golub et al. 2023). This phenomenon could be related to the fact that in hardwater systems, as the study reservoir, dissolved inorganic carbon available for photosynthesis is mainly in the form of HCO₃⁻. Some species of microalgae and cyanobacteria have developed a mechanism that increases the bicarbonate inside the cells at the carboxysome, where some enzymes (e.g., carbonic anhydrase) convert this HCO₃⁻ into CO₂ to photosynthesis (Moroney and Ynalvez 2007). Therefore, photosynthesis coupled to calcite formation removes two bicarbonates: one HCO₃⁻ for calcite formation (CaCO₃ + H⁺) and the other one for photosynthesis that converts HCO₃⁻ into CO₂ and then in organic matter. Intracellular calcite formation during photosynthesis induces a substantial decrease in bicarbonate and does not release CO₂ into the water column.

In hardwater reservoirs, seasons or locations with high photosynthesis have been linked to calcite formation further suggesting that the coupling of the two processes may be an overlooked C sink with a substantial reduction in alkalinity (Deemer et al. 2020; Escoffier et al. 2023; Perolo et al. 2023). High temperatures promote cyanobacteria blooms (Paerl and Paul 2012) and, consequently, during summertime, this process could be accentuated, and CO₂ emissions reduced. López et al. (2011) showed that carbonate precipitation in summertime affects the metabolism–CO₂ emission relationship, and Waldo et al. (2021a) found that the best predictors of CO₂ emissions were surface pH and bicarbonate concentrations. The CO₂ emissions also showed a positive linear correlation with the wind speed (Fig. 2d). The higher wind speeds will increase CO₂ emissions since wind intensity promotes gas transfer from the water surface to the atmosphere. This observation is consistent with previous studies on the circadian scale (Liu et al. 2016).

The main fixed predictors of the diffusive CH₄ emissions were water temperature (Fig. 3a,d), water level (Fig. 3b,e), and NEP (Fig. 3c,f) (Supporting Information Table S1). Water level ranged from 636.16 to 639.46 m above sea level (i.e., from 3.01 to 6.26 m depth), and NEP ranged from 0.06 to 1.04 mg C L⁻¹ d⁻¹. In the case of the ebullitive fluxes, the main predictors were only water temperature (Fig. 3a,g) and water level (Fig. 3b,h). These robust correlations between water temperature and diffusive and ebullitive CH₄ emissions agree with previous studies (Yang et al. 2013; Yvon-Durocher et al. 2014;

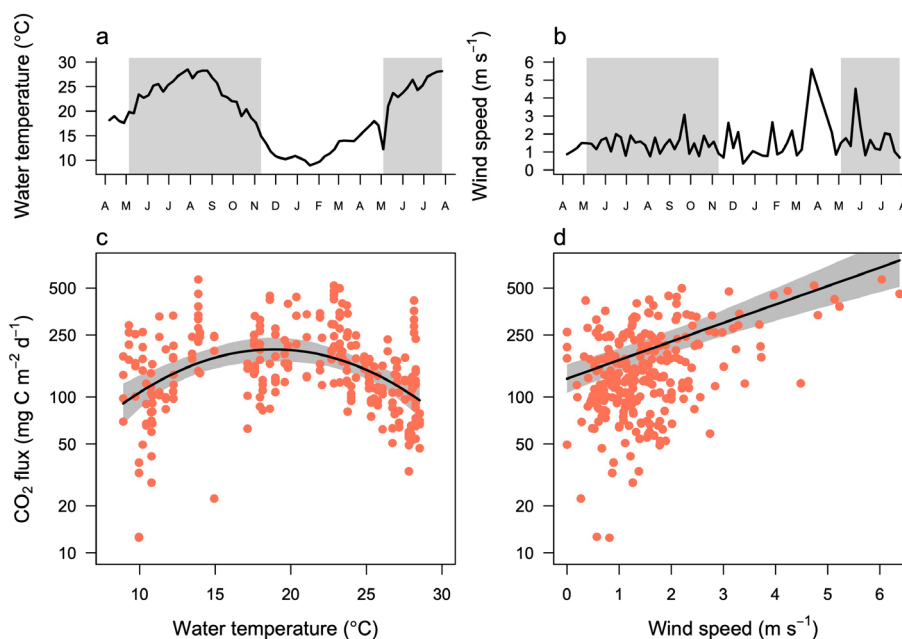


Fig. 2. LMMs for the main drivers of the CO₂ fluxes. Time series of water temperature (a) and wind speed (b). CO₂ emission as a function of water temperature (c) and wind speed (d). The red dots are the observations ($n = 268$), the black lines are the fit lines, and the gray areas are the 95% confidence intervals. Conditional $R^2 = 0.72$; marginal $R^2 = 0.36$.

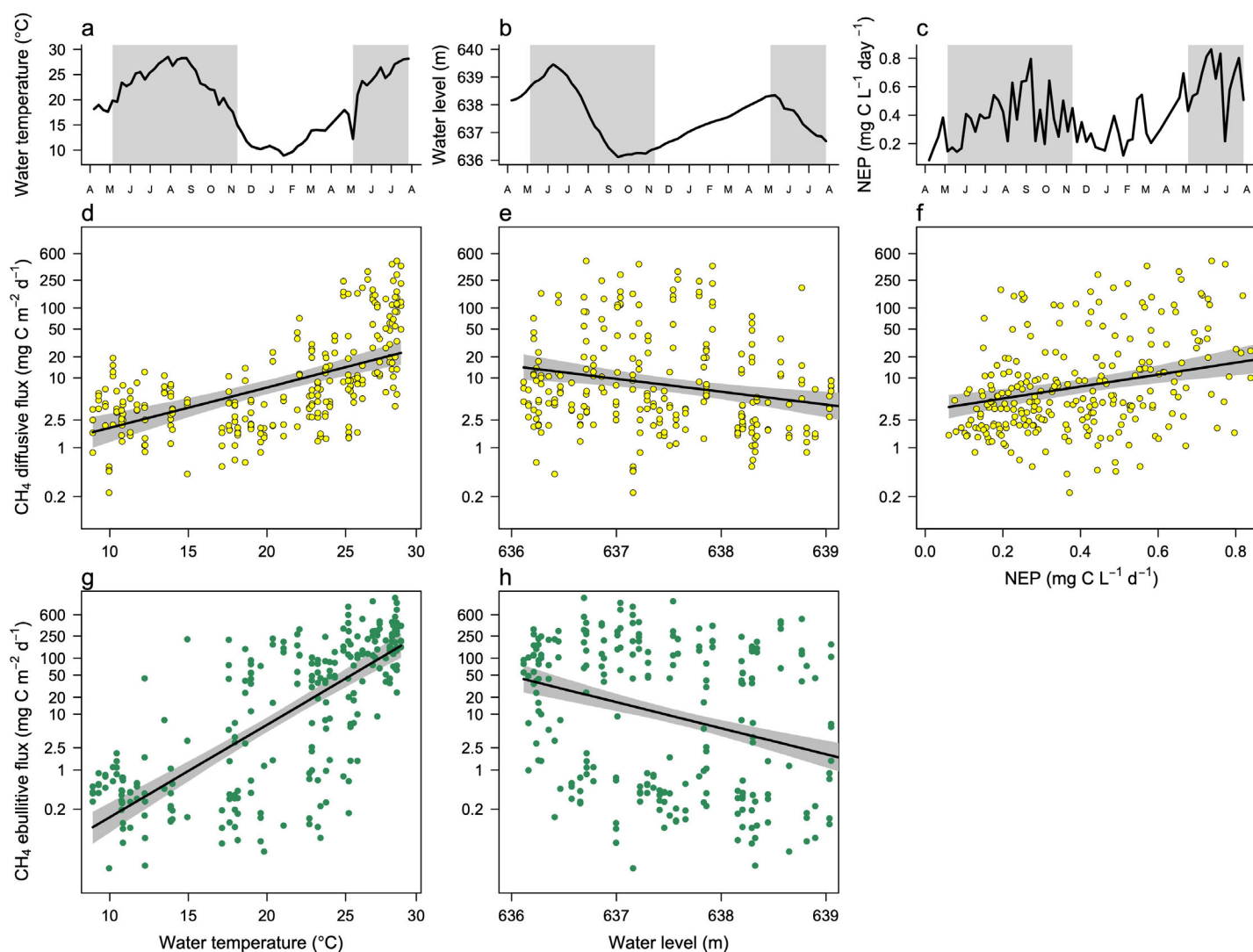


Fig. 3. LMMs for the main drivers of the CH₄ diffusive (conditional $R^2 = 0.74$; marginal $R^2 = 0.48$) and ebullitive fluxes (conditional $R^2 = 0.72$; marginal $R^2 = 0.63$). Time series of water temperature (**a**), water level (**b**), and NEP (**c**). CH₄ diffusive emission as function of water temperature (**d**), water level (**e**), and NEP (**f**). The yellow dots are the observations ($n = 268$), the black lines are the fit lines, and the gray areas are the 95% confidence intervals. CH₄ ebullitive flux as a function of water temperature (**g**) and water level (**h**). The green dots are the observations ($n = 250$), the black lines are the fit lines, and the gray areas are the 95% confidence intervals. Water level means meters above sea level.

Mosher et al. 2015; Linkhorst et al. 2020). High temperatures can intensify CH₄ emissions due to changes in solubility and an increase in sediment methanogenesis (Duc et al. 2010; Yvon-Durocher et al. 2014). In the study reservoir, this relationship is even steeper for the ebullitive flux (slope = 0.17) than for the diffusive flux (slope = 0.06) (Supporting Information Table S1; Fig. 3d,g). The effect of temperature on CH₄ emissions is maybe more noticeable in shallow systems, as the heat is transferred more efficiently throughout the water column to the sediments (Natchimuthu et al. 2016). The water level (i.e., hydrostatic pressure) was negatively correlated with both fluxes (Fig. 3e,h). This predictor, like the temperature, was more influential in the ebullition fluxes with a higher slope (−0.46)

than in the diffusion fluxes with a lower slope (−0.18) (Supporting Information Table S1). Previous studies have also reported similar results (DelSontro et al. 2011; Xiao et al. 2013; Harrison et al. 2017; Linkhorst et al. 2020; León-Palmero et al. 2020a). Harrison et al. (2017) and Beaulieu et al. (2018) showed that water level decline triggers the release of CH₄-rich bubbles from the sediments. Finally, the diffusive flux was also positively correlated with NEP (Fig. 3f). There are at least two non-exclusive underlying mechanisms to explain this relationship. First, there is evidence of a direct link between CH₄ production and photosynthesis by picoeukaryotes and cyanobacteria in surface waters (e.g., Klintzsch et al. 2019; Bižić et al. 2020; León-Palmero et al. 2020b), and previous works

have reported maximum fluxes associated with spring algae blooms (Waldo et al. 2021b). Second, methanogenesis in sediments appears to be boosted by phytoplanktonic organic matter exported to sediments (West et al. 2015, 2016). Indeed, we have observed a lagged response between phytoplankton-derived particulate organic matter and methane emissions in the study reservoir (Martínez-García et al. 2024). Similarly, Bertolet et al. (2020) found a relationship between the CH₄ storage in the hypolimnion and the GPP. This result could explain why we did not find a synchronous relationship between NEP and the CH₄ ebullitive flux. This pattern with maximum CH₄ emissions during summer was also observed in previous studies of temperate latitudes (Jacinthe et al. 2012; Beaulieu et al. 2014; McClure et al. 2020), and it differs from that observed in boreal lakes with peaks shortly after ice-off (Denfeld et al. 2018), in other temperate lakes with peaks during the fall mixing (Encinas Fernández et al. 2013) or in reservoirs that experience flood control drawdowns (Harrison et al. 2017).

The main fixed drivers of N₂O emissions were water temperature (Fig. 4a,d), wind speed (Fig. 4b,e), and GPP (Fig. 4c,f) (Supporting Information Table S1). The GPP values ranged from 0.22 to 3.29 mg C L⁻¹ d⁻¹ (Fig. 4c). Surprisingly, we did not find a significant relationship between the N₂O emissions and nitrate and these emissions decreased with increases of GPP. Previous studies have also shown positive relationships between N₂O emissions and temperature (Zhu et al. 2013; Musenze et al. 2014; Xiao et al. 2019; Yang et al. 2023) associated with increased microbial activity. An increment in wind

speed also stimulates the transference of N₂O to the atmosphere. León-Palmero et al. (2020a) also reported a relevant influence of wind speed on N₂O emissions. In previous studies, however, positive correlations between Chl *a* concentration (that could be considered as a surrogate of primary production) and the emission of N₂O have been reported (Xiao et al. 2019).

All the results above highlight the importance of temperature in the Mediterranean zone as a driver of the three GHGs, although the quadratic function obtained for CO₂ emissions needs further exploration to unravel its ultimate causes. Furthermore, autochthonous production only correlated synchronously with diffusive CH₄ emissions, but surprisingly not with CO₂ emissions.

Finally, we compared our results (black and white triangles in Fig. 5) with previous studies that reported GHG seasonal dynamics in reservoirs and lakes across latitudes (Supporting Information Tables S2–S4). In the case of CO₂ fluxes (Fig. 5a), the variability range was broader—including negative fluxes—in tropical/subtropical (0° to 30°), temperate low latitudes (light orange bands 30° to 45°) that include the Mediterranean zone, and temperate high latitudes (45° to 60°) than in the boreal (> 60°) latitude that presented a lower amplitude. In the case of CH₄ fluxes (Fig. 5b), all values were always positive (i.e., CH₄ sources). The most wide-ranging variability was also found in the systems from the 30°–45° latitudes (light orange bands), and the lowest ones in the boreal systems (Fig. 5b). The same observation was also reported by Johnson et al. (2021) using a

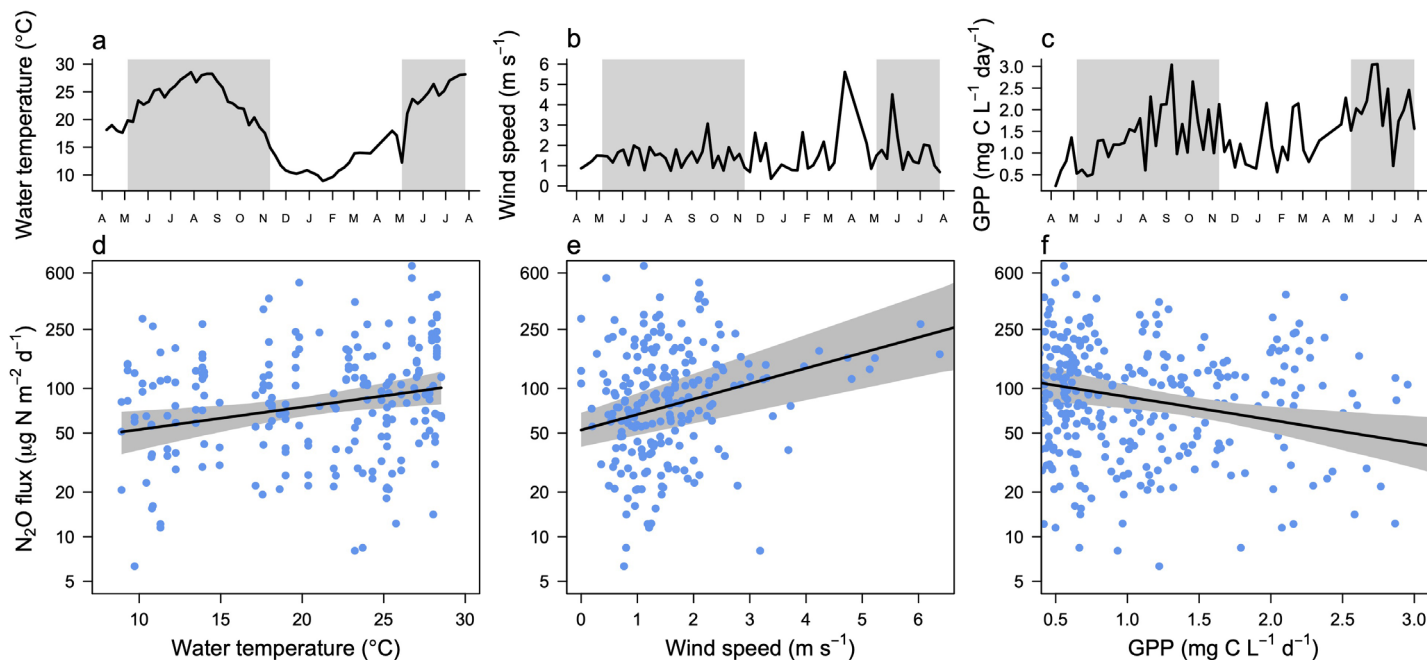


Fig. 4. LMMs for the main drivers of the N₂O fluxes. Time series of water temperature (a), wind speed (b), and GPP (c). N₂O emission as function of water temperature (d), wind speed (e), and GPP (f). The blue dots are the observations ($n = 218$), the black lines are the fit lines, and the gray areas are the 95% confidence intervals. Conditional $R^2 = 0.46$; marginal $R^2 = 0.13$.

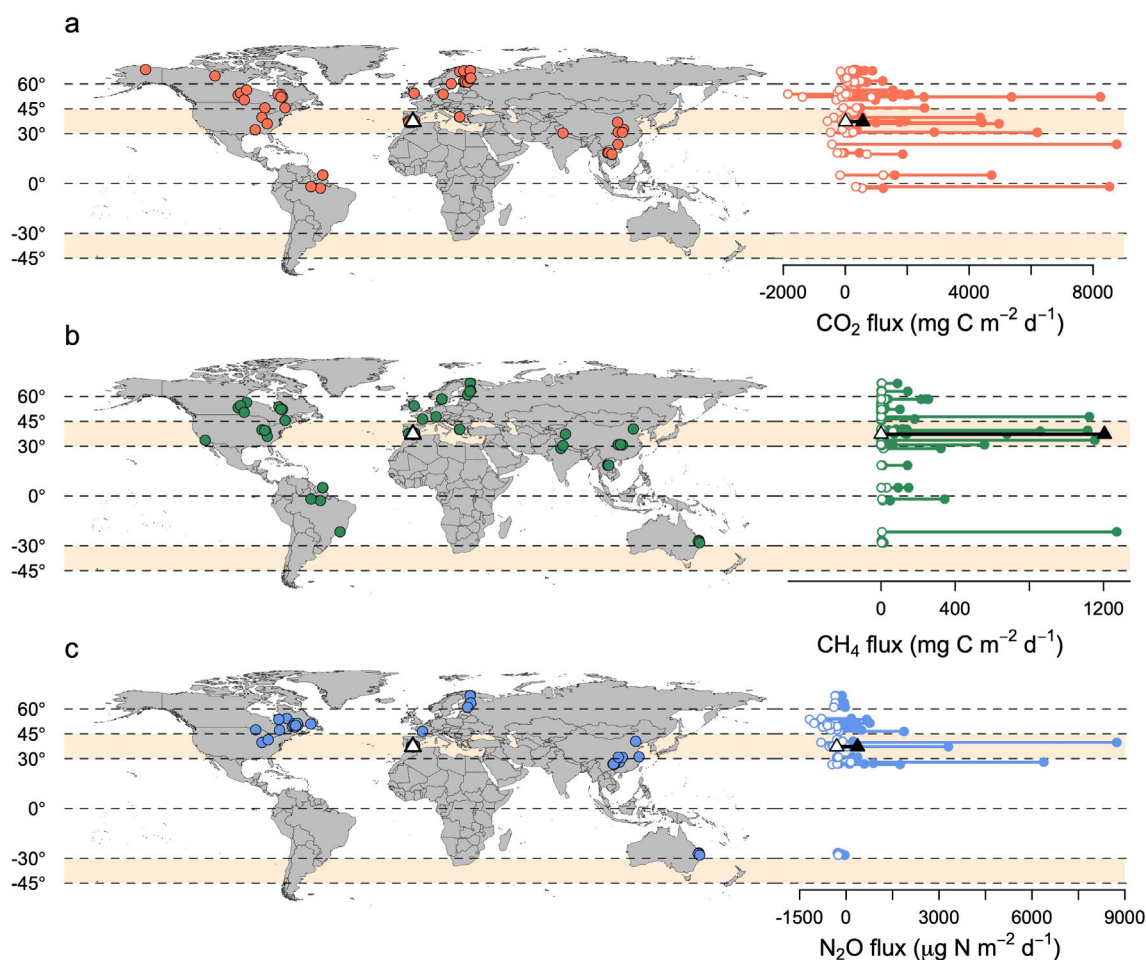


Fig. 5. Locations of the reservoirs and lakes with temporal variability of CO₂ (orange dots), CH₄ (green dots), and N₂O (blue dots) fluxes found in the literature (Supporting Information Tables S2–S4) and our study reservoir (white triangle). The magnitude of seasonal variability in the fluxes is represented by the line, the white dots represent the minimum values and the colored dots the maximum values, including the negative values for CO₂ fluxes (**a**) CH₄ fluxes (**b**), and N₂O fluxes (**c**). The variability of the study reservoir is represented by a line with white (minimum values) and black (maximum values) triangles. The temperate low latitude from 30° to 45° (including the Mediterranean zone) is represented with a light orange shadow.

different approach. For N₂O fluxes (Fig. 5c), the most extensive variability—including negative fluxes—was also observed in the systems located at 30° to 45° latitudes, and the smallest again in the boreal latitudes. Despite the lack of data for some regions, this comparison emphasizes the relevance of including seasonal variability, particularly from temperate low latitudes from 30° to 45°, in the upscaling models and exploring local drivers that can be easily obtained using remote sensing approaches such as primary productivity and inundation surface. In addition, new models should consider other sources of variability not included here, such as the differences in fluxes between day and night (Liu et al. 2016; Golub et al. 2023) and the high spatial variability in some reservoir fluxes (Colas et al. 2020; Linkhorst et al. 2020; Liu et al. 2021). This model improvement could provide more accurate projections under future scenarios of increasing temperatures and more extended stratification periods (Woolway and Merchant 2019) resulting

in anoxic hypolimnia (Jane et al. 2021), which can enhance methanogenesis and denitrification, as well as more severe eutrophication (Beaulieu et al. 2019).

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