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Oxic Methane Production Dynamics in Lakes

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Département F.-A. Forel des

Professeur Daniel Frank McGinnis

sciences de l'environnement et de l'eau

Oxic Methane Production Dynamics in Lakes

THÉSE

presenté à la Faculté de Sciences de l'Université de Genève pour obtenir le grade de Docteur ès Sciences, mention Science de l'Environnement

par

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de

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«Oxic Methane Production Dynamics in Lakes»

La Faculté des sciences, sur le préavis de Monsieur D. F. MCGINNIS, professeur associé et directeur de thèse (Département F.-A. Forel des sciences de l'environnement et de l'eau), Monsieur B. W. IBELINGS, professeur ordinaire (Département F.-A. Forel des sciences de l'environnement et de l'eau), Monsieur A. LORKE, professeur (Institute for Environmental Sciences, Natural and Environmental Sciences, The University of Koblenz-Landau, Landau, Germany), autorise l'impression de la présente thèse, sans exprimer d'opinion sur les propositions qui y sont énoncées.

Genève, le 30 mai 2022

Thèse - 5655 -

Le Doyen



Summary

Over the last decades, methane (CH₄) oversaturation in surface oxic waters in the ocean and inland waters has been widely reported, contrasting the paradigms that CH₄ is only produced in anoxic conditions (called "The methane paradox"). During the last decade, several studies have shown that different organisms are able to produce CH₄ in oxic conditions, however the contribution of oxic methane production (OMP) to CH₄ emissions from lakes versus the CH₄ produced in anoxic sediment has been hotly debated. Recent estimates show that CH₄ is responsible for about 20% of the radiative force and that emissions from freshwaters are about 30 to 60% of the global CH₄ budget. Since OMP occurs in surface waters, it can be rapidly emitted to the atmosphere. Therefore, to be able to predict CH₄ emissions from lakes, it is crucial to understand the OMP dynamics and its contribution to emissions in different lakes and at different time scales. In this doctoral thesis, modelling approaches are used to quantify the net production rate (P_{net}), which is defined as the balance between OMP (adds CH₄) and CH₄ oxidation (MOx, removes CH₄) in the lake surface mixed layer. I investigated the main physical and biochemical lake features that drive P_{net} rates and its contribution to diffusive CH₄ emissions over the stratified season in five natural lakes. Chapter 1 provides an overview of the state of the art on OMP and the CH₄ cycle in lakes.

Using an extensive dataset across trophic state gradients, Chapter 2 studies the occurrence and OMP contribution to diffusive CH_4 emissions in four pre-alpine lakes. With both a 0-D full-scale mass balance model and a 1-D lateral transport model, P_{net} rates in the surface mixed layer (SML) are quantified during three different times over the stratified season. In previous studies, these two models showed contradictory results mainly due to the use of parameterizations to estimate surface diffusive CH_4 emissions. Nevertheless, Chapter 2 results show a good agreement between the two models when correct boundary conditions are utilized. The conclusions illustrate that eutrophic lakes have higher P_{net} rates than oligotrophic lakes and the rates tended to be higher in late spring and decrease towards the end of the summer. Including literature data from additional lakes, light climate, Secchi disk depth and chlorophyll-a concentrations were identified as the main drivers of P_{net} , suggesting that photoautotrophs play an important role in CH_4 production in surface oxic waters. Using these data, an empirical up-scaling method is proposed to estimate P_{net} rates in different lake ecosystems. Chapter 2

concludes that OMP can be the dominant source of CH₄ in the pre-alpine lakes during the stratified season, but it highly depends on the physical and biochemical conditions in each lake.

So far, only one investigation reported the same trend that I observed in Chapter 2 using one year of data on Lake Stechlin (Germany). To further investigate the seasonal cycle of P_{net} rates, Chapter 3 analyzes data collected over four years in a small eutrophic lake (Soppensee, Switzerland). Using the same approaches from Chapter 2, P_{net} rates were estimated in the SML. The results confirm that P_{net} seasonal dynamics observed previously tend to occur every year. The stable isotopic signature ($\delta^{13}C_{CH_4}$) suggest that MOx dominated over OMP at the end of the summer as the residual CH_4 concentration became highly ^{13}C enriched. Chapter 3 also corroborates the role of photoautotrophs in OMP and highlights the importance of including the temporal seasonal variation to estimate the OMP contribution to the CH_4 atmospheric budget.

Using different phytoplankton cultures, it was shown that OMP was higher under light compared to dark conditions. In addition to the results observed in Chapter 2 and Chapter 3 regarding the role of phytoplankton on P_{net} rates, Chapter 4 investigates the diurnal variation of P_{net} rates in a small eutrophic lake located in the pre-alpine region (Lac de Bretaye, Switzerland). Using day and night measurements over three days, it is demonstrated that the variation of CH₄ concentration in time was only explained when P_{net} was included as an additional sources/sink in the SML. OMP tended to dominate over MOx during the day while during the night MOx dominated over OMP. Until now, all reported OMP studies were conducted during the day, therefore Chapter 4 concludes that OMP contribution to CH₄ emissions from lakes may be overestimated due to the lack of nighttime data and illustrates the importance of including the temporal diurnal cycle of P_{net} in the CH₄ budget.

Finally, Chapter 5 gathers the implications of this doctoral thesis on the role of OMP in the atmospheric CH_4 budget, the impact of the feedback between climate change, lake eutrophication and CH_4 emissions from lakes, and discusses potential future research. This doctoral thesis mainly focusses on OMP and its contribution to CH_4 emissions from lakes at different temporal scales and with varying physico-chemical conditions. However, more research is needed to validate the upscaling approach proposed to predict P_{net} rates in different lake ecosystems. Understanding the main sources of CH_4 emissions from freshwaters and identifying

their contribution on a global scale is necessary to be able to predict, adapt and mitigate future changes in the CH_4 cycle in inland waters.

Résumé

Au cours des dernières décennies, la sursaturation en méthane (CH₄) des eaux de surface oxiques de l'océan et des eaux intérieures a été largement signalée, contrastant avec les paradigmes selon lesquels le CH₄ n'est produit que dans des conditions anoxiques. Cette observation est appelée "le paradoxe du méthane". Ces dernières années, plusieurs études ont montré que différents organismes sont capables de produire du CH₄ dans des conditions oxiques, mais la contribution de la production de méthane oxique (PMO) aux émissions de CH₄ des lacs par rapport au CH₄ produit dans des sédiments anoxiques a été largement débattue. Des estimations récentes montrent que le CH₄ est responsable d'environ 20 % du forcage radiative et les émissions provenant des eaux douces représentent environ 30 à 60 % du budget global du CH₄. Comme la PMO est produit dans les eaux de surface, il peut être rapidement émis dans l'atmosphère. Par conséquent, pour être en mesure de prédire les émissions de CH₄ des lacs, il est crucial de comprendre la dynamique de la PMO et sa contribution aux émissions dans différents lacs et à différentes échelles de temps. Dans cette thèse de doctorat, des approches de modélisation sont utilisées pour quantifier le taux de production net (P_{net}), qui est défini comme l'équilibre entre la PMO (qui ajoute du CH₄) et l'oxydation du CH₄ (MOx, qui élimine le CH₄) dans la couche mixte de surface du lac. J'ai étudié les principales caractéristiques physiques et biochimiques des lacs qui déterminent les taux de P_{net} et sa contribution aux émissions diffusives de CH₄ pendant la saison stratifiée dans cinq lacs naturels. Le chapitre 1 donne une vue d'ensemble de l'état de l'art sur le P_{net} et le cycle du CH₄ dans les lacs.

En utilisant un grand ensemble de données avec un gradient d'état trophique, le chapitre 2 étudie l'occurrence et la contribution de la PMO aux émissions diffusives de CH₄ dans quatre lacs préalpins. Avec un modèle 0-D de bilan de masse à l'échelle réelle et un modèle 1-D de transport latéral, les taux de P_{net} dans la couche mélangée de surface (CMS) sont quantifiés à trois moments différents de la saison stratifiée. Dans les études précédentes, ces deux modèles ont montré des résultats contradictoires, principalement en raison de l'utilisation de paramétrages pour estimer les émissions diffusive de CH₄ à la surface. Néanmoins, les résultats du chapitre 2 montrent un bon accord entre les deux modèles lorsque les conditions correctes aux limites sont utilisées. Les conclusions montrent que les lacs eutrophes ont des taux de P_{net} plus élevés que les lacs oligotrophes et que les taux ont tendance à être plus important à la fin du printemps et

à diminuer vers la fin de l'été. En incluant des données bibliographiques provenant d'autres lacs, des facteurs tel que le climat lumineux, la profondeur du disque de Secchi et les concentrations de chlorophylle-a sont identifiés comme les principaux facteurs de P_{net} , ce qui suggère que les photoautotrophes jouent un rôle important dans la production de CH_4 dans les eaux oxiques de surface. En utilisant ces données, une méthode empirique à grande échelle est proposée pour estimer les taux de P_{net} dans différents écosystèmes lacustres. Le chapitre 2 conclut que la P_{met} dans différents écosystèmes lacustres. Le chapitre 2 conclut que la P_{met} dans les lacs préalpins pendant la saison stratifiée, mais cela dépend fortement des conditions physiques et biochimiques de chaque lac.

Jusqu'à présent, une seule étude rapporté la même tendance que celle observée dans le chapitre 2 en utilisant les données d'une année sur le lac Stechlin (Allemagne). Pour approfondir le cycle saisonnier des taux de P_{net} , le chapitre 3 analyse quatre ans de données provenant d'un petit lac eutrophique (Soppensee, Suisse). En utilisant les mêmes approches que dans le chapitre 2, les taux de P_{net} ont été estimés dans la CMS. Les résultats confirment que la dynamique saisonnière observée précédemment a tendance à se produire chaque année . La signature isotopique stable ($\delta^{13}C_{CH_4}$) suggère que le MOx a dominé sur la PMO à la fin de l'été lorsque la concentration résiduelle de CH_4 est devenue fortement enrichie en ^{13}C . Le chapitre 3 corrobore également le rôle des photoautotrophes dans la PMO et souligne l'importance d'inclure la variation saisonnière temporelle pour estimer la contribution de la PMO au bilan total du CH_4 atmosphérique.

En utilisant différentes cultures de phytoplancton, il a été démontré que la P_{net} était plus élevée dans des conditions lumineuses que dans des conditions d'obscurité. En plus des résultats observés dans les chapitres 2 et 3 concernant le rôle du phytoplancton sur les taux de P_{net}, le chapitre 4 étudie la variation diurne des taux de P_{net} dans un petit lac eutrophe situé dans la région pré-alpine (Lac de Bretaye, Suisse). En utilisant des mesures diurnes et nocturnes sur trois jours, il a eté démontré que la variation de la concentration de CH₄ dans le temps est expliquée lorsque P_{net} est inclus comme source/puits supplémentaire dans la CMS. La PMO a eu tendance à dominer le MOx pendant la journée alors que, pendant la nuit, le MOx a dominé la PMO. Par conséquent, le chapitre 4 conclut que la contribution de la PMO aux émissions de CH₄ des lacs peut être surestimée en raison du manque de données nocturnes et illustre l'importance d'inclure le cycle temporel diurne de P_{net} dans le bilan de CH₄.

Enfin, le chapitre 5 rassemble les implications de cette thèse de doctorat sur le rôle de la PMO sur le bilan atmosphérique du CH₄, l'impact de la rétroaction entre le changement climatique, l'eutrophisation des lacs et les émissions de CH₄ des lacs, et discute des recherches futures potentielles. Cette thèse de doctorat se concentre principalement sur l'étude de la PMO et sa contribution aux émissions de CH₄ des lacs à différentes échelles temporelles et à plusieurs conditions physico-chimiques. Cependant, des recherches supplémentaires sont nécessaires pour valider l'approche de mise à l'échelle proposée pour prédire les taux de P_{net} dans différents écosystèmes lacustres. La compréhension des principales sources d'émissions de CH₄ des eaux douces et l'identification de leur contribution à l'échelle mondiale sont nécessaires pour pouvoir prévoir, adapter et atténuer les changements futurs du cycle du CH₄ dans les eaux intérieures.

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Geneva, May 2022

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Curriculum Vitae

Introduction

1.1 Climate crisis

The global average surface temperature during 2011 - 2020 was 1.09 °C higher than the 1850 - 1900 time period (Fig. 1.1) and each of the last four decades have been successively warmer than the preceding decade since 1850 (IPCC 2021). Anthropogenic greenhouse gases (GHG) emissions are the main drivers of this tropospheric warming (Eyring et al. 2021). We are already experiencing several changes around the world caused by global warming, for example the increase by 0.2 m of the sea level between 1901 and 2018, the retreat of Arctic Sea ice, the acidification of the ocean and the increase on average global precipitation (Gulev et al. 2021).

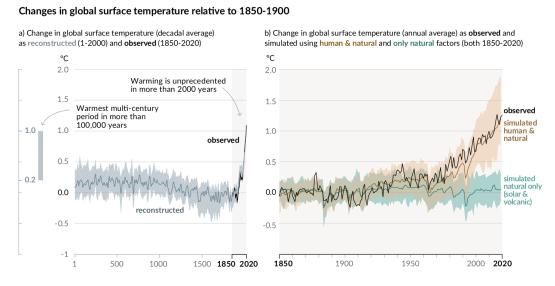


Figure 1.1. Change of global temperature relative to 1850-1900. Figure from: (IPCC 2021).

Gulev et al. 2021 showed that most of these changes are unprecedented. The rate of warming of the global air temperature since 1970 has not been observed for at least the last 2000 years (Fig. 1.1). During the summer of 2020, the sea ice area in the Arctic was the smallest in last 1000 years and in the last 2 million years, the ocean was not as acid as it is today.

The Intergovernmental Panel on Climate Change (IPCC) established that "It is unequivocal that human influence has warmed the atmosphere, ocean and land." (IPCC 2021). Global warming will continue with every emission scenario until at least 2050 and with every increment of global warming the environmental changes will be larger (Gulev et al. 2021). The ocean acidification and deoxygenation, the retreat of mountain and polar glaciers, and the sea level rise are irreversible for centuries to millennia (Lee et al. 2021).

Projected global GHG emissions from NDCs announced prior to COP26 would make it likely that warming will exceed 1.5°C and also make it harder after 2030 to limit warming to below 2°C.

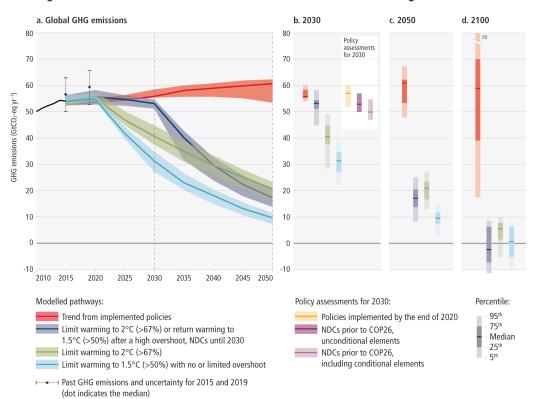


Figure 1.2. Global GHG emissions of modelled pathways (funnels in Panel $\bf a$ and associated bars in Panels $\bf b$, $\bf c$, $\bf d$) and projected emission outcomes from near-term policy assessments for 2030 (Panel $\bf b$). Panel $\bf a$ shows global GHG emissions over 2015-2050 for four types of assessed modelled global pathways. Figure from: IPCC 2022.

A decrease in the anthropogenic CO₂ emissions to at least net-zero and a rapid reduction

in CH₄ and other GHGs emissions are needed to limit the effect of global warming on the climate system (Fig. 1.2).

1.2 Methane in the atmosphere

A greenhouse gas is a gas that absorbs and emits radiation within the range of the radiation emitted by the Earth's surface, by the atmosphere and by clouds, causing the greenhouse effect. The primary GHGs are water vapor (H_2O), carbon dioxide (CO_2), nitrous oxide (N_2O), methane (CH_4), and ozone (O_3). After CO_2 , CH_4 is the second most important carbon-based GHG contributing about 16 - 23% of the additional radiative forcing to date (Fig. 1.3). Despite its low atmospheric concentration compared to CO_2 , its global warming potential (GWP) is \sim 84 times greater than CO_2 in a 20-yr timescale (Myhre *et al.* 2013).

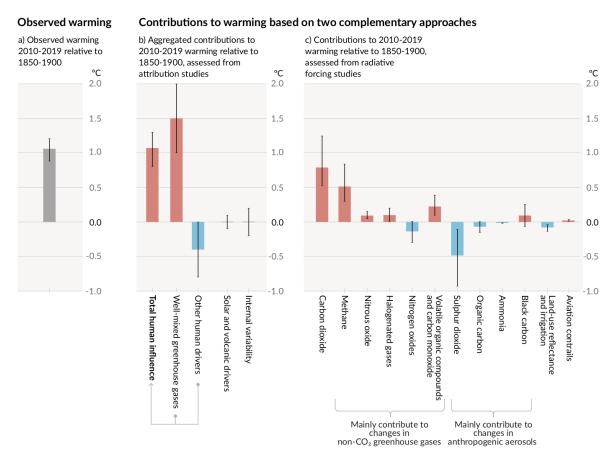


Figure 1.3. Contribution to observed warming in 2010 to 2019 relative to 1850 to 1900. Figure from: (IPCC 2021).

The atmospheric CH₄ concentration has nearly tripled from the pre-industrial era in

1750 to the current average value of ~ 1.89 part per million (ppm) in 2021 (Fig. 1.4). Data analysis from ice cores shows that this concentration level is unprecedented on the last 800,000 years (Meure et al. 2006; Loulergue et al. 2008). After a temporary pause in the increment of the CH₄ concentrations between late 1990s to 2006, they have been continuously increasing again at a rate of 7 ± 3 ppb/yr between 2006 to 2019 (Saunois et al. 2020). The renewed growth of CH₄ concentration after 2006 is due to an imbalance of about 3.5 % (20 Tg yr⁻¹) between the global sources and sinks. Currently, the uncertainty in the different components of the CH₄ budget greatly exceed this value. Therefore, the causes of the change in the trends over the last 30 years is highly debated (Turner et al. 2019).

Since the lifetime of CH₄ in the atmosphere is about 12.4 yr (Myhre *et al.* 2013), the radiative forcing caused by the increase of CH₄ in the atmosphere can be slowed down by reducing CH₄ emissions (Nisbet *et al.* 2021).

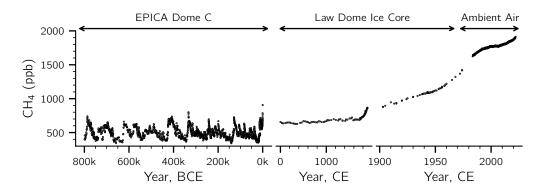


Figure 1.4. Atmospheric CH_4 concentration of the last 800,000 years. The BCE period (800,000 – 0 BCE) are from Vostok, EPICA Dome C ice cores (Loulergue *et al.* 2008). The data from 0 year CE are taken from Law Dome ice core analysis (Meure *et al.* 2006). And the ambient air concentration are taken from NOAA (www.esrl.noaa.gov/gmd/ccgg/trends_ch4, retrieved 26 April 2022). BCE = before current era, CE = current era.

1.3 Global methane sources and sinks

During 2000 - 2017, the global CH₄ emissions to the atmosphere estimated based on top-down and bottom-up methodologies are about 737 and 576 Tg CH₄ yr⁻¹ respectively (Fig. 1.5). The top-down methodology estimates CH₄ emissions based on satellite observations and inverse modelling whereas the bottom-up is the result of global anthropogenic inventories, land surface models for wetland emissions, and literature inventories from other natural sources (Saunois *et al.*

2020). About 50 to 60% of CH₄ emissions come from anthropogenic sources, where agricultural and waste activity are the largest contributors (about 211 Tg CH₄ yr⁻¹). The major natural sources are aquatic ecosystems such as wetlands and lakes (Saunois *et al.* 2020; Rosentreter *et al.* 2021). Chemical loss in the atmosphere (mainly by reaction with OH radicals) is the major sink of CH₄ with a minor contribution by oxidation in soils. The current imbalance between sources and sinks in the atmosphere is about 110 Tg CH₄ yr⁻¹ for the top-down and 13 Tg CH₄ yr⁻¹ for the bottom-up methodology. Lakes, rivers, reservoirs and wetlands are the greatest sources of uncertainty in the global CH₄ emissions to the atmosphere and therefore it is necessary to improve these estimates (Saunois *et al.* 2020).

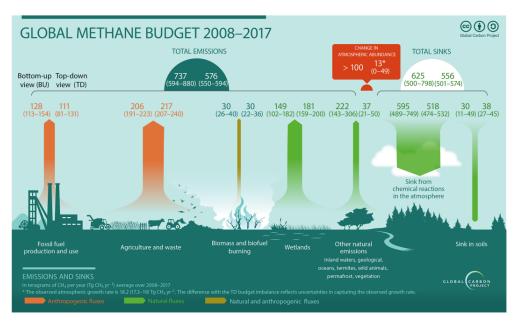


Figure 1.5. Global CH_4 budget between 2008 - 2017. Both bottom-up (left) and top-down (right) estimates (Tg CH_4 yr⁻¹). Figure from: (Saunois *et al.* 2020).

Methane emissions from lakes and reservoirs range about 69 to 179 Tg CH₄-C yr⁻¹ (Bastviken et al. 2008; DelSontro et al. 2018b; Rosentreter et al. 2021). The major sources of uncertainty are linked to the understanding of how CH₄ is produced and consumed in these ecosystems, the different transport pathways (ebullition, diffusion, plant mediation), the temporal and spatial variability, and statistical up-scaling methods (Bastviken et al. 2008; Rosentreter et al. 2021). To decrease the high variability of these data and predict future CH₄ emissions, further research is needed to understand the mechanisms of CH₄ productions and emissions in lakes.

1.4 Methane pathways in lakes

In lakes, CH₄ is generally produced under anaerobic conditions in the sediment as a final step of organic matter decomposition after other electron acceptors have been exhausted (Wetzel 2001). There are two dominant methanogenic pathways: acetoclastic methanogenesis using acetate as a substrate and hydrogenotrophic methanogenesis using CO₂ and H₂ as substrates (Conrad 2005).

$$CH_3COO_H \longrightarrow CO_2 + CH_4$$
 (1.1)

$$CO_2 + 4H_2 \longrightarrow 2H_2O + CH_4$$
 (1.2)

During these metabolic processes, CH₄ producing bacteria (methanogens) prefer lighter 12 C over heavier 13 C which alters the stable isotopic signature δ^{13} C (Mach et al. 2015). Methane oxidation (i.e., consumption of CH₄) in lake sediments and the water column is primarily attributed to aerobic pathways (King 1992). Aerobic CH₄ oxidizers preferentially utilize light carbon, leaving behind a CH₄ pool with a heavier residual δ^{13} C_{CH₄} signature (Whiticar 1999). Therefore, knowing the stable-C isotopes can help determine the origin and fate of CH₄ in aquatic systems.

In stratified lakes, CH₄ that is produced in the sediment is transported by diffusion to the water column (F_s) and can also be introduced by sediment-released bubbles (McGinnis *et al.* 2006). Methane is ~27 times less soluble than CO₂ (Sander 2015), and when its production is faster than CH₄ oxidation (MOx) and diffusion, bubbles can be formed (Schmid *et al.* 2017). These bubbles efficiently pass through the diffusive barrier at the sediment-water interface and depending on the depth of the lake, CH₄ in bubbles can bypass oxidation and reach the atmosphere with some fraction of the CH₄ partially dissolving into the water column (R_{dis}) (McGinnis *et al.* 2006).

Dissolved CH₄ in the water column is transported vertically by turbulent diffusion (F_z). This flux is directly proportional to the vertical concentration gradient of CH₄ and turbulent diffusivity (K_z), which depends on the kinetic energy in the water column and the stratification strength (Imboden & Wüest 1995). Right below the oxycline/thermocline dissolved CH₄ is mainly oxidized in a narrow zone (Rudd *et al.* 1976; Oswald *et al.* 2015; Thottathil *et al.* 2019).

In stratified lakes with anoxic hypolimnia, CH₄ is stored in the hypolimnion until turnover when a portion is released to the atmosphere (Schubert *et al.* 2010; Vachon *et al.* 2019).

Once CH_4 reaches the surface it is emitted to the atmosphere via diffusion (F_a), which depends on the surface water concentration and the gas transfer velocity (k_{CH_4}). The latter is mainly controlled by the turbulence level at the surface layer (MacIntyre *et al.* 2010), and potentially enhanced by microbubbles (Melack & Kilham 1974; McGinnis *et al.* 2015). Fig. 1.6 shows the classical scheme of the components of the CH_4 budget in lakes considering only anoxic sources during the stratified season.

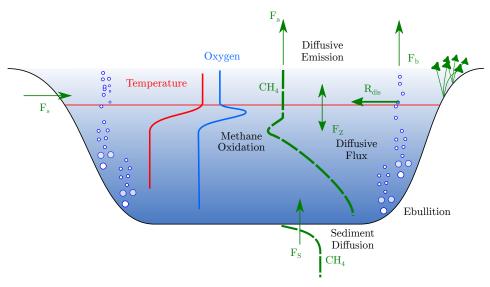


Figure 1.6. Schematic of CH_4 budget in lakes during the stratified period with typical profiles of temperature, oxygen (O_2) and CH_4 concentrations considering only anoxic sources. CH_4 mass balance components: diffusive CH_4 emissions to the atmosphere (F_a) , vertical turbulent diffusion (F_z) , CH_4 diffusive flux from littoral sediments (F_s) , bubbles emissions (F_b) , CH_4 oxidation and CH_4 dissolution from bubbles (R_{dis}) .

1.5 The methane paradox

The paradigm that CH₄ is only produced in anoxic conditions has been challenged by measurements showing widespread CH₄ oversaturation in oxic surface water in the ocean and inland waters (Karl et al. 2008; Tang et al. 2016). Using laboratory lake water incubations, Grossart et al. 2011 were the first to demonstrate that CH₄ production in oxic conditions (also called oxic methane production or OMP) could occur in lakes and could substantially contribute to epilimnetic CH₄ oversaturation and thus emissions. Subsequently, Bogard et al. 2014 showed

that a new source of CH₄ needed to be included in the epilimnion using an in-lake mesocosm experiment. Tang et al. 2014 demonstrated the CH₄ concentration peak found in the thermocline in Lake Stechlin did not originate from anoxic sources (CH₄ production in littoral or bottom sediments, or anoxic microniches) and in-situ CH₄ production needed to be included. Donis et al. 2017, using a full-scale mass balance, illustrated that OMP rates were higher in the surface mixed layer (SML) than in the metalimnion in Lake Hallwil (Switzerland). This new recognized source of CH₄ in lakes surface waters can therefore be rapidly emitted to the atmosphere (Tang et al. 2016).

Numerous mechanisms have been proposed for OMP in lakes and oceans such as archaeal methanogenesis in anoxic micro-niches (Oremland 1979; Grossart et al. 2011), associations between methanogens and photoautotrophs (Grossart et al. 2011; Bogard et al. 2014; Tang et al. 2014; Yao et al. 2016b) and dissolved organic matter photoproduction (Xie et al. 2019; Li et al. 2020). Moreover, OMP has been shown to be produced by methylphosphonate (MPn) biodegradation in phosphorus limited environments (Karl et al. 2008; Repeta et al. 2016; Wang et al. 2017; Khatun et al. 2019) and microbial CH₄ formation as a product of nitrogen fixation (Zheng et al. 2018; Luxem et al. 2020).

Recent studies show strong evidence that CH₄ can be produced by several phytoplankton species (Bižić et al. 2020; Günthel et al. 2020; Hartmann et al. 2020; Morana et al. 2020). Correlations between CH₄ concentrations and light, oxygen and phytoplankton concentrations suggest a direct role of metabolic processes in the production of CH₄. These processes might be linked with the ability that all organisms may have to produce CH₄ based on the combined reactions between reactive oxygen species, iron species and methyl donors (Ernst et al. 2022).

1.5.1 OMP controversy

Donis et al. 2017 and Günthel et al. 2019 showed, for the first time, that during the stratified season OMP could contribute up to $\sim 70\%$ to the diffusive CH₄ emissions from lakes. Both studies performed a mass balance assuming the SML of the lake as a well-mixed reactor and measured sources and sinks of CH₄. These results contrast with other investigations that show using water incubations (Morana et al. 2020) and 1-D lateral transport model (Peeters et al. 2019) that CH₄ diffusive flux from littoral sediments was enough to compensate the loses of

 CH_4 in the SML.

The CH₄ concentration in the SML is mostly defined by the lateral inputs from littoral sediments, OMP and MOx which can be affected by environmental factors in the lake. For example, the presence of oxygen on top of littoral sediment may decrease the CH₄ flux release to the water column (Damgaard et al. 1998; Liikanen et al. 2002). Conversely, warmer temperatures can increase CH₄ production in the sediment (Nozhevnikova et al. 1997) potentially increasing F_s. In the water column, laboratory water incubations experiments have shown that MOx can be inhibited by light (Shelley et al. 2017; Morana et al. 2020) and by elevated oxygen concentrations (Thottathil et al. 2019). However, only a few studies have so far focused on the effect of environmental conditions on OMP. Recent studies showed that OMP rates follow the dark/light cycle (Bižić et al. 2020; Hartmann et al. 2020) and increase at warmer temperatures (Klintzsch et al. 2020). Nevertheless, the drivers, temporal dynamics and contribution to diffusive CH₄ emissions are still unknown.

1.6 Climate change and eutrophication impacts on CH₄ cycle

Climate change and eutrophication are the two mayor threat to lakes and inland waters. The increase of air temperature due to climate change will lead to warmer surface water temperatures in lakes (Adrian et al. 2009; Woolway et al. 2020). Since it has been shown that both methanogenesis and ebullition exponentially increase with temperature (Nozhevnikova et al. 1997; DelSontro et al. 2010), the littoral zone will be highly impacted by the increase of surface temperature. MOx also increases with temperature, but slower than CH₄ production rates (Thottathil et al. 2019). Longer stratification periods are also projected (Woolway et al. 2021) promoting anoxia in the hypolimnion (Hadley et al. 2014) which could potentially increase methanogenesis in pelagic sediments and hypolimnetic CH₄ storage (Vachon et al. 2019). Global warming will also increase the ice free period, allowing CH₄ to be emitted for a longer time. The projected increase of human population will also enhance the loadings of agricultural fertilizers and sewage to inland waters (Beaulieu et al. 2019 and references therein). Moreover, heavier rainfall would enhance the load of dissolved organic matter and nutrients in aquatic ecosystems. The combination of increased nutrient loading and higher temperatures will promote the growth of algae in inland waters (Rigosi et al. 2014). Correlations between CH₄ emissions

and chlorophyll-a concentrations (DelSontro et al. 2018a) suggest an increase of CH₄ emissions from lakes as a result of lake eutrophication (Beaulieu et al. 2019).

High uncertainties remain as to the magnitude of the impact that climate change and eutrophication may simultaneously have on the CH₄ cycle in lakes. Numerical models considering physical and biochemical processes have been developed to investigate the fate of CH₄ concentrations in lakes (Tan et al. 2015; Stepanenko et al. 2016; Schmid et al. 2017). Nevertheless, several parameters used in these models are often unknown or poorly constrained. Moreover, processes such as OMP need to be included to be able to project CH₄ emissions from lakes in a changing environment. Therefore, further research needs to be conducted to understand the mechanisms behind the CH₄ cycle in lakes.

1.7 Objectives

Most of the CH₄ investigations have been focused on CH₄ emissions from lakes and less research has been done to understand the basic processes driving those emissions and their interaction with environmental changes (Soued & Prairie 2021). OMP is a newly recognize source of CH₄ to the atmosphere but its importance in the CH₄ cycle and its occurrence remain uncertain. This doctoral thesis investigates the CH₄ sources and sinks using in-situ measurements and modelling approaches with focus on the physical aspects.

This thesis was conducted in five natural lakes, four in the Swiss pre-alpine region and one in the Swiss plateau. Detailed information about the study sites can be found on each individual chapters.

This doctoral thesis addresses the following research goals:

- 1. Use mass balance approaches to determine P_{net} in the surface mixed layer and its contribution to diffusive CH₄ emissions in five natural lakes.
- 2. Use a multi-lake survey to investigate the occurrence of OMP and its main drivers.
- 3. Determine the seasonal and diel P_{net} cycle during the stratified season.

1.8 Chapter organization

The investigations of the PhD thesis are organized in three individual manuscripts (Chapters 2 to 4) and conclusions and perspective can be found in Chapter 5.

- Chapter 2 Towards a global scaling of oxic methane production: evaluation of the methane paradox in four Swiss pre-alpine lakes.

 First author publication. Ordóñez, C., DelSontro, T., Langenegger, T., Donis, D., Suarez, E. L. & McGinnis, D. F. Towards a Global Scaling of Oxic Methane Production: Evaluation of the Methane Paradox in Four Swiss Pre-Alpine Lakes.

 Nature Communications (submitted) (2022)
- Chapter 3 Temporal dynamics of oxic methane production in a hypertrophic Swiss lake.

 First author publication. Ordóñez, C., Massot, A., DelSontro, T., Langenegger,

 T. & McGinnis, D. F. Temporal Dynamic of Oxic Methane Production in a

 Hypereutrophic Swiss Lake. Frontiers in Environmental Science (submitted)

 (2022)
- Chapter 4 Diurnal cycle of oxic methane production in a pre-alpine lake (Lac de Bretaye).

 First author manuscript to be submitted for publication 2022.
- Chapter 5 Conclusions and Perspective

Towards a global scaling of oxic methane production: evaluation of the methane paradox in four Swiss pre-alpine lakes

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Content of Chapter 2 was submitted in:

Ordóñez, C., DelSontro, T., Langenegger, T., Donis, D., Suarez, E. L. & McGinnis, D. F. Towards a Global Scaling of Oxic Methane Production: Evaluation of the Methane Paradox in Four Swiss Pre-Alpine Lakes. *Nature Communications (submitted)* (2022)

Abstract

Oxic methane production (OMP) in freshwaters is more widely accepted, but the mechanisms behind it and its contribution to local and global methane (CH₄) emissions are not well constrained. We use a full-scale mass balance and a lateral transport model to analyze sources and sinks of CH₄ in the surface mixed layer of four pre-alpine lakes in the Swiss Alps. With the appropriate parameterizations, we find that these two approaches agree well, suggesting that OMP occurs in the studied lakes and is often the dominant source of total CH₄ emissions. Correlations between OMP versus algal concentration, Secchi depth and mixed layer depth suggest a link between photosynthesis and CH₄ production that provides an empirical approach to upscale OMP rates globally. OMP is a newly recognized direct CH₄ source in surface waters that is rapidly emitted to the atmosphere. The extent of OMP and drivers need better understanding, especially considering the potential positive feedback between climate change, phytoplankton production, OMP and CH₄ emissions to the atmosphere.

2.1 Introduction

The widely reported methane (CH₄) oversaturation in surface oxic waters in oceans (Karl et al. 2008) and lakes (also referred to as "the methane paradox"; Tang et al. 2014) contrasts with the current understanding that biogenic CH₄ formation occurs exclusively under anoxic conditions (Conrad 2009). Methane production in oxic conditions (also called oxic methane production or OMP) has been reported for an increasing number of lakes (Grossart et al. 2011; Bogard et al. 2014; Tang et al. 2014; Donis et al. 2017; Günthel et al. 2019; Hartmann et al. 2020). While recent studies have shown that OMP may contribute up to 80% of the lake diffusive CH₄ emissions (Donis et al. 2017; Günthel et al. 2019), other researchers argued that CH₄ produced in anoxic littoral sediments is enough to resolve the methane paradox (Hofmann et al. 2010; Encinas Fernández et al. 2016; Peeters et al. 2019; Morana et al. 2020). Thus, the OMP contribution to global lake CH₄ emissions thus remains unclear and hotly debated (Günthel et al. 2021; Peeters & Hofmann 2021).

Atmospheric CH₄ concentrations have more than doubled since the industrial era (Hartmann, D. L. et al. 2013; Nisbet et al. 2019). Although CH₄ is less abundant in the

2.1. INTRODUCTION 15

atmosphere than carbon dioxide (CO₂), its global warming potential (GWP) is about 81 times higher than carbon dioxide (CO₂) in a 20-year period (Forster et al. In Press). Its GWP combined with its ~12 year lifetime means that reducing CH₄ emissions is a priority for mitigating climate change (Nisbet et al. 2021). Lakes represent between 20 to 50 % of natural CH₄ atmospheric sources, but large uncertainties remain about the contribution of internal sources and sinks (Bastviken et al. 2011; Saunois et al. 2020; Rosentreter et al. 2021). Methane in lakes can be emitted to the atmosphere through bubbles (ebullition) and diffusive fluxes at the air-water interface (AWI) (Bastviken et al. 2011). Ebullition is mainly driven by high CH₄ production rates in anoxic sediments (Langenegger et al. 2019) and the diffusive fluxes at the AWI are driven by CH₄ concentrations and turbulence level in surface waters (MacIntyre et al. 2010). Unlike anoxic CH₄ sources in sediments and bottom waters, OMP occurs in oxic surface waters where is quickly emitted to the atmosphere (Tang et al. 2014); however, the processes behind OMP and its drivers remain unconstrained.

Several mechanisms have been proposed for OMP such as archaeal methanogenesis in anoxic micro-niches (Oremland 1979; Grossart et al. 2011), associations of methanogens with photoautotrophs (Grossart et al. 2011; Bogard et al. 2014; Tang et al. 2014; Yao et al. 2016a), dissolved organic matter photoproduction (Xie et al. 2019; Li et al. 2020), biodegradation of methylphosphonate (MPn) in phosphorus limited environments (Karl et al. 2008; Repeta et al. 2016; Wang et al. 2017; Khatun et al. 2019) and CH₄ as a byproduct of nitrogen fixation (Zheng et al. 2018; Luxem et al. 2020). However, recent studies have shown strong evidence pointing to a direct role of phytoplankton in OMP (Bižić et al. 2020; Günthel et al. 2020; Hartmann et al. 2020) indicated by positive correlations between CH₄, oxygen and phytoplankton concentrations (Grossart et al. 2011; Hartmann et al. 2020). Moreover, it has been shown that OMP follows light-dark cycles in different phytoplankton cultures (Bižić et al. 2020; Günthel et al. 2020), but the processes behind photosynthesis-derived CH₄ are unclear. It is likely that multiple pathways produce CH₄ in oxic lake environments, and that these may vary from lake-to-lake and seasonally on the basis of trophic properties and light conditions.

OMP rates have been reported using different methodologies, such as laboratory and in situ water incubations (Grossart *et al.* 2011; Bižić *et al.* 2020; Günthel *et al.* 2020), inlake mesocosms (Bogard *et al.* 2014; Günthel *et al.* 2019), a physical lateral transport model

(DelSontro et al. 2018b), and lake mass balances (Donis et al. 2017; Günthel et al. 2019). OMP has been reported from different freshwater environments, including both temperate and arctic regions (DelSontro et al. 2018a; Li et al. 2019), high altitudes lakes (above 2500 m.a.s.l) (Perez-Coronel & Beman 2020) and across a range of trophic states (DelSontro et al. 2018a; Khatun et al. 2020; Ye et al. 2020). While these studies show the occurrence of OMP in lakes across a geographic and trophic gradient, OMP has so far not been investigated in pre-alpine lakes.

Although not typically thought to promote CH₄ emissions (Beaulieu et al. 2019), prealpine lakes (from 1300 to 2000 m.a.s.l) have been shown to be sources of CH₄ to the atmosphere (Rinta et al. 2017; Pighini et al. 2018). Pre-alpine lakes are also disproportionately experiencing climate change (Thompson et al. 2005; Råman Vinnå et al. 2021) with air temperature increasing two times faster in the European Alps than the global mean (Cannone et al. 2008). Such an increase in air temperature can induce the following changes in pre-alpine lakes with implications for aquatic CH₄ emissions (Thompson et al. 2005; Råman Vinnå et al. 2021; Sharma et al. 2021): (1) a longer ice-free season that allows CH₄ to be emitted for a longer period (Guo et al. 2020); (2) an increase in surface water temperature that enhances littoral production rates of CH₄ (Bastviken et al. 2008); (3) a longer stratified season that allows for more CH₄ accumulation in the hypolimnion (Vachon et al. 2020). These impacts, however, will differ across lakes depending on the light regime and trophic state (Flaim et al. 2016), therefore the precise impact of climate change on the CH₄ budget in pre-alpine lakes needs further investigation.

This study focuses on four adjacent pre-alpine lakes (~ 1700 m.a.s.l) with identical climate forcing but different trophic states. We quantified net CH₄ production (P_{net}, Fig. 2.1), which is defined as the balance between OMP (that adds CH₄) and CH₄ oxidation (MOx, that removes CH₄) from the surface mixed layer (SML) (King 1992; Bastviken *et al.* 2002; Thottathil *et al.* 2019). P_{net} in the SML was estimated using two independent approaches: a 0-D full-scale mass balance following Donis *et al.* 2017 and a 1-D lateral transport model similar to Peeters *et al.* 2019. In the full-scale 0-D model, we used CH₄ bulk sources and sinks in the oxic SML. The lateral transport model of Peeters *et al.* 2019 was modified to include additional relevant terms, such as diffusive CH₄ flux across the thermocline and CH₄ bubble dissolution. In this model, we included all of the CH₄ sources in the SML and the loss to the atmosphere

2.2. RESULTS 17

to determine P_{net} when the simulated concentrations were best-fit to the CH₄ concentrations measured across the lake.

In previous studies, these two models have shown contradictory results mainly due to the use of literature parameterizations to estimate surface diffusive emissions (e.g. Donis et al. 2017; Günthel et al. 2019; Peeters et al. 2019; Günthel et al. 2021; Peeters & Hofmann 2021). We used direct flux measurements from floating chambers and we found an excellent agreement between the two models. The results indicate that OMP substantially contributes to diffusive emissions during the stratified period of pre-alpine lakes with different trophic states. Moreover, we performed a sensitivity analysis using five diffusive flux literature parameterizations and surface flux measurements to analyze the impact that modeled versus measured atmospheric diffusive fluxes has on OMP estimation. Finally, we present two OMP upscaling approaches: the first based on chlorophyll a (Chla) concentrations, light penetration, and SML depth, and the second based on dissolved inorganic nitrogen (DIN) concentrations in the SML. Our findings highlight the need for OMP to be included in CH₄ lake budgets and for more research to understand OMP drivers and its response to climate change.

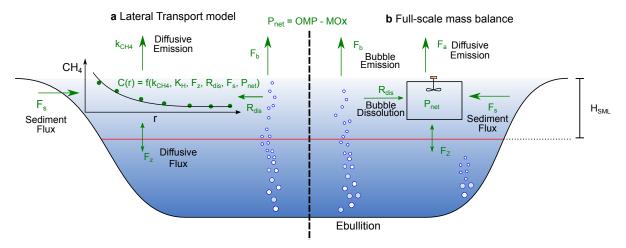


Figure 2.1. Conceptual schematic of the CH_4 budget components in the SML used for the **a** Lateral transport model and the **b** Full-scale mass balance.

2.2 Results

Here we present results from four pre-alpine lakes using two mass balance approaches described in Eq. 2.2 and Eq. 2.3. First, we present a general description of the study sites and then the measured sources and sinks of CH₄ that represent the boundary conditions for the applied models (Fig. 2.1). We then quantify and compare the CH₄ production rates in the surface mixed layer (SML) obtained from both models.

2.2.1 Study sites.

The four pre-alpine lakes studied - Lac Bretaye (BRE), Lac Noir (NOI), Lac des Chavonnes (CHA) and Lac Lioson (LIO) - are located between 1650 to 1850 m.a.s.l in the Swiss Alps and are hypereutrophic, eutrophic, mesotrophic and oligotrophic, respectively (Table S.A.1). NOI and BRE are small lakes with a maximum depth of ~ 9 m, while CHA and LIO have a maximum depth of ~ 28 m (Fig. S.A.1 and Table S.A.1). Throughout the three sampling campaigns conducted in June 2018, September 2018 and July 2019, the surface waters of all four lakes were oxic but oversaturated in CH₄ (Fig. S.A.2; Table 2.1). Temperature and CH₄ concentration profiles at the deepest point of the lakes show that all the lakes were stratified with a SML thickness between 1 - 6 m. Secchi disc depths, dissolved inorganic nitrogen (DIN), dissolved phosphorus (DP) and Chla concentrations in the SML are summarized in Table 2.1.

Table 2.1. Spatial average of surface CH_4 concentration and its stable isotopic signature ($\delta^{13}C_{CH_4}$) along each transect. Average dissolved inorganic nitrogen (DIN), Chla, and dissolved phosphorus (DP) concentrations in the SML. Secchi and SML depth (H_{SML}) at each sampling campaign in each lake. $\Delta CH_4/CH_4$ shore is the percentage variation of the CH_4 concentration at center and the shore. The values marked with * signify there is a significant difference between shore and center as determined with an ANOVA analysis.

Lake	Date	$\mathrm{CH_4} \pmod{\mathrm{m}^{-3}}$	$\delta^{13}C_{CH_4} \ (\%)$	$\begin{array}{c} \Delta \mathrm{CH_4/CH_4~shore} \\ (\%) \end{array}$	Secchi depth (m)	H _{SML} (m)	$\frac{\mathrm{Chl}a}{(\mathrm{mg}\mathrm{m}^{-3})}$	$\mathrm{DIN} \atop \left(\mathrm{mgm^{-3}}\right)$	$\mathrm{DP} \atop \left(\mathrm{mgm^{-3}}\right)$
Bretaye	June 2018	6.7 ± 2.3	-52.0	54	3.7	1.3	3.01	18	9.0
	Sept 2018	3.5 ± 0.5	-38.0	22*	3.0	5.2	4.08	29	7.3
	July 2019	2.8 ± 1.6	-48.8	4	4.7	2.6	4.05	4	57
Noir	June 2018	1.4 ± 0.1	-54.5	18*	2.8	0.9	8.81	18	2.3
	Sept 2018	1.8 ± 0.4	-45.5	19	6.1	5.4	4.71	13	2.7
	July 2019	3.9 ± 0.3	-49.9	23	3.8	1.9	8.48	BD	BD
Chavonnes	June 2018	0.1 ± 0.1	-62.3	59*	4.6	1.3	3.73	235	2.0
	Sept 2018	0.2 ± 0.1	-62.4	22	5.2	4.6	2.51	167	1.0
	July 2019	0.1 ± 0.0	-61.2	120	3.8	2.0	5.02	189	BD
Lioson	June 2018	0.1 ± 0.0	-50.9	33*	9.0	0.9	1.52	126	2.0
	Sept 2018	0.4 ± 0.6	-50.1	12*	10.5	6.1	3.01	45	1.0
	July 2019	0.2 ± 0.2	-54.0	14	5.5	2.2	4.64	71	BD

2.2.2 Surface methane concentration and isotopic signature.

For each lake, surface CH₄ concentrations and their stable isotopic signatures ($\delta^{13}C_{CH_4}$) were measured at the deepest point of the lake and along a transect from shore to shore to resolve the spatial variability (Fig. 2.2 and Fig. S.A.3). All four lakes were oversaturated with CH₄ values near the shore 33 \pm 32% higher than in the center, although only 40% of the time this difference

2.2. RESULTS 19

was significant (Table 2.1). The eutrophic lakes BRE and NOI, on average, had one order of magnitude higher surface concentrations (3.13 \pm 2.09 mmol m⁻³) than the oligo/mesotrophic lakes LIO and CHA (0.15 \pm 0.13 mmol m⁻³) (Table 2.1). The spatial average $\delta^{13}C_{CH_4}$ signature ranged between -62 to -38 % (Table 2.1). Isotopically enriched CH₄ ($\delta^{13}C_{CH_4} \sim -40$ %) was observed at the end of summer in the SML of the eutrophic lakes, while in the oligotrophic lakes $\delta^{13}C_{CH_4}$ was relatively consistent between sampling dates (Table S.A.4). Relatively constant $\delta^{13}C_{CH_4}$ values were observed along the transect for most of the lakes, except for CHA in June 2018 when lighter $\delta^{13}C_{CH_4}$ was observed at the shore (~ -65 %) than in the center of the lake (~ -60 %) (Fig. S.A.3).

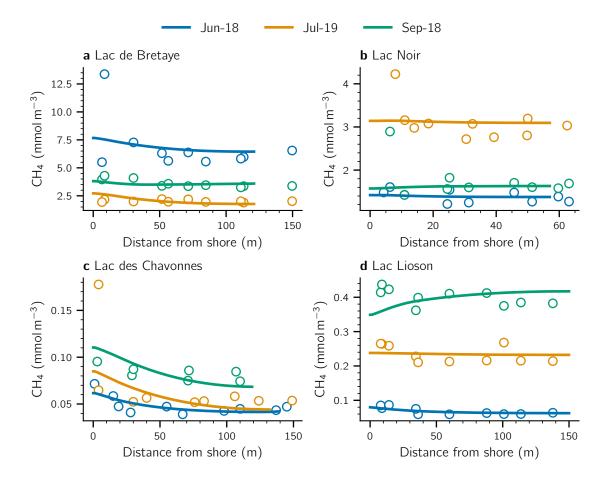


Figure 2.2. Surface CH₄ concentrations along the transects sampled in each lake. Lines represent the CH₄ concentration simulated using the lateral transport model and dots are the measured values. Since the lateral transport model assumes that the CH₄ concentrations in the SML are radially symmetric, the concentrations are shown from shore to center.

2.2.3 Diffusive CH₄ emissions to the atmosphere.

Diffusive CH₄ emissions (F_a) at the air-water interface (AWI) were measured in each lake using a floating chamber (McGinnis *et al.* 2015) at the deepest point of the lake and along the CH₄ concentration transects. Average surface fluxes (mean \pm SD) measured in eutrophic NOI and BRE (3.24 \pm 0.88 mmol m⁻² d⁻¹) were an order of magnitude higher than in LIO and CHA (0.29 \pm 0.43 mmol m⁻² d⁻¹). Surface diffusive fluxes of CH₄ remained relatively similar between sampling dates in each lake (Table 2.2).

Surface diffusive CH_4 fluxes can be estimated using Fick's 1st Law using (Prairie & del Giorgio 2013; McGinnis *et al.* 2015)

$$F_a = k_{\text{CH}_4} \left(C_w - H_{cp} \cdot \text{pCH}_{4,\text{atm}} \right); \quad [\text{mmol/m}^2/\text{d}]$$

 $k_{\text{CH}_4} = k_{600} (600/\text{Sc})^n; \quad [\text{m d}^{-1}]$ (2.1)

where C_w is the CH₄ concentration in the surface water, pCH_{4,atm} is the partial pressure of atmospheric CH₄, H_{cp} is the Henry constant of CH₄ dissolution at in situ temperature, Sc is the Schmidt number for CH₄ and the exponent is taken as n = 2/3 for wind speed $< 3.7 \,\mathrm{m\,s^{-1}}$ and n = 1/2 for wind $> 3.7 \,\mathrm{m\,s^{-1}}$. Several parameterizations have been proposed for the mass transfer coefficient (k_{600}) to estimate diffusive emission to the atmosphere (Klaus & Vachon 2020 and references therein). We compared chamber-based CH₄ mass transfer coefficients (k_{600}^{cb}) based on our chamber flux data to five k_{600} parameterizations: CC98 based on Cole & Caraco 1998; MA10-NP (negative buoyancy), MA10-MB (mixed buoyancy), and MA10-PB (positive buoyancy) based on MacIntyre *et al.* 2010; and VP13 based on Vachon & Prairie 2013 (Fig. S.A.4). These parameterizations weakly correlated with k_{600}^{cb} ($r^2 = [0.01 - 0.037]$; (Fig. S.A.5) and all of them underestimated k_{600}^{cb} (MNB=[16 - 81%]) (Fig. S.A.5). The best agreement was found with MA10-NB based on convective mixing ($r^2=[0.01 - 0.37]$, RMSE=[0.63 - 4.65 m d⁻¹], MNB=[16 - 57%]; Fig. S.A.5).

2.2.4 Diffusive CH₄ fluxes from littoral sediments.

Diffusive CH₄ fluxes at the sediment-water interface (SWI) in the littoral zone (F_s) were estimated using benthic chambers and porewater measurements of dissolved CH₄ (Fig. S.A.6). On average, three cores above the thermocline depth were taken in the epilimnion on September 2018 and

2.2. RESULTS 21

July 2019 (Table S.A.3) to estimate the littoral sediment fluxes at each lake (Methods). Benthic chamber fluxes were calculated based on the temporal change of CH₄ concentration inside the chamber, while Fick's 1st Law was used for the porewater method (Methods). Hypereutrophic BRE showed the highest littoral sediment flux (8.3 \pm 6.7 mmol m⁻² d⁻¹), followed by NOI (eutrophic), CHA (mesotrophic) and LIO (oligotrophic) with the lowest value (0.3 \pm 0.1 mmol m⁻² d⁻¹) (Table 2.2). $\delta^{13}C_{CH_4}$ in the upper part of the sediments ranged between -66 to -48% (Table S.A.4). $\delta^{13}C_{CH_4}$ in the littoral sediment was around 20% lighter than in the surface waters of NOI and BRE but similar for CHA (-60%, Table S.A.4 and Fig. S.A.6). No porewater measurements were performed in LIO due to the rocky nature of the littoral sediments (Methods).

2.2.5 CH₄ ebullition rates and bubble dissolution.

CH₄ ebullition rates at the SWI were estimated using the gas composition of bubbles collected during each sampling campaign, the measured CH₄ fluxes at the SWI (Table S.A.3), and modeling the dissolved porewater gas concentration in the sediments following Langenegger et al. 2019. Bubble dissolution rates (R_{dis}) were calculated as the average dissolution rates in the SML obtained using a discrete bubble model (McGinnis et al. 2006) (see Methods). The spatially-averaged ebullition rates for BRE and NOI were 1.06 and 0.56 mmol m⁻² d⁻¹, respectively, while the bubble dissolution in the SML for BRE and NOI ranged between 17 - 51 μ mol m⁻³ d⁻¹ (Table 2.2). Ebullition was not detected in CHA and LIO.

2.2.6 Vertical diffusive fluxes from/to the epilimnion.

The vertical transport from/to the epiliminion (F_z) is driven by turbulent vertical diffusivity (K_z) and concentration gradients at the bottom of the epilimnion. K_z was determined at each lake for each sampling campaign (Fig. S.A.7) using CTDs and microstructure temperature profiles following the approach from Lorke & Wüest 2002 (Methods). K_z values at the top of the thermocline ranged between 0.03 - 14.4 × 10⁻⁶ m s⁻¹ (Table 2.2). F_z was determined by Fick's 1st Law using K_z and the concentration gradient just below the SML (Table 2.2). In all lakes, F_z was typically low (-0.1 - 0.5 mmol/m²/d), except in BRE and NOI at the end of the summer when fluxes were 13.3 and 3.1 mmol/m²/d, respectively.

2.2.7 Horizontal dispersion.

In the lateral transport model, we estimated the horizontal dispersion coefficient (K_H) for each lake using Peeters & Hofmann 2015 parametrization in which the length scale is equal to the equivalent radius, $R_{\rm eq} = \sqrt{A_a/\pi}$, where A_a is lake surface area as a function of water level. Water level fluctuations were low in BRE, NOI and LOI. ($\pm 1 \,\mathrm{m}$). In CHA, the highest water level was observed at the beginning of summer after ice-off and slowly decreased during the summer by about 4 m (Fig. S.A.9). This water level change was accounted for when estimating horizontal dispersion in CHA on each sampling date. The calculated K_H values were 2034, 903 and 2564 m² d⁻¹ for BRE, NOI, and LIO, respectively, and ranged between 2004 - 2366 m² d⁻¹ for CHA (Table 2.2).

Table 2.2. Inputs for the lateral transport model and full-scale mass balance in the surface mixed layer (mean \pm SD).

Lake	Date	K_H	C_{hyp}	K_z	$\overline{k}_{\mathrm{CH_4}}$	F_s	F_{a}	F_z	R _{dis}
		$(m^2 d^{-1})$	$\mid (\text{mmol m}^{-3}) \mid$	$(10^{-6}\mathrm{ms^{-1}})$	$(m d^{-1})$		$(\mathrm{mmol}\mathrm{m}^{-2}\mathrm{d}$	-1)	$ (\mu mol m^{-3} d^{-1}) $
Bretaye	June 2018 Sept 2018 July 2019	2034	4.0 161.8 2.3	4.09 0.96 0.94	0.67 1.00 2.12	$ \begin{vmatrix} 8.3 \pm 6.7 \\ \text{(n=3)} \end{vmatrix} $	4.6 ± 1.8 3.7 ± 1.5 3.7 ± 1.6	0.5 ± 0.3 13.3 ± 7.7 0.02 ± 0.01	
Noir	June 2018 Sept 2018 July 2019	903	1.3 13.7 2.3	0.91 30.1 0.07	1.75 1.48 0.69	1.5 ± 0.3 (n=4)	2.4 ± 0.8 2.2 ± 1.0 2.9 ± 1.7	0.03 ± 0.02 3.1 ± 1.8 -0.01 ± 0	$ \begin{array}{c cccccccccccccccccccccccccccccccccc$
Chavonnes	June 2018 Sept 2018 July 2019	2366 2004 2246	0.1 0.1 0.4	14.14 0.74 1.02	2.23 1.49 1.12	$\begin{array}{ c c } \hline 0.4 \pm 0.4 \\ (n=3) \\ \hline \end{array}$	0.1 ± 0.02 0.2 ± 0.1 0.1 ± 0.1	-0.1 ± 0.03 0.0 ± 0 0.03 ± 0.02	$ \begin{array}{c c} 0 \pm 0 \\ 0 \pm 0 \\ 0 \pm 0 \end{array} $
Lioson	June 2018 Sept 2018 July 2019	2564	0.1 0.6 0.3	0.89 0.03 4.80	2.22 3.30 1.29	$\begin{array}{ c c } 0.3 \pm 0.1 \\ (n=3) \end{array}$	0.2 ± 0.04 1.2 ± 0.6 0.4 ± 0.2	0 ± 0 0 ± 0 0.01 ± 0.01	$ \begin{array}{c c} 0 \pm 0 \\ 0 \pm 0 \\ 0 \pm 0 \end{array} $

2.2.8 Surface mass balances.

The full-scale mass balance (0-D) proposed by Donis et al. 2017 and a modified version of the lateral transport model (1-D) proposed by Peeters et al. 2019 were used to determine P_{net} in the SML of each lake and campaign based on the input values listed in Table 2.2. P_{net} is the net result of OMP and MOx, which adds and removes CH₄ to the SML, respectively. Thus, when P_{net} is positive the true OMP rate is actually higher than P_{net} .

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Full-scale mass balance.

The full-scale mass balance approach assumes that at each sampling date the surface layer can be modeled as a well-mixed reactor and $P_{net,fs}$ can be estimated as follows:

$$\frac{\partial C}{\partial t} \forall_{\text{SML}} = A_s F_s - A_a F_a + A_z F_z + R_{\text{dis}} \forall_{\text{SML}} + P_{\text{net,fs}} \forall_{\text{SML}}; \quad [\text{mol d}^{-1}]$$
 (2.2)

where C is surface CH₄ concentration, \forall_{SML} is SML volume, and A_s , A_a and A_z are sediment area, lake surface area, and planar area at the bottom of the SML (Table S.A.5), respectively.

Lateral transport model.

Using a modified version of the lateral transport model presented by Peeters et al. 2019, P_{net,lt} rates for each lake were obtained by finding the simulated transect CH₄ concentrations that best-fit to the measured CH₄ concentrations. In this study, the lateral transport model includes vertical diffusive CH₄ flux through the bottom of the SML and bubble dissolution:

$$\frac{\partial C}{\partial t} = K_H \frac{1}{H(r)r} \frac{\partial}{\partial r} \left(H(r)r \frac{\partial C(r)}{\partial r} \right) + \frac{1}{H(r)} K_z \frac{C_{\text{hyp}} - C(r)}{\Delta z} - \frac{\overline{k}_{\text{CH}_4}}{H(r)} \left(C(r) - H_{\text{cpp}} \text{CH}_{4,\text{atm}} \right) + \frac{F_s}{H(r)} + R_{\text{dis}}(r) + P_{\text{net,lt}}; \qquad [\text{mol m}^{-3} \, \text{d}^{-1}]$$
(2.3)

where H(r) is the spatially varying thickness of the SML. The mass transfer coefficient for CH₄ was calculated based on the average gas transfer coefficient obtained from the flux chambers $(\overline{k}_{\text{CH}_4})$, C_{hyp} is the CH₄ concentration 1 m below the bottom of the SML and $\Delta z = 1$ m (Table 2.2; see Methods for each term calculation).

Despite the different modeling approaches and underlying assumptions, we obtained an excellent agreement between the $P_{\rm net}$ rates obtained with both models assuming steady state conditions (Fig. S.A.10, $R^2=0.97$). Monte Carlo simulations were applied to assess uncertainties in both models during the stratified period (Methods). The average $P_{\rm net}$ rates for the three sampling dates were 316, 1434, 12 and 223 μ mol m⁻³ d⁻¹ for BRE, NOI, CHA, and LIO, respectively (Fig. 2.3). $P_{\rm net}$ rates in BRE and NOI were, on average, about seven times higher than CHA and LIO. A decrease of $P_{\rm net}$ rates from the beginning to the end of the

summer were observed in NOI and BRE, whereas in CHA and LIO P_{net} remained relatively consistent across campaigns.

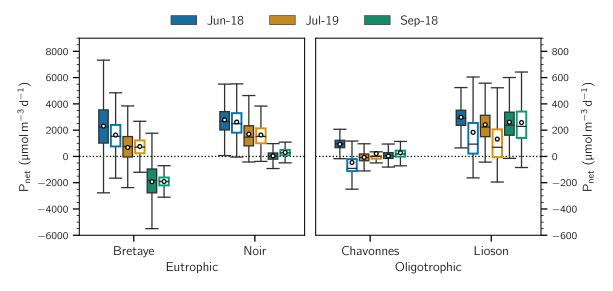


Figure 2.3. $P_{\rm net}$ estimation using the full-scale mass balance ($P_{\rm net,fs}$; filled boxes) and lateral transport model ($P_{\rm net,lt}$; open boxes). Boxes show the first and third quartiles with the median (line), whiskers extend to most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of the $P_{\rm net}$ distribution. Note different scales on Y-axes of the two panels.

2.2.9 Sensitivity analysis of surface diffusive emission to the atmosphere.

Several studies have used k_{600} literature parameterization to estimate F_a (Eq. 2.1, Tan *et al.* 2021 and references there in), however other studies have shown that these estimates often do not correspond with field measurements (Klaus & Vachon 2020 and Fig. S.A.4). Therefore, we analyzed the impact of k_{600} parameterization on P_{net} as it is one of the main parameters affecting the mass balance in the epilimnion.

Since the P_{net} results from both models were similar, we used P_{net} from the full-scale mass balance in the following sensitivity analysis as it is simpler to interpret. Therefore, in the lateral transport model (Eq. 2.3), we simulated surface CH_4 concentrations either with the addition of OMP (i.e., P_{net}), as obtained from the full-scale mass balance approach ($P_{net}=P_{net,fs}$), or without any addition from OMP (i.e., $P_{net}=0$). We also used five different mass transfer coefficient parameterizations (k_{600}) to model diffusive CH_4 emission to the atmosphere in the lateral transport model (Table 2.3). Thus, the resulting surface CH_4 concentrations were obtained from the combinations of P_{net} and k_{600} , as they determined different boundary conditions of the

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mass balance in the SML. The analysis is focused on the best and worst mass transfer coefficient parameterizations (MA10-NB and CC98, respectively) when compared with chamber-based estimations ($\overline{k}_{\text{CH}_4}$) (Figs. S.A.4 and S.A.5). The results of the three remaining parameterization comparisons are available in the Fig. S.A.11 and Table 2.3.

Table 2.3. Combinations of each simulation performed with the lateral transport model for the sensitivity analysis. Root mean square error (RMSE), correlation coefficient (R²) and mean normalized bias (MNB) are shown for the comparison between simulated and measured surface CH₄ concentration. $P_{\text{net,fs}}$ refers to the P_{net} rates obtained from the full-scale mass balance. k_{CH_4} were calculated from the k_{600} literature parameterizations (Eq. 2.1) to be used in (Eq. 2.3). U_{10} , wind speed at 10 m (m s⁻¹); A_s , surface lake area (km²); k_{600} , gas transfer coefficient (cm h⁻¹).

Configuration name P _{net}		$k_{\mathrm{CH_4}}$	RMSE	\mathbb{R}^2	MNB
$P_{\text{net}}0-\overline{k}_{\text{CH}_4}$	0	$\overline{k}_{\mathrm{CH_4}}$	2.44	0.36	-1.60
$P_{\rm net}$ 0-CC98	0	$k_{600} = 2.07 + 0.215U_{10}^{1.7}$ (Cole & Caraco 1998)	4.28	0.35	-0.35
$P_{net}0-MA10-NB$	0	$k_{600} = 2.045U_{10} + 2$ (MacIntyre et al. 2010)	1.80	0.43	-1.27
$P_{net}0$ -MA10-MB	0	$k_{600} = 2.25U_{10} + 0.16$ (MacIntyre et al. 2010)	1.98	0.56	-0.66
$P_{net}0$ -MA10-PB	0	$k_{600} = 1.75U_{10} - 0.15$ (MacIntyre <i>et al.</i> 2010)	3.13	0.61	0.11
$P_{\rm net}0$ -VP13	0	$k_{600} = 2.51 + 1.48U_{10} + 0.39U_{10}\log_{10}(A_s)$ (Vachon & Prairie 2013)	2.54	0.36	-0.93
$\overline{\mathrm{P}_{\mathrm{net}}}$ - $\overline{k}_{\mathrm{CH_4}}$	P _{net,fs}	$\overline{k}_{\mathrm{CH_4}}$	0.56	0.96	0.07
P_{net} -CC98	$P_{\text{net,fs}}$	$k_{600} = 2.07 + 0.215U_{10}^{1.7}$ (Cole & Caraco 1998)	2.62	0.89	1.75
P_{net} -MA10-NB	$P_{\text{net,fs}}$	$k_{600} = 2.04U_{10} + 2 \text{ (MacIntyre } et \ al. \ 2010)$	0.72	0.88	0.71
P_{net} -MA10-MB	$P_{\text{net,fs}}$	$k_{600} = 2.25U_{10} + 0.16$ (MacIntyre et al. 2010)	2.08	0.84	1.01
P_{net} -MA10-PB	$P_{\text{net,fs}}$	$k_{600} = 1.74U_{10} - 0.15$ (MacIntyre <i>et al.</i> 2010)	4.23	0.83	1.57
P _{net} -VP13	$P_{\rm net,fs}$	$k_{600} = 2.51 + 1.48U_{10} + 0.39U_{10}\log_{10}(A_s)$ (Vachon & Prairie 2013)	1.06	0.9	0.72

The best agreement between measured and simulated CH₄ concentrations was found using P_{net} from the full-scale mass balance ($P_{net,fs}$) and the chamber-based mass transfer coefficient (\bar{k}_{CH_4}) (P_{net} - \bar{k}_{CH_4} , Table 2.3, Fig. 2.4b). When using \bar{k}_{CH_4} with P_{net} set to zero ($P_{net}0$ - \bar{k}_{CH_4}), average CH₄ concentrations along the transect were underestimated relative to the measured values (MNB=-1.60, Table 2.3, Fig. 2.4a). Using $P_{net,fs}$ and MA10-NB or CC98 parameterizations (P_{net} -MA10-NB and P_{net} -CC98) resulted in an overestimation of CH₄ concentrations (Figs. 2.4d and 2.4f and Table 2.3), whereas when P_{net} was set to zero ($P_{net}0$ -MA10-NB and $P_{net}0$ -CC98) using both parameterizations, the average CH₄ concentrations along the transect were underestimated (Table 2.3, Figs. 2.4c and 2.4e). The lowest correlation (R^2 =0.35) and highest error (RMSE=4.28) of the all the combinations was obtained when P_{net} was set to zero and CC98 was used to estimate the CH₄ diffusive flux to the atmosphere (Table 2.3).

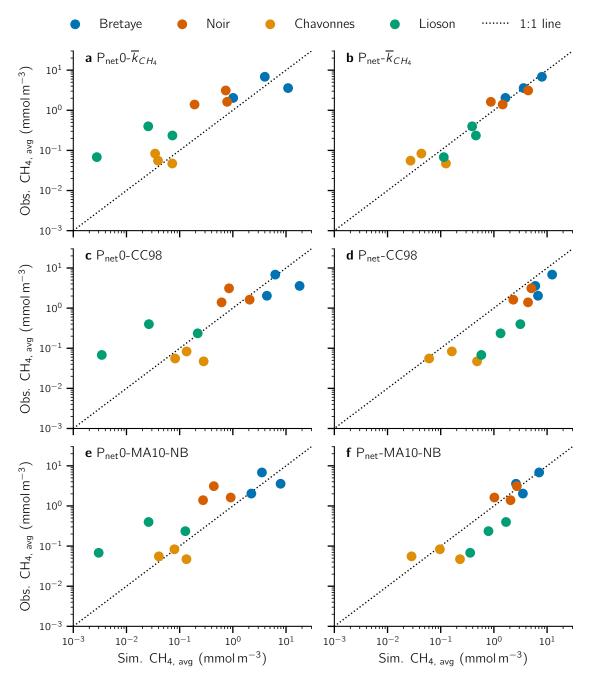


Figure 2.4. Comparison of observed and simulated average surface CH₄ concentration along transects for each campaign. Simulated CH₄ concentrations were obtained with the lateral transport model using k_{600} for diffusive emissions either with (panels **b**, **d** and **f**) or without P_{net} (panels **a**, **c** and **e**). The k_{600} was either the chamber-based k_{600} ($\overline{k}_{\text{CH}_4}$, panels **a** and **b**), Cole & Caraco 1998 (CC98, panels **c** and **d**) and MacIntyre *et al.* 2010 negative buoyancy (MA10-NB, panels **e** and **f**). The statistical results of each panel are shown in Table 2.3.

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2.2.10 Contribution of methane sources to atmospheric diffusive emissions.

As P_{net} rates were similar from both models, we again use the P_{net} rates from the full-scale mass balance approach to calculate the contribution of each CH₄ SML source to the diffusive atmospheric CH₄ emissions from each lake and during each campaign (Methods). F_s and P_{net} were the two major sources of CH₄ in the SML. On average, P_{net} contributed about 30% of the CH₄ emissions in BRE and CHA, while it reached up to 60% and 90% for NOI and LIO, respectively (Fig. 2.5). P_{net} was a dominant source in all lakes in June and July except for CHA in July. Negligible P_{net} contributions (<8%) were found in all lakes in September 2019, except for LIO (91%). On average, F_s contributed about 10%, 30%, 50% and 65% to the CH₄ emissions in LIO, NOI, BRE and CHA, respectively. For CHA and NOI, the F_s contribution increased at the end of the summer and reached up to 90% for CHA in September. For BRE and LIO, the F_s contribution was relatively constant during the different months. On average, F_s contributed the same in the oligotrophic and eutrophic lakes. The vertical turbulent flux (F_z) contributed about 50% of the atmospheric CH₄ from BRE and NOI in September and about 30% from CHA in July, but was negligible (<9%) for the other campaigns. The contribution from bubble dissolution (R_{dis}) in BRE and NOI was also negligible (<4%).

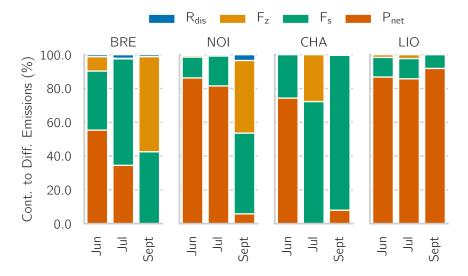


Figure 2.5. Contribution to diffusive atmospheric CH_4 emissions (F_a) from the sediment flux (F_s) , diffusive flux from hypolimnion (F_z) , bubble dissolution (R_{dis}) and net production (P_{net}) in the SML of Lac de Bretaye (BRE), Lac Noir (NOI), Lac des Chavonnes (CHA) and Lac Lioson (LIO).

2.3 Discussion

In most of our study lakes during all campaigns, the $P_{\rm net}$ values were positive, indicating that the oxic SML acts as a CH₄ source during the stratified season (Fig. 2.3). $P_{\rm net}$ was around zero in CHA, which is the meso-oligotrophic lake with the largest water level changes throughout the summer, in contrast to the other pre-alpine lakes in our study that maintained relatively consistent water levels. The observed average $P_{\rm net}$ rates were within the range of values previously reported (Günthel *et al.* 2021), except for NOI with the highest reported $P_{\rm net}$ (2308 \pm 2024 μ mol m⁻³ d⁻¹).

P_{net} rates were temporally variable and across study sites. At the beginning of the summer, highly positive P_{net} rates indicated that OMP was an active source of CH₄ to the atmosphere. By the end of the stratified season, P_{net} became negative indicating that MOx was dominating in surface waters of each lake except for the oligotrophic LIO (Fig. 2.3). This seasonal trend in OMP was also observed by Günthel et al. 2019 and may be related to the CH₄ production rates of different algal species (Günthel et al. 2020) and their concentrations at the end of the growing season. In addition, the eutrophic lakes BRE and NOI had P_{net} rates one order magnitude higher than the meso-oligotrophic and oligotrophic lakes (CHA, LIO), suggesting that OMP may also be related to trophic state. From this perspective, productive lakes in general may experience higher OMP rates than less productive ones.

The OMP contribution to epilimnetic CH_4 budgets and diffusive CH_4 emissions to the atmosphere is hotly debated (Günthel *et al.* 2021; Peeters & Hofmann 2021). In our study, P_{net} is a significant contributor to atmospheric diffusive CH_4 emissions. Its contribution to emissions decreased significantly by the end of the summer in the eutrophic lakes, as did P_{net} rates themselves (Figs. 2.3 and 2.5). Although higher P_{net} rates were found in the eutrophic lakes, the P_{net} contribution fraction to surface diffusive CH_4 emissions was independent of the trophic status of the lake. For example, the fraction of P_{net} contribution to emissions was similar and even higher in oligotrophic LIO than that in eutrophic NOI. This was mainly due to the substantial contribution of CH_4 from the littoral sediments of the eutrophic lakes to the SML. Therefore, increasing CH_4 production and emission from littoral sediments will actually reduce the fractional contribution of OMP to atmospheric emissions even though OMP rates may be

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higher in such productive systems.

As a first attempt at upscaling OMP, Günthel et al. 2019 proposed that the OMP contribution to diffusive CH_4 emissions from lakes can be estimated as a function of littoral sediment area and SML volume. In our study, the P_{net} contribution to diffusive CH_4 flux to the atmosphere at the AWI was highly variable and disagreed with this simple upscaling approach (Fig. S.A.14). While it is plausible that OMP contribution may partially depend on lake bathymetry (i.e., the fraction between the sediment area and the SML volume), our results indicate that OMP is a complex phenomenon more likely to be related to the lake trophic properties (e.g. productivity).

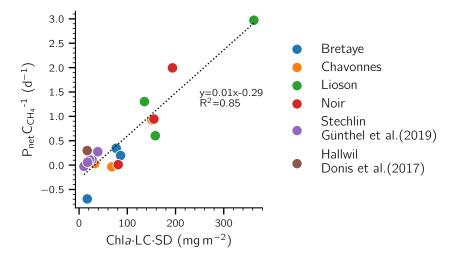


Figure 2.6. Interaction between $P_{\rm net}$ (mmol m⁻³ d⁻¹) and Chla (mg m⁻³), LC (-) and SD (m) suggest a direct role of photosynthesis on OMP. Specific production/oxidation rate calculated as $P_{\rm net}$ normalized by CH₄ concentration (mmol m⁻³) ($P_{\rm net}C_{\rm CH_4}^{-1}$) versus Chla × light climate (LC = $2.5\frac{\rm SD}{H_{\rm SML}}$) × SD. $C_{\rm CH_4}$ is the average surface concentrations, Chla is the surface average concentrations obtained from CTD's profiles at the center of the lake, SD is the Secchi depth and $H_{\rm SML}$ is the SML depth. All the parameters were calculated at each sampling campaign.

Two primary mechanisms have been the focal points of CH_4 production in oxic waters. Methylphosphonate (MPn) biodegradation has been shown to be responsible for CH_4 production in oxic waters of the ocean (Karl et al. 2008) and lakes (Wang et al. 2017), specifically in phosphorus-limited environments (Karl et al. 2008; Wang et al. 2017). In our pre-alpine lakes, however, we did not observe any correlation between P_{net} and phosphorus in the SML (Fig. S.A.12). The other suggested OMP mechanism is the production of CH_4 in nitrogen-limited environments via the transformation of CO_2 , nitrogen gas, and hydrogen by the nitrogenase

enzyme (Zheng et al. 2018; Luxem et al. 2020) that is commonly present in cyanobacteria. While we observed that P_{net} negatively correlated with dissolved inorganic nitrogen (DIN) (Fig. S.A.13), which could indicate the use of nitrogen for OMP, to our knowledge CH₄ production due to nitrogenase activity in cyanobacteria has not yet been observed. Ultimately, although our data suggests links between OMP and trophic parameters, similar to relationships found in Günthel et al. 2020, the underlying OMP mechanisms still need to be deciphered.

While it is premature to construct a mechanistic model to estimate OMP in lakes, based on our data we suggest an empirical approach using simple correlations between physical and biochemical parameters as a tool to identify the potential for the occurrence of P_{net} in lakes (Fig. 2.6). Higher surface CH_4 concentrations were observed in eutrophic lakes (Table 2.1); therefore, CH₄ concentration in the SML is used as a proxy to reflect the trophic state of each lake and to normalize P_{net} rates found in the eutrophic and oligotrophic lakes (Fig. 2.3). Although the mechanism is still unclear, recent evidence indicates that OMP could be a photosynthesis-derived process (Bižić et al. 2020; Günthel et al. 2020; Hartmann et al. 2020). In this approach, Chla is used as a proxy of phytoplankton biomass, Secchi depth as light availability for photosynthesis and MOx inhibition (Thottathil et al. 2019) and LC (light climate) defines the variability of light conditions that phytoplankton can be subjected to in the SML during the day (MacIntyre 1993). This interaction between $P_{\rm net}$ normalized by the SML CH₄ concentration versus Chl $a \times LC \times CH_4$ Secchi depth indicates the direct role of phytoplankton and light availability in CH₄ production under oxic conditions (Bižić et al. 2020; Günthel et al. 2020; Hartmann et al. 2020). Including the data from Donis et al. 2017 and Günthel et al. 2019, this parameterization explains around 85% of the dataset ($R^2=0.85$). While more data are needed to understand this parameterization, it provides a step towards estimating P_{net} in the SML that would help to identify OMP dynamics across systems, identify lakes with potentially high OMP rates, and develop a global upscaling of OMP (or P_{net}) driven emissions.

The methodologies for determining P_{net} are limited by the accuracy of the boundary conditions of the mass balance (i.e., diffusive CH_4 emissions at the AWI, CH_4 flux from littoral sediment, ebullition, etc.). These boundary conditions are sometimes based on very few measurement locations using uncertain methodologies and are naturally variable. The variability and uncertainty of such estimations leads to the observed range of P_{net} in mass

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balance approaches obtained with the Monte Carlo simulations (Fig. 2.3). Therefore, to assess the robustness and the validity of the models used, we compared the boundary conditions components (F_a and F_s) with literature values and examined how their variability may alter the outcome of the two mass balance models.

Diffusive CH₄ emissions to the atmosphere are temporally and spatially variable. We accounted for the spatial variability by using the average of ten surface flux measurements along a lake-wide transect for each P_{net} calculation. The impact of diffusion variability at the AWI on P_{net} rates were considered in the Monte Carlo simulations conducted on both modeling approaches. In addition, the average diffusive CH₄ emissions estimated for lakes NOI, CHA and BRE are well within the range reported for the stratified season of these lakes in previous studies (0.06 - 4.38 mmol/m²/d; Rinta et al. 2017). There are no previous data for LIO.

A large uncertainty in the estimation of diffusive CH_4 emissions is the mass transfer coefficient (k_{600}) parameterization. Therefore, we applied five alternative k_{600} parameterizations (Cole & Caraco 1998; MacIntyre et al. 2010; Vachon & Prairie 2013, negative, mixed and positive buoyancy) to estimate CH_4 diffusion at the AWI in the four pre-alpine lakes and compared these fluxes with direct measurements using floating chambers. The comparison of $k_{600}^{\operatorname{cb}}$ with all the tested parameterizations resulted in a low correlation ($\operatorname{R}^2<0.38$) and clear underestimation of the measured k_{600} values (Fig. S.A.4), reflecting the limitations of the k_{600} models across different lakes (Klaus & Vachon 2020). The underestimation by k_{600} parameterizations has also been reported in previous studies (Tan et al. 2021 and references there in). We hypothesize that the presence of oxygen microbubbles produced by photosynthesis in the water column (Koschorreck et al. 2017) might enhance the mass transfer coefficient (McGinnis et al. 2015). This phenomenon would be more relevant in high altitudes lakes due to the lower air pressure and oxygen saturation concentration.

In our analysis of the k_{600} parameterizations for the lateral transport model, we observed that when using the CC98 parameterization for surface CH₄ fluxes, OMP was not required to close the mass balance. Similarly, OMP seemed irrelevant when Peeters *et al.* 2019 re-analyzed the data from Donis *et al.* 2017 using the CC98 k_{600} parameterization instead of the measured surface CH₄ fluxes Donis *et al.* 2017. Moreover, when comparing observed surface

CH₄ concentrations with simulated CH₄ concentrations, we obtained a lower correlation and higher error using simulated CH₄ concentrations based on the CC98 parameterization than those based on $k_{600}^{\rm cb}$ or MA-NB parameterizations (Fig. 2.4). This is largely explained by the fact that the CC98 parameterization grossly underestimates $k_{600}^{\rm cb}$ for all lakes (MNB = [57 - 71 %]) and does not correlate well with $k_{600}^{\rm cb}$ (R²=[0.01 - 0.38], Fig. S.A.5). Thus, a generalized k_{600} parameterization from a single lake (e.g. CC98) does not apply to all lakes. It is therefore much more reliable to use in situ measured fluxes to compute k_{600} in a mass balance.

The littoral diffusive sediment fluxes we measured were within the range of values reported in the literature (0.001 - 8.8 mmol m⁻² d⁻¹ (Huttunen et al. 2006; Bastviken et al. 2008; Peeters et al. 2019)). In the full-scale mass balance, we assumed that the lateral flux to the center of the lakes was equal to the diffusive CH_4 flux coming from the sediment in the littoral area. It has been shown that CH_4 production rates in sediments increase with increasing temperature (Bastviken et al. 2008); thus, it has been hypothesized that sediment CH_4 diffusion will also follow this relationship (Peeters et al. 2019). As most of our sediment flux measurements in the littoral zone were performed in July when the temperatures were highest in all lakes (Table S.A.3), we can assume that those observed sediment fluxes were on the higher end of possible values. Ultimately, using a presumably high sediment flux from July for the mass balance of other months would result in a conservative $P_{\rm net}$ estimate.

We also investigated the impact of littoral sediment flux on mass balances. Assuming that OMP does not occur (i.e. $P_{net}=0$) in the full-scale mass balance, we would need littoral sediment fluxes that are two to three times higher than our measured fluxes to compensate for the diffusive CH₄ emissions measured at the AWI (Fig. 2.7). In oligotrophic LIO, we would need a littoral sediment flux of about $2.23 \pm 1.12 \text{ mmol m}^{-2} \text{d}^{-1}$ which is very unlikely given that flux is an order of magnitude higher than what we measured (0.3 \pm 0.1 mmol m⁻² d⁻¹). In fact, in BRE we have measured one of the highest littoral sediment fluxes yet reported (8.3 \pm 6.7 mmol m⁻² d⁻¹) (Huttunen *et al.* 2006; Bastviken *et al.* 2008; Hardenbroek *et al.* 2012; Peeters *et al.* 2019), and we still required one of the highest P_{net} rates ever reported in the literature (June 2018: 2314 \pm 2046 µmol m⁻³ d⁻¹) to close the CH₄ budget in that lake. Therefore, littoral sediment CH₄ flux alone cannot account for the diffusive CH₄ emissions measured in our lakes and OMP needs to be included to close the CH₄ budget.

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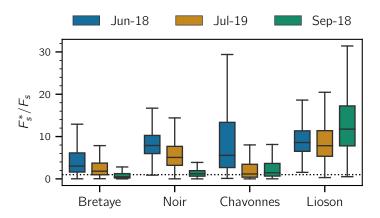


Figure 2.7. Proportion of littoral sediment flux needed (F_s^*) to compensate the surface diffusive emissions compare to measured littoral sediment flux (F_s) considering no OMP in the surface mixed layer.

In this study, we quantified the P_{net} rates of CH₄ (i.e. net balance between OMP and MOx) in the oxic SML of four pre-alpine lakes using two models that have previously produced contradictory results when resolving OMP in lowland lakes (Donis et al. 2017; Günthel et al. 2019; Peeters et al. 2019; Günthel et al. 2021; Peeters & Hofmann 2021). The good agreement between both approaches we found shows that there are no methodological issues with the models themselves when the appropriate boundary conditions are used to estimate OMP (or P_{net}, in our case). We also conducted thorough sensitivity analyses on the two main parameters that lead to the highest uncertainties. This analysis shows that measured surface fluxes have to be used instead on literature k_{600} parameterizations to estimate the diffusive CH₄ flux to the atmosphere and an unlikely high CH₄ diffusive flux from littoral sediment are need it to close the mass balance in the SML without including $P_{\rm net}$. Moreover, our results indicate that in three out of four lakes a positive $P_{\rm net}$ (i.e. a net input of CH_4 from OMP) needs to be included in the SML CH₄ budget. In fact, we show that up to 85% of atmospheric CH₄ emissions from these lakes at the beginning of summer results from OMP, and even in our systems with some of the highest recorded littoral sediment fluxes, we still obtained some of the highest epilimnetic P_{net} (or OMP) rates ever reported.

Finally, while the mechanisms behind OMP need further investigation, this study (in agreement with previous ones) show that nitrogen, light, and photoautotrophs may play a significant role in OMP. Consequently, future changes in light availability and temperature may induce positive feedbacks by promoting algal species capable of producing the greenhouse gas CH₄ in oxic environments at the AWI where it can easily be emitted into the atmosphere. Although the contribution of OMP to total diffusive emissions from inland waters is hotly debated, we have shown that it can be a dominant source from lakes in the pre-alpine region that is already experiencing climatic changes at higher rates than average. It is thus crucial to continue quantifying the contribution of OMP from various aquatic systems and identify the main drivers of OMP to better understand the impact of OMP on the global CH₄ cycle and how to predict or even mitigate its impact in a changing climate.

2.4 Methods

2.4.1 Study sites.

Lac de Bretaye (BRE), Lac Noir (NOI), Lac des Chavonnes (CHA) and Lac Lioson (LIO) are pre-alpine lakes (above 1600 m.a.s.l) located in Canton Vaud, Switzerland (Table S.A.1). All lakes are of glacial origin and have a wide-range of trophic states (oligotrophic-hypereutrophic). BRE, NOI and CHA are $\sim 500\,\mathrm{m}$ away from each other, while LIO is located about 7 km away from the others . BRE and NOI are small and shallow lakes without inflow or outflow streams located in alpine meadows which are utilized for animal grazing (Lods-Crozet & Reymond 2008). CHA has a small inflow stream while LIO has a small creek outflow that is the origin of the Hongrin River.

2.4.2 Limnological measurements.

At each sampling date, water column profiles were measured at the deepest point of each lake (M1, Fig. S.A.1) with a CTD profiler (Conductivity-Temperature-Depth, Seabird SBE19plus) equipped with temperature, conductivity, oxygen, PAR, turbidity, Chl-a and pH sensors.

Total (TP) and dissolved phosphorus (DP), dissolved inorganic nitrogen as nitrate plus nitrite (DIN), dissolved silica (DSIL) and total carbon concentration (TC) were measured at each campaign in the upper mixed layer (from the surface to the bottom of the thermocline) and in the hypolimnion (Table S.A.2). Water samples were collected with a Niskin sampler and equal amounts of water were transferred into two 1L glass bottle (Duran, GmbH, Mainz, Germany).

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50 mL of water were filtered through 0.45 μm (PES) Syringe filters to measure dissolved nutrient fractions. An AQ2 Discrete Analyzer (SEAL Analytical) based on spectophotometric methods was used to measure TP and DP by Acidic molybdate/antimony with ascorbic acid reduction (USEPA 1993), Nitrate-N plus Nitrite-N by Cadmium coil reduction followed by sulfanilamide reaction in the presence of N-(1-naphthylethylenediamine) (USEPA 1993) and DSIL by Acidic molybdate with ANSA reduction (USEPA 1983). A Shimadzu carbon analyzer (TOC-L_{CPH/CPN}) measured TC.

2.4.3 Mass balance.

The mass balance of CH₄ in the SML was calculated using two models (Eqs. 2.2 and 2.3). A sonar survey was performed to obtain the bathymetry of each lake (Fig. S.A.1). The bottom of the SML ($H_{\rm SML}$) was define when $\partial T/\partial z$ becomes smaller than $-1\,^{\circ}{\rm C}\,{\rm m}^{-1}$ (Read *et al.* 2011) (Table 2.1). The SML water volume ($\forall_{\rm SML}$), the surface area (A_a), the sediment area (A_s) and the planar area at the bottom of the SML (A_z) were determined using the software Surfer® (Golden Software, LCC) (Table S.A.5).

2.4.4 Full-scale mass balance.

The net CH₄ production (P_{net}) was estimated using a mass balance in the surface mixed layer (Eq. 2.2) assuming steady state conditions ($\frac{\partial C}{\partial t} \forall = 0$). We assumed that the lateral contribution to the mass balance is equal to the littoral sediment flux times the area of the sediment. The spatial average values for the surface fluxes (F_a), bubble dissolution rates (R_{dis}) in the SML and hypolimnetic fluxes (F_z) were used as boundaries conditions (Table 2.2).

2.4.5 Lateral transport model.

The lateral transport model simulates the transect CH₄ concentrations in the SML similar to Peeters et al. 2019 but including the additional flux terms from bubble dissolution and diffusion across the thermocline (Eq. 2.3). This model considers that the surface layer is fully mixed in the vertical and, therefore, the vertical CH₄ concentrations are homogeneous within the SML.

In the simulations of each lake, we assumed that the SML, sources and sinks are radially symmetric in the horizontal. Therefore, the development of CH₄ concentration can be described

based on the radial distance r from the shore to the center of the lake $(r_{\text{max}} = \sqrt{A_a/\pi})$.

Two regions were defined in the model, the littoral zone $(r \leq r_s = \sqrt{(A_a - A_s)/\pi})$ and the pelagic waters $(r > r_s)$. The SML thickness (H(r)) is equal to the mixed layer depth in the pelagic region and, within the littoral zone, H(r) decreases linearly with r from the mixed layer depth to zero at the shore. The littoral sediment flux is zero in the pelagic zone $(r < r_s)$ and equal to the measured average littoral sediment flux $(\overline{\mathbb{F}_s})$ in the shallow region $(r \geq r_s)$ as:

$$F_s(r,t) = \begin{cases} \overline{F_s} & \text{for } r \ge r_s \\ 0 & \text{for } r < r_s \end{cases}$$
 [mmol/m²/d] (2.4)

The vertical turbulent transport (F_z) is zero in the littoral areas and in the pelagic waters F_z is calculated as:

$$F_z = K_z \frac{C_{\text{hyp}} - C(r)}{\Delta z} \tag{2.5}$$

where K_z is the turbulent vertical diffusivity, $C_{\rm hyp}$ is the CH₄ concentration 1 m below the bottom of the SML and $\Delta z = 1$ m. Average bubble dissolution rates $(R_{\rm dis}(h(r)))$ as a function of lake depth (h) were included in the SML. At the boundaries, horizontal fluxes were assumed as zero. To estimate the horizontal dispersion coefficient (K_H) we used Peeters & Hofmann 2015 parameterization:

$$K_H = 1.4 \times 10^{-4} L^{1.07} \qquad [\text{m}^2 \,\text{s}^{-1}]$$
 (2.6)

where the length scale L [m] was calculated as $L = r_s$ (Table S.A.5). Eq. 2.6 is the average of the results 1, 3 and 4 found on Table 2 of Peeters & Hofmann 2015.

 P_{net} rates were obtained using least square method optimization solver implemented with the *curve fit* function from Scipy (Virtanen *et al.* 2020) in Python.

2.4.6 Monte Carlo Simulation

To assess uncertainties, Monte Carlo simulation were performed (10000 iterations) when solving the full-scale mass balance model. P_{net} , R_{dis} and F_z were selected within a normal distribution resulting from the mean (μ) and their standard deviation (σ) retrieved from the field measure-

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ments. To prevent negative values, F_a and F_s were chosen from a gamma distribution defined by shape $(\kappa = \mu^2/\sigma^2)$ and the scale $(\theta^2 = \sigma^2/\mu)$. Here the gamma distribution has the density $f(x) = \left(x^{\kappa-1} \frac{e^{-x/\theta}}{\theta^{\kappa}\Gamma}\right)$ where Γ is the gamma function. Random.normal and random.gamma functions from the Numpy package (Harris *et al.* 2020) in Python were used for each normal and gamma distributions, respectively.

2.4.7 Water column CH_4 and $\delta^{13}C_{CH_4}$ signature.

At each sampling campaign CH_4 and $\delta^{13}C_{CH_4}$ concentration profiles were taken at the deepest location of each lake (M1, Fig. S.A.1). A transect composed of 10 or 11 stations across the lake (shore to shore) was also performed at each lake (T1-T11, Fig. S.A.1) were surface CH_4 and $\delta^{13}C_{CH_4}$ concentrations and surface CH_4 fluxes were measured.

Dissolved CH₄ concentration profiles were performed at a maximum depth resolution of 0.5 m where the metalimnetic CH₄ gradient was expected. For the profile, the water samples were obtained with a 5-L Niskin bottle and then gently transferred into a 1 L glass bottle (Duran GmbH, Mainz, Germany) while for the transect the samples were obtained directly with a 1 L glass bottle (Duran GmbH, Mainz, Germany). For both methodologies, the water was overflowing to replace the volume three times. CH₄ concentrations and $\delta^{13}C_{CH_4}$ were measured using the headspace method following Donis *et al.* 2017. The samples were measured on a Cavity Ring-Down Spectrometer analyzer (Picarro G220-i, Santa Clara, CA, USA) for CH₄ concentrations in the gas phase (ppm) and stable isotope ration ($\delta^{13}C_{CH_4}$ in %). Water CH₄ concentrations were back calculated according to Wiesenburg & Guinasso 1979 accounting for water temperature, air concentration and the headspace/water ratio in the bottle.

2.4.8 CH_4 diffusive fluxes to the atmosphere.

Diffusive CH₄ emissions to the atmosphere (F_a) were measured using a floating chamber attached to a portable GHG analyzer (UGGA; Los Gatos Research, Inc.). Instrument-specific precision at ambient concentrations (1- σ of 100 s average) for [12 CH₄] is 0.25 ppb. The floating chamber consist of an inverted plastic container with foam elements for floatation (as McGinnis *et al.* 2015). To minimize artificial turbulence effects, the buoyancy element was adjusted that only \sim 2 cm of the chamber penetrated below the water level. The chamber was painted white to

minimize heating. Two gas ports (inflow and outflow) were installed at the top of the chamber via two 5 m gas-impermeable tubes (Tygon 2375) and connected to the GHG analyzer measuring the gaseous CH₄ concentrations in the chamber every 1 s. Transects were performed with the chamber deployed from a boat. The chamber was allowed to freely drift to minimize artificial disturbance. Fluxes were obtained by the slopes of the resolved CH₄ curves over the first \sim 5 min when the slopes were approximately linear (R² > 0.97).

To simulate the fluxes to the atmosphere in the lateral transport model, chamber-based mass transfer coefficient $(k_{\text{CH}_4}^{\text{cb}})$ was calculated as:

$$k_{\text{CH}_4}^{\text{cb}} = \frac{F_a}{\left(\text{CH}_{4,w} - H_{\text{cp}} \cdot \text{pCH}_{4,\text{atm}}\right)}; \quad [\text{mmol m}^{-2} \,\text{d}^{-1}]$$
 (2.7)

where pCH_{4,atm} and CH_{4,w} are the CH₄ atmospheric partial pressures and water concentration, respectively. H_{cp} is the Henry constant of CH₄ dissolution at in situ temperature obtained from Sander 2015.

2.4.9 Sediment sampling.

Littoral sediment cores were taken in most of the lakes, except for LIO where the rocky bottom made it impossible to take a sample. Sampling was performed with a gravity sediment corer (Uwitech, Mondsee, Austria) equipped with an acrylic liner of 70 cm in length and with an internal diameter of 6 cm. The liner had pre-drilled holes to fit 3 mL syringes at 1 cm intervals. Additionally, in-site core liners where hammered into shallow littoral sediment where benthic chambers were performed. The location and depth of each core are shown in Fig. S.A.1 and Table S.A.3.

2.4.10 Porewater CH_4 concentration and $\delta^{13}C_{CH_4}$ signature.

3 mL of sediment was sub-sampled at 1 - 2 cm depth intervals with headless 3 mL syringes through the pre-drilled holes from the selected depths. The sediment sub-sample was immediately placed into 1 L glass bottle (Duran GmbH, Mainz, Germany) containing 500 mL of lake water previously bubbled with air to reach equilibrium with the atmosphere. The subsequent procedure followed the same as for the water column head-space method. Porewater CH₄ concentrations were back calculated from the headspace concentrations accounting for dilution of sediment porewater

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in the lake water (assuming that aerated lake water is in equilibrium with the atmosphere), temperature, headspace ratio, and assuming a porosity of 0.9.

2.4.11 Methane benthic fluxes.

The littoral CH_4 sediment flux (F_s) at each lake was determined as the average flux provided by two independent methods which are described below.

Porewater method.

Methane fluxes at the sediment–water interface were calculated with Fick's 1st Law over the linear top 2 - 3 cm of the porewater concentration profile.

$$F_s = -\phi D_{\text{CH}_4} \theta^{-2} \frac{\partial C}{\partial z}; \qquad [\text{mmol m}^{-2} \, \text{d}^{-1}]$$
 (2.8)

where F_s is the diffusive CH₄ flux at the sediment–water interface, ϕ the porosity of the sediments (assumed as 0.9), D_{CH_4} the diffusion coefficient for CH₄ in water (1.5 × 10⁻⁵ cm² s⁻¹ (Broecker & Peng 1974), θ^2 the square of tortuosity (1.2) (Boudreau 1997) and $\partial C/\partial z$ the measured vertical concentration gradient.

Benthic chamber.

Benthic chambers were deployed in situ or by sediment cores retrieved from the littoral sediment of the lake. The core was covered leaving $\sim 5 \,\mathrm{cm}$ of headspace and ~ 30 - $50 \,\mathrm{cm}$ of water. The lid was connected to a GHG analyzer creating a closed loop where partial pressure of CH_4 (P_{CH_4}) was measured over time. At the beginning and at the end of the experiment, water CH_4 concentration (C_w) was also measured. Each experiment deployment lasted about 1 h while the surface water was gently stirred to increase the mass transfer coefficient (k_{bc}) at the air-water interface without producing sediment resuspension. The sediment flux was calculated using three methods:

 \bullet Integrated mass balance: F_s is obtained using the beginning and final air and gas CH_4

concentration and performing a mass balance in the water and the air phase as:

$$F_s A_{bc} = \frac{V_{air}}{RT_a} \frac{\Delta P_{CH_4}}{\Delta t} + \frac{V_w \Delta C_w}{\Delta t}; \quad [mmol d^{-1}]$$
 (2.9)

where V_w and V_{air} are the volume of the water and air phases, respectively. R is the ideal gas constant, T_a is the air temperature and A_{bc} is the surface area of the chamber.

• Transient mass balance: solving the mass balance over time we obtain that:

$$\frac{\partial P_{\text{CH}_4}}{\partial t} = \frac{RT_a}{b} \left(wF_s - (wF_s - bk_{\text{cb}}C_o) e^{-bk_{\text{bc}}t} \right); \qquad [\text{Pa d}^{-1}]$$
 (2.10)

where $w = A_{\rm bc}/V_w$, $a = A_{\rm bc}/V_a$, $C_0 = C_w(0) - H_{\rm cp}P_{\rm CH_4}$ and $b = (w - H_{\rm cp}RT_aa)$. The sediment flux is estimated fitting k_{cb} and F_s to the measured $\partial P_{\rm CH_4}/\partial t$ using least square method optimization solver implemented on the *curve fit* function from Scipy (Virtanen *et al.* 2020) in Python. The k_{cb} boundaries were set from 0 - 40 m d⁻¹ for the fitting.

• Equilibrium mass balance: after ~ 1 h of measurements, we assume that the exponential part of the curve of Eq. 2.10 becomes negligible. Therefore, F_s can be estimated with the last 5 min of the CH₄ partial pressure as:

$$P_{\text{CH}_4} = \frac{RT_a}{b} w F_s t; \qquad [\text{Pa}]$$
 (2.11)

The flux from the benthic chamber was calculated as the average of the results of the three methods described above.

2.4.12 CH₄ bubble dissolution and ebullition rates.

The CH₄ dissolution from a single bubble released from the sediment was calculated using McGinnis et al. 2006. For each bubble we considered a diameter of 5 mm and the water column CH₄, CO₂ and O₂ concentrations and temperature profiles at each date. The initial bubble composition at each depth was estimated from a linear interpolation from bubble content obtained following the same methodology as Langenegger et al. 2019. The total bubble dissolution rate ($R_{dis}(z)$) was calculated considering the contribution from all bubbles released below that depth

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as:

$$R_{\text{dis}}(z) = \frac{\sum_{\text{bottom}}^{z} r_i \frac{F_{\text{eb,SWI},i}}{n_{0,i}} \Delta A_{\text{sed},i}}{A_p(z)} \qquad [\mu \text{mol m}^{-3} \, \text{d}^{-1}]$$

$$(2.12)$$

where r_i is the bubble dissolution from an individual bubble at depth i (µmol bub⁻¹), $F_{\text{eb,SWI},i}$ is the CH₄ ebullition flux released at the sediment-water interface (SWI) at depth i (mmol/m²/d) and $n_{0,i}$ is the initial amount of CH₄ in a single bubble (µmol bub⁻¹). $\Delta A_{\text{sed},i}$ is the sediment area between the depth interval i to i+1 (m²). $F_{\text{eb,SWI},i}$ was estimated using Langenegger et al. 2019's model. This model assumes an exponential profile for the CH₄ production rates as a function of the sediment depth ($W_{\text{CH}_4}(z) = ae^{-bz}$), where a and b are solved using as inputs the bubble content and the sediment flux at each depth (Langenegger et al. 2019). These inputs were estimated using linear and polynomial interpolation respectively from data collected at each lake.

2.4.13 Vertical diffusive CH₄ flux from/to hypolimnion.

To estimate the transport of CH₄ into the SML via turbulent diffusion we applied the Fick's First Law as:

$$F_z = -K_z \frac{\partial C}{\partial z}; \qquad [\text{mmol m}^{-2} \, \text{d}^{-1}]$$
 (2.13)

where F_z is the average vertical CH₄ diffusive flux, z is depth (m), $\frac{\partial C}{\partial z}$ is the vertical gradient measured at 1 m depth resolution approximately. K_z is the vertical diffusivity derived from temperature profiles (sampling rate 4 Hz) and the Osmidov method Thorpe & Deacon 1977 as:

$$K_z = \gamma_{mix} L_T^2 N; \quad [\text{m}^2 \, \text{d}^{-1}]$$
 (2.14)

where γ_{mix} is the mixing efficiency (assumed 0.15, Wüest & Lorke 2003), N is the Brunt-Väisälä buoyancy frequency and L_T is the Thorpe scale estimated from the maximum displacement length (L_{max}) as Lorke & Wüest 2002:

$$L_T = \frac{\sqrt{2}}{7.3} L_{\text{max}}; \quad [m]$$
 (2.15)

This estimation was tested using microstructure profiles measured with a self-contained autonomous microstructure profiler (SCAMP; PME, Inc.) during the summer of 2021 in BRE, NOI and CHA (Fig. S.A.8), where turbulence profiles were resolved after Kreling *et al.* 2014.

2.4.14 Sources contribution to diffusive CH_4 emissions.

We studied the importance of each source contribution (SC) to the diffusive surface flux by computing:

$$SC_i = \frac{S_i}{\sum_j S_j} \cdot 100;$$
 [%]

where S_i is each source term (mol d^{-1}) such as bubble dissolution $(R_{\text{dis}}\forall_{\text{SML}})$, sediment flux (F_sA_s) , net production $(P_{\text{net}}\forall_{\text{SML}})$ and vertical diffusive fluxes (F_zA_z) . If $S_i \leq 0$ then $S_i = 0$ where i is each source term.

Temporal dynamics of oxic methane production in a hypertrophic Swiss lake

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Abstract

The occurrence of methane production in oxic conditions (OMP) is increasingly reported in diverse aquatic ecosystems. However, its seasonal and yearly dynamics, contribution to global methane (CH₄) emissions and key drivers, remains uncertain due to lacking measurements covering the spatial and temporal OMP variability. We used three models (full-scale mass balance, wind-based and a lateral transport) to estimate the net production of CH_4 (P_{net}) in the oxic surface mixed layer of a hypereutrophic Swiss lake. Over the stratified season, we found that OMP occurs in the studied lake, and was regularly the dominant source of CH_4 diffusive emissions during the four years. Moreover, we observed a clear pattern with high P_{net} rates at the beginning of the summer and a decrease towards the end of the stratified season. Correlations between P_{net} versus Secchi depth, chlorophyll-a concentrations and light climate suggest that photoautotrophs play a role on CH_4 production in agreement with recent findings. This work highlights that the OMP temporal variability needs to be included in CH_4 lake budget to estimate the OMP contribution on a global scale.

3.1 Introduction

During the last decades methane (CH₄) oversaturation in oxic waters in the ocean and lakes has been widely reported and named "The Methane Paradox" (Karl et al. 2008; Tang et al. 2014). Oxic methane production (OMP) has been proposed as an explanation for the Methane Paradox (Grossart et al. 2011; Bogard et al. 2014; Donis et al. 2017; Günthel et al. 2019), contrasting with the current understanding that CH₄ is only produced in anoxic conditions (Conrad 2009). While several investigations have reported that OMP does occur in the surface mixed layer in lakes (Grossart et al. 2011; Bogard et al. 2014; Donis et al. 2017; Günthel et al. 2019; Günthel et al. 2020; Chapter 2), its seasonal patterns and main drivers still remain unresolved. Two main biotic mechanisms have been suggested to produce CH₄ under oxic condition: dimetilation of methylphosphonate under phosphorus limited environment (Karl et al. 2008; Repeta et al. 2016; Wang et al. 2017; Khatun et al. 2019) and CH₄ production as a bioproduct of nitrogen fixation (Zheng et al. 2018; Luxem et al. 2020). However, positive correlations between CH₄, oxygen and chlorophyll-a concentration (Chla) point out to a direct role of phytoplankton on OMP (Bižić et al. 2020; Hartmann et al. 2020; Günthel et al. 2021). Moreover, the discovery that

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OMP follows light-dark cycles (Bižić et al. 2020; Hartmann et al. 2020) and a strong positive correlation of light and temperature (Klintzsch et al. 2020) links OMP with photosynthesis. Since primary productivity varies seasonally and from lake to lake, it is likely that OMP also varies on the basis of the trophic properties, algae species and light conditions.

Following these insights, in Chapter 2 it was proposed an empirical upscaling approach that relates Chla, secchi disk depth (SD) and light climate (LC) with the net production rates (P_{net}) normalized by the surface CH₄ concentrations. P_{net} is defined as the balance between OMP (that adds CH₄) and methane oxidation (MOx, that consume CH₄). Despite that this approach was tested on seven lakes with different trophic status, more studies on seasonal OMP changes and different types of lakes are needed to verify this trend on a larger scale.

An alternative explanation of the Methane Paradox is the transport of CH₄ from littoral sediments (Encinas Fernández et al. 2016; Peeters et al. 2019; Morana et al. 2020) but this source alone cannot support the CH₄ diffusive emissions to the atmosphere (Chapter 2). In fact, recent studies show that OMP contribution to the diffusive CH₄ emissions is up to $\sim 80\%$ in lakes during the stratified season (Donis et al. 2017; Günthel et al. 2019; Chapter 2). OMP occurs in surface waters (Tang et al. 2014; Donis et al. 2017) and, unlike anoxic CH₄ sources, can be quickly emitted to the atmosphere. Despite the limited data points on a temporal scale, Günthel et al. 2019 and Chapter 2 show that P_{net} rates have a tendency to decrease towards the end of the summer. However, it is no known if this is driven by an increase in MOx, a decrease of the OMP rates or both.

About 20% of the warming of the planet can be attributed to CH₄, and freshwater are responsible for about \sim 20 - 40% of the total CH₄ emissions to the atmosphere (202 - 425 Tg CH₄ yr⁻¹, Saunois *et al.* 2020; Rosentreter *et al.* 2021; Forster *et al.* In Press. Considering that 50% of the CH₄ emissions by lakes results from OMP in the SML and 30% is oxidized (Günthel *et al.* 2020), Bizic 2021 estimates that OMP from freshwater is about 47 - 210 Tg CH₄ yr⁻¹. These rates highlight the potential contribution of OMP globally, however there is a large uncertainty due to lacking of sufficient measurements covering the spatial and temporal OMP variability.

In this study we estimate $P_{\rm net}$ for a hypertrophic lake for four years during the stratified

period. We used a full-scale mass balance, a wind-based and a lateral transport model to calculate the $P_{\rm net}$ and analyze its seasonal pattern during the thermal stratification period. In addition, we compared each methodology and calculated the $P_{\rm net}$ contribution to diffusive emissions to the atmosphere. Finally, we tested the upscaling approach proposed in Chapter 2 to estimate $P_{\rm net}$ rates.

3.2 Methods

3.2.1 Study site and sampling.

The study was conducted in a small hypereutrophic lake, Soppensee (47.09 °N, 8.08 °E, 596 m above the sea level) located in the Canton Lucerne, Switzerland. This glacially formed lake has a surface area of 0.24 km² and a maximal depth of 26 m. The lake watershed is about 1.6 km² and lies in an area of intense agriculture (Lotter 1989).

This study relies on data from the water column profiles of CH₄, $\delta^{13}C_{CH_4}$, temperature, O₂ as well as CH₄ diffusive fluxes to the atmosphere and CH₄ ebullition rates during April 2016 to January 2018 from Vachon *et al.* 2019. Additional water column data were collected during 2018 and 2019 including surface CH₄ concentrations, $\delta^{13}C_{CH_4}$ and surface diffusive fluxes to the atmosphere (F_a) along transects performed in May and August 2018 and two transects in July 2019. Ten locations were sampled in each transect from the shore to the deepest point of the lake. The sample locations and the bathymetric map are shown in Fig. S.B.1.

In the deepest point of the lake (M1; Fig. S.B.1) a dissolved O_2 probe (miniDot, Precision Measurements Engineering) was installed at $\sim 1 \,\mathrm{m}$ depth and logged temperature, dissolved O_2 concentration (mg O_2/L) and saturation (%) every minute from April 2016 to October 2019. In addition, manual CTD (Conductivity-Temperature-Depth) profiles were performed in M1 using a multiparameter sonde (EXO2 Yellow Spring Instrument, after May 2017 a Seabird SBE19plus CTD profiler). The EXO2 profiler was equipped with a temperature, conductivity, oxygen and pH sensor while the SBE19plus was equipped additionally with turbidity, PAR, and chlorophyll-a (Chla) sensors. The temperature (T) profiles were used to define the bottom of the surface mixed layer (SML) when $\frac{\partial T}{\partial z}$ becomes smaller than $-1 \,^{\circ}\mathrm{C}\,\mathrm{m}^{-1}$ (Read $et\ al.\ 2011$). The stratified period was defined when the difference between the surface

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and bottom waters density exceeded $0.7 \,\mathrm{kg} \,\mathrm{m}^{-3}$. Water density was calculated using T and the specific conductivity at $20 \,^{\circ}\mathrm{C}$ (κ_{20}) from the CTD profiles (Imboden & Wüest 1995). In August 2016, May and Aug 2018; and July 2019 manual water profiles were performed with a spectrofluorometer (bbe Moldaenke GmbH, Schwentinental, Germany) at the deepest point of the lake to measure total Chla concentrations. Lower concentrations were found from measurements performed with the SBE19plus CTD profiler compared with the spectrofluorometer at the same time in the lake (Fig. S.B.2). Therefore, the average Chla concentration was calculated as the average value in the SML measured with the spectrofluorometer and the corrected values from SBE19plus CTD profiler using an empirical correction (Fig. S.B.2).

Diffusive CH₄ fluxes from littoral sediment (see definition below) from Langenegger et al. 2022 were also included. Additional littoral sediment cores were also retrieved to determine CH₄ diffusive fluxes using benthic chambers and Rhizons (see methods below) at the same location of the cores (a, b, and c) reported by Langenegger et al. 2022. Finally, the Secchi disk depth was also measured at each sampling campaign after August 2016. An overview of the periods, data, and methods used is given in Table 3.1.

3.2.2 Surface mass balance

The full-scale mass balance (0-D) proposed by Donis et al. 2017, a wind-based mass balance (0-D) and the lateral transport model (1-D) proposed in Chapter 2 were used to estimate net production rates (P_{net}) in the SML during the stratified period from 2016 to 2019. P_{net} is the result of the OMP and MOx ($P_{net} = OMP - MOx$), which adds and removes CH₄ in the SML, respectively. Thus, when P_{net} is positive, the true OMP rate is actually higher than P_{net} .

3.2.3 Full-scale mass balance

 $P_{\text{net,fs}}$ was determined using a mass balance in the SML during the stratified period (Eq. 3.1). This approach assumed that the SML can be modeled as a well-mixed reactor and the lateral contribution to the mass balance is equal to the littoral sediment flux times the sediment area. The spatial average of the measured values of the CH₄ concentrations (C), F_a , R_{dis} and F_z were used as boundary conditions at each sampling campaign. At the beginning of the stratified

Table 3.1. Measurements,	modelled derived	l parameters and	data sources	analyzed from	2016 to 2019
over the stratified season.					

Description	Symbol	Eq.	Method	Period	Data Source
Water column CH ₄	C		Head Space	Apr 2016 to Jan 2018	Vachon et al. 2019
Water column CH ₄	C		Head Space	2018-2019	This study
Transect CH ₄	C		Head Space	May and Aug 2018 and Jul 2019	This study
Water column $\delta^{13}C_{CH_4}$	$\delta^{13}C_{CH_4}$		Head Space	Apr 2016 to Jan 2018	Vachon et al. 2019
Water column $\delta^{13}C_{CH_4}$	$\delta^{13}C_{CH_4}$		Head Space	2018-2019	This study
Transect $\delta^{13}C_{CH_4}$	$\delta^{13}C_{CH_4}$		Head Space	2018-2019	This study
Diff. CH ₄ emissions	$F_{\mathbf{a}}$		Floating chambers	Apr 2016 to Jan 2018	Vachon et al. 2019
Diff. CH ₄ emissions	F_{a}		Floating chambers	2018-2019	This study
Diff. CH ₄ emissions transect	$F_{\mathbf{a}}$		Floating chambers	May and Aug 2018 and Jul 2019	This study
Diff. CH ₄ flux littoral sediments	F_s	Eq. 3.5	Cut-syringe	2016-2019	Langenegger et al. 2022
Diff. CH ₄ flux littoral sediments	F_s	Eq. 3.5	Rhyzons	2018-2019	This study
Diff. CH ₄ flux littoral sediments	F_s		Benthic chambers	2018-2019	This study
Total CH ₄ eb. flux below 8 m	F_{eb}		Inverted funnels	Apr 2016 to Jan 2018	Vachon et al. 2019
CH ₄ ebullition rate at SWI	R_{eb}		Bubble composition	2018-2019	Langenegger et al. 2019 and this study
CH ₄ flux at the bottom of SML	F_z	Eq. 3.8	Fick's First Law	2016-2019	This study
CH ₄ bubble dissolution rate in the SML	$R_{ m dis}$		McGinnis et al. 2006	2016-2019	Langenegger et al. 2019 Vachon et al. 2019 and this study
Net CH ₄ production rate	P _{net,fs}	Eq. 3.1	Full-scale mass balance	2016-2019	This study
Net CH ₄ production rate	$\mathrm{P}_{\mathrm{net},t}$	Eqs. 3.1 and 3.2	Wind-based mass balance	2016-2019	This study
Net CH ₄ production rate	$\mathrm{P}_{\mathrm{net,lt}}$	Eq. 3.3	Lateral transport	2016-2019	This study

period at each year $\frac{\partial C \forall_{\text{SML}}}{\partial t}$ was assumed zero.

$$\frac{\partial C \forall_{\text{SML}}}{\partial t} = A_s(t) F_s - A_a F_a(t) + A_z(t) F_z(t) + R_{\text{dis}}(t) \forall_{\text{SML}}(t)
+ P_{\text{net,fs}}(t) \forall_{\text{SML}}(t); \quad [\text{mol d}^{-1}]$$
(3.1)

The SML volume (\forall_{SML}) , the sediment area (A_s) , lake surface area (A_a) , and planar area at the bottom of the SML (A_z) are shown in Table S.B.1.

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3.2.4 Wind-based mass balance

 $P_{\text{net,t}}$ was determined using Eq. 3.1 and the diffusive CH_4 flux to the atmosphere estimated using Fick's 1st Law (Eq. 3.2). The CH_4 concentration and the sources and sinks described in Eq. 3.1 were linearly interpolated to obtain daily P_{net} rates during the stratified period.

$$F_a = k_{\text{CH}_4} (C - \text{H}_{cp} P_{\text{CH}_4}); \quad [\text{mmol m}^{-2} d^{-1}]$$
 (3.2)

The Henry constant of CH₄ dissolution at in situ temperature (H_{cp}) was the obtained following Sander 2015. The atmospheric partial pressure of CH₄ ($P_{\text{CH}4}$) was obtained from air samples taken at Soppensee and the local atmospheric pressure in each sampling campaign. The mass transfer coefficient for CH₄ ($k_{\text{CH}4}$) was estimated using the standardized gas transfer coefficient (k_{600}) following MacIntyre et al. 2010 for negative buoyancy, using 10 min average wind speed measured from a meteorological station situated about 12 km north of the lake (Egolzwil Station, Federal Office of Meteorology and Climatology MeteoSwiss). Then, $k_{\text{CH}4}$ was calculated following McGinnis et al. 2015, where CH₄ Schmidt number was estimated from the surface water temperature (Wanninkhof 1992) and the exponent n = -2/3 when $U_{10} < 3.7 \,\text{m s}^{-1}$ and n = 1/2 when $U_{10} > 3.7 \,\text{m s}^{-1}$.

3.2.5 Lateral transport model

The lateral transport model simulates the transect CH₄ concentrations in the SML following Peeters et al. 2019 and modified according to Chapter 2 (Eq. 3.3). This model considers that the surface layer is fully mixed in the vertical and, therefore, the vertical CH₄ concentrations are homogeneous within the SML.

$$\begin{split} \frac{\partial C}{\partial t} = & K_H \frac{1}{H(r)r} \frac{\partial}{\partial r} \left(H(r) r \frac{\partial C(r)}{\partial r} \right) + \frac{1}{H(r)} K_z \frac{C_{\text{hyp}} - C(r)}{\Delta z} - \frac{\overline{k}_{\text{CH}_4}}{H(r)} \left(C(r) - \text{H}_{\text{cp}} P_{\text{CH}_4} \right) \\ & + \frac{F_s}{H(r)} + R_{\text{dis}}(r) + P_{\text{net,lt}}; \qquad [\text{mol m}^{-3} \, \text{d}^{-1}] \end{split} \tag{3.3}$$

The mass transfer coefficient for CH₄ was calculated based on the average gas transfer coefficient obtained from the flux chambers ($\overline{k}_{\text{CH}_4}$), C_{hyp} is the CH₄ concentration 1 m below the bottom of the SML and $\Delta z = 1$ m and H(r) is the spatially varying thickness of the SML. At the boundaries, horizontal fluxes were assumed as zero. To estimate the horizontal dispersion

coefficient (K_H) we used Peeters & Hofmann 2015 parameterization:

$$K_H = 1.4 \times 10^{-4} L^{1.07}; \quad [\text{m}^2 \,\text{s}^{-1}]$$
 (3.4)

where the length scale L [m] was calculated as $L = r_s$. Eq. 3.4 is the average of the results 1, 3 and 4 found in Table 2 of Peeters & Hofmann 2015.

P_{net,lt} rates for each date were calculated by minimizing the root-mean-square error between the simulated transect CH₄ concentrations to the measured CH₄ concentrations using an optimization solver implemented with the *curve fit* function from Scipy (Virtanen *et al.* 2020) in Python. Note that the lateral transport model was only applicable to the transects measured in May and August 2018, and July 2019.

3.2.6 Water column CH₄ concentration and $\delta^{13}C_{CH_4}$ signature

The surface CH₄ concentrations and $\delta^{13}C_{CH_4}$ were measured using the methodology described by Vachon et al. 2019 for the water profiles and in Chapter 2 for the transect. For each sample, $\mathrm{CH_{4}}$ concentration and its $\delta^{13}\mathrm{C_{CH_{4}}}$ were measured using the headspace method following Donis et al. 2017. The samples were measured on a Cavity Ring-Down Spectrometer analyzer (Picarro G220-i, Santa Clara, CA, USA) for CH₄ concentrations in the gas phase (ppm) and stable isotope ration ($\delta^{13}C_{CH_4}$ in %). Water CH_4 concentrations were back calculated according to Wiesenburg & Guinasso 1979 accounting for water temperature, air concentration and the headspace/water ratio in the bottle. Water column profiles of CH₄ concentrations were measured at the deepest point of the lake (M1) and in June and August 2018 and July 2019 they were also measured at 12 m deep point (M2) and along 10 surface waters points in each transect. Since several measurements were taken in the SML in M1 and M2 compared with only one value at the surface of each point in the transects, the average CH₄ concentration value at the SML in M1 and M2 was assumed as part of the transect to calculate the surface average CH₄ concentration for the entire lake. This procedure minimizes the bias that can be caused by the several measurements taken at M1 and M2 in comparison with only one sample for the rest of the transect. When transect data were not available, the average value for the entire lake was calculated with the CH₄ concentrations measured in the SML.

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3.2.7 CH₄ diffusive fluxes to the atmosphere

The diffusive fluxes to the atmosphere (F_a) were estimated using a floating chamber attached to a portable GHG analyzer (UGGA; Los Gatos Research, Inc.). The floating chamber consist of an inverted plastic container with foam elements for floatation (as McGinnis *et al.* 2015). Fluxes were obtained by the slopes of the resolved CH_4 curves over ~ 5 min when the slopes where approximately linear ($R^2 > 0.97$). F_a was measured at M1 and, in June and August 2018 and July 2019, was also measured M2 and along the same locations where we measured the CH_4 concentrations transects. Since several measurements were taken at M1 and M2 compared with only one value at each point in the transects, the average value at M1 and M2 was assumed as part of the transect to calculate the average flux F_a for the entire lake. Similar to the CH_4 concentrations, this approach minimizes the bias that can be caused by the fact of several measurements taken at M1 and M2 compared to only one sample for the rest of the transect.

3.2.8 Sediment cores extraction

Additional littoral sediment cores were retrieved at the same locations as the ones presented by Langenegger et al. 2022 (Cores a, b, c, Table S1) where porewater concentrations were measured with Rhizons (Rhizosphere Research Products, The Netherlands). The sediment cores were retrieved with a gravity sediment corer (Uwitech, Mondsee, Austria) equipped with an acrylic liner of 70 cm in length and with an internal diameter of 6 cm. The liner had pre-drilled holes every 1 cm to fit Rhizons and 1 and 3 mL syringes respectively. Additional sediment cores were also retrieved in those same locations, where benthic chambers were performed to calculate the CH₄ diffusive flux from the sediment. The location and depth of each core are shown in Fig. S.B.1 and Table 3.2 respectively.

3.2.9 Porewater CH₄ concentration and $\delta^{13}C_{CH_4}$ signature

The Rhizons are porewater samplers designed to extract small volumes of water from soil and sediments in a minimally disturbing way. They are made of an inert microporous tube (0.15 µm) which were connected to a needle through a flexible 1 mm tube. About 2 mL of porewater was extracted using the Rhizons and transported through a needle into a septum-capped vacuumed 49 mL glass vials (Infocroma). Vials were shaken vigorously for 2 min after filling the vials with zero air to create reach equilibrium with the headspace. The headspace was extracted with a

glass syringe (Fortuna Optima) by gently injecting zero air (Synthetic air 5.6, Pangas AG) in the vial with a second glass syringe prior to flushing the headspace three times between the two syringes. The headspace was then measured with the Cavity Ring-Down Spectrometer analyzer (Picarro G220-i, Santa Clara, USA) directly from the glass syringe for CH₄ and δ^{13} C_{CH₄} concentrations in the gas phase. Water CH₄ concentrations were back calculated according to Wiesenburg & Guinasso 1979 accounting for water temperature, the headspace/water ratio in the vial and the dilution from the extraction with the syringes.

3.2.10 Methane diffusive fluxes from littoral sediments

Using the porewater measurements from the cut-syringes used by Langenegger *et al.* 2022 and the Rhizon methods, the littoral CH_4 sediment flux (F_s) was determined using Fick's First Law over the linear top 2 - 3 cm of the porewater concentration profile.

$$F_s = -\phi D_{\text{CH}_4} \theta^{-2} \frac{\partial C}{\partial z}; \qquad [\text{mmol m}^{-2} \, \text{d}^{-1}]$$
(3.5)

where F_s is the diffusive CH₄ flux at the sediment–water interface, ϕ the porosity of the sediments (0.9), D_{CH_4} the diffusion coefficient for CH₄ in water (1.5 m² s⁻¹ (Broecker & Peng 1974)), θ^2 the square of tortuosity (1.2) (Boudreau 1997) and $\partial C/\partial z$ the measured vertical concentration gradient.

The benthic chamber consists of a core cover with a lid leaving ~ 5 cm of headspace and ~ 30 - 50 cm of water. The lid was connected to a GHG analyzer (UGGA, Los Gatos Research, Inc.) creating a closed loop where the partial pressure of CH₄ (P_{CH_4}) was measured over time. Water CH₄ concentration (C_w) was measured at the beginning and at the end of the experiment. Benthic chamber fluxes were calculated based on the temporal change of CH₄ concentration inside the chamber following the methodology described in Chapter 2.

The littoral sediment flux was calculated as the average flux at each location, determined as the average of all the methods (when possible). The littoral sediment flux was assumed constant during the entire period of analysis.

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3.2.11 Bubble CH₄ dissolution and ebullition rates

Ebullition rates for June and September 2018 and July 2019 were obtained using measurements of bubble composition at those dates and diffusive sediment fluxes from 2016 to 2019 measurements following Langenegger et al. 2019. Using these results, we calculated the average proportion of total ebullition rates that is emitted above 8 m (χ) and we used it to estimate the total ebullition rates ($R_{\rm eb}$) from the ebullition fluxes below 8 m ($F_{\rm eb,8m}$) for 2016 and 2017 reported by Vachon et al. 2019 as:

$$R_{\rm eb}(t) = F_{\rm eb, 8m}(t)A_{\rm s,8m}(1+\chi); \quad [\rm mmol\,d^{-1}]$$
 (3.6)

where $A_{s,8m}$ is the sediment area above 8 m and t are the dates of each sampling campaign conducted during 2016 and 2017.

The temporal variation of $R_{\rm eb}$ for 2018 and 2019 was estimated assuming that $R_{\rm eb}$ for July 2018 and August 2019 are the maximum rate during each year. Next, the temporal distribution of the missing data of those years were obtained from the temporal distribution of 2016 as:

$$R_{\text{eb, i}}(t) = R_{\text{eb, i}}^{\text{max}} \frac{R_{\text{eb,2016}}(t)}{R_{\text{eb,2016}}^{\text{max}}}; \quad [\text{mmol d}^{-1}]$$
 (3.7)

where i is the year 2018 or 2019 and t is each sampling campaign conducted each year.

The bubble dissolution rates $(R_{dis}\forall_{SML})$ in the SML were estimated using the slope of the linear interpolation of the proportion of $R_{dis}\forall_{SML}$ in the SML and R_{eb} at May and August 2018 and July 2019 (Fig. S.B.3). The R_{dis} rates for May and August 2018 and July 2019 were estimated following Chapter 2, where the dissolution from a single bubble released from the sediment was calculated using McGinnis *et al.* 2006 considering a 5 mm bubble diameter.

3.2.12 Vertical CH₄ diffusive fluxes from/to the SML

Diffusive fluxes to/from the SML (F_z) were estimated using Fick's First Law:

$$F_z = K_z \frac{\partial C}{\partial z} \tag{3.8}$$

where $\frac{\partial C}{\partial z}$ at the bottom of the SML was estimated using the water column profiles of CH₄ concentrations and the K_z was determined from the heat budget method (Powell & Jassby 1974)

at the bottom of the SML.

3.2.13 Monte Carlo simulation

Monte Carlo simulations were performed (10000 iterations) when solving both mass balance models following Chapter 2. $P_{\rm net}$, $R_{\rm dis}$ and F_z were selected within a normal distribution resulting from the mean (μ) and their standard deviation (σ) retrieved from the field measurements. To avoid negative values for F_a and F_s a gamma distribution defined by the shape $(\kappa = \mu^2/\sigma^2)$ and the scale $(\theta^2 = \sigma^2/\mu)$ was chosen. Here the gamma distribution has the density $f(x) = \left(x^{\kappa-1}\frac{e^{-x/\theta}}{\theta^{\kappa}\Gamma}\right)$ where Γ is the gamma function. Random.normal and random.gamma functions from the Numpy package (Harris et~al.~2020) in Python were used for each normal and gamma distributions, respectively.

3.2.14 Sources contribution to CH₄ diffusive fluxes to the atmosphere

Following Chapter 2, each source contribution (SC) to the diffusive surface flux was estimated as:

$$SC_i = \frac{S_i}{\sum_j S_j} \cdot 100;$$
 [%]

where S_i is each source term (mol d⁻¹) such as: bubble dissolution ($R_{dis} \forall_{SML}$), sediment flux ($F_s A_s$), net production ($P_{net} \forall_{SML}$) and vertical diffusive fluxes ($F_z A_z$). If $S_i \leq 0$ then $S_i = 0$ where i is each source term.

3.3 Results

3.3.1 Limnological measurements

Surface waters remained oxygenated over the four years of measurements during the stratified period (2016 - 2019; Fig. S.B.4). The surface water temperature during the stratified period was consistent between the different years with an average value of 21 ± 3 °C. The surface mixed layer (SML) depth ranged between 1.4 - 7.5 m during the stratified period with a clear tendency to deepen towards the end of the summer (Fig. S.B.5). The average Secchi disk depth (SD) and chlorophyll-a (Chla) concentrations during the stratified period of the four years were 2.4 ± 1.1 m and 5.1 ± 3.1 mg m⁻³, respectively (Table S.B.1 and Fig. S.B.6).

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3.3.2 Surface CH₄ concentration and isotopic signature

The surface CH₄ concentrations ranged between 0.3 - 1.3 mmol/m³ during stratified period with an average value of 0.93 ± 0.28 mmol/m³ (Fig. S.B.7). In fact, the CH₄ concentrations showed a general tendency to be lower in May, increased in June and July, and decrease in September. There was a slight increase CH₄ concentrations in October likely due to the start of convective mixing (Fig. 3.1A). $\delta^{13}C_{CH_4}$ in the SML ranged between -60 to -30 % with a tendency to become more enriched towards the end of the summer (Fig. 3.1B).

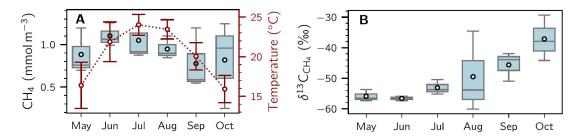


Figure 3.1. (A) Monthly average surface CH_4 concentrations (boxes) and surface temperature (red line) from 2016 to 2019. The error bars of the temperature represent the standard deviation for each month. (B) Monthly average of surface isotopic signature of CH_4 ($\delta^{13}C_{CH_4}$). Each box shows the first and third quartiles with the median (line), whiskers extend to the most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of each distribution.

Higher surface CH₄ concentrations were observed in the transect conducted in August 2019 compared to May and September 2018 (Fig. 3.2). On average, the CH₄ concentrations near the shore were 11 % higher than at the center of the lake, however the difference along each transect was not significant (ANOVA analysis). Surface $\delta^{13}C_{CH_4}$ along the transect ranged between -57 to -38 % (Fig. S.B.8). Lighter $\delta^{13}C_{CH_4}$ was observed in May 2018 compared with the other two campaigns (Fig. S.B.8). Significant differences between $\delta^{13}C_{CH_4}$ at the shore and the center of the lake were only observed on August 2019, where $\delta^{13}C_{CH_4}$ near the shore was 8% lighter than at the center (Fig. S.B.8).

3.3.3 Diffusive CH₄ emissions to the atmosphere

Diffuse CH₄ emissions (F_a) at the air-water interface (AWI) ranged between 0.1 - $3.7 \,\mathrm{mmol/m^2/d}$ during the stratified season (Fig. S.B.9). On average, in September F_a reached its lowest value ($1.0 \pm 0.5 \,\mathrm{mmol/m^2/d}$) and increased during the fall up to $2.1 \pm 1.4 \,\mathrm{mmol/m^2/d}$ (Fig. S.B.10a). The yearly average value of F_a during the stratified season was higher in 2016

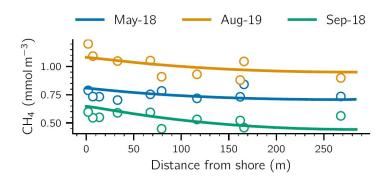


Figure 3.2. Surface CH₄ concentration along the transect sampled in Soppensee. Lines represent the CH₄ concentration simulated using the lateral transport model and dots are the measured values.

 $(1.4 \pm 1.0 \,\mathrm{mmol/m^2/d}; \,\mathrm{number~of~days~averaged~n=5})$ and 2017 $(1.8 \pm 0.8 \,\mathrm{mmol/m^2/d}, \,\mathrm{n=6})$ followed by a decrease during 2018 $(1.0 \pm 1.1 \,\mathrm{mmol/m^2/d}, \,\mathrm{n=3})$ and 2019 $(0.5 \pm 0.4 \,\mathrm{mmol/m^2/d}, \,\mathrm{n=2})$ (Fig. S.B.10b).

Five literature k_{600} parameterizations were compared with the chamber-based $k_{600,cb}$ (Fig. S.B.11). The best fit between the k_{600} literature parameterizations and $k_{600,cb}$ was obtained with MA10-NB (MacIntyre *et al.* 2010 negative buoyancy, Table S.B.2). The other four parameterizations underestimate $k_{600,cb}$ (MNB = [0.14 - 0.5]) and show a low determination coefficient ($\mathbf{r}^2 = [-26 \text{ to } 0.9]$).

Wind velocity at 10 m were used to calculate k_{CH_4} during 2016, 2017 and 2018 over the stratified period in Soppensee. Daily wind velocity at the Egolzwil meteorological station ranged between 0.6 - $8.2\,\mathrm{m\,s^{-1}}$ with an average value of $2\,\mathrm{m\,s^{-1}}$ (Fig. S.B.12). The average value of k_{CH_4} estimated with MA10-NB was $1.3\pm0.3\,\mathrm{m\,s^{-1}}$ from 2016 to 2018 with minimum and maximum values of 0.68 and $2.8\,\mathrm{m\,s^{-1}}$ respectively (Fig. S.B.12). No significant differences were found between years for the wind velocity or k_{CH_4} .

3.3.4 Diffusive CH₄ fluxes from littoral sediment

The deepest SML during the stratified season was 7.5 m, therefore sediment above the maximum SML depth was classified as littoral. Diffusive CH₄ fluxes from littoral sediments (F_s) were obtained via different methods (Table 3.2). The average F_s was $1.56 \pm 1.19 \, \text{mmol/m}^2/\text{d}$ for the entire period. About 20% lower values were obtained with the Rhizons compared to the benthic chambers and cut-syringe methods. No correlation was observed between sampled depth and F_s

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within the SML depth range.

Table 3.2. CH₄ diffusive flux from littoral sediments. Data from this work and previously published data from Langenegger *et al.* 2022 and Langenegger *et al.* 2019

Core	Date	Water Depth (m)	Temp (°C)	Diffusive flux $(\text{mmol m}^{-2} d^{-1})$				
				Benthic Chamber	Rhizons	Cut-Syringe	Average	
C1	08/12/19	0.80	22.40	3.23	0.29	1.27	1.60	
C2	08/12/19	1.00	22.40	2.38	0.05	3.14	1.86	
C3	08/12/19	1.00	22.40	0.50	0.32	1.11	0.64	
C4	07/10/17	3.00	23.70	_	-	1.43	1.43	
C5	09/13/16	4.00	21.00	_	_	3.64	3.64	
C6	07/10/17	6.00	10.30	-	-	0.22	0.22	
Average sediment flux $(\text{mmol m}^{-2} d^{-1})$						1.56 ± 1.19		

3.3.5 CH₄ ebullition rate and bubble dissolution

Monthly average ebullition rates ranged between $0.9 - 2.0 \,\mathrm{mmol/m^2/d}$ (Fig. S.B.13a). On average, the lowest and highest fluxes were found in June and September, respectively. The highest ebullitive fluxes were observed in $2018 \, (2.3 \,\mathrm{mmol/m^2/d})$ and the lowest in $2016 \, (\mathrm{Fig. S.B.13b})$. As the CH₄ dissolution rates in the SML are directly related to the ebullitive fluxes, we observed the same temporal dynamics for dissolution rates as we did for ebullition. The dissolution rates into the SML layer ranged between $0.001 - 0.06 \,\mathrm{\mu mol/m^3/d}$ (Fig. S.B.14).

3.3.6 Vertical diffusive fluxes from/to the SML

The vertical diffusive flux from/to the SML is driven by the vertical turbulent diffusivity K_z and the concentration gradient just below the bottom of the SML. K_z ranged between 0.1 - $8 \times 10^{-6} \,\mathrm{m^2 \, s^{-1}}$ with a clear tendency to be lower towards the end of the summer (Figs. S.B.16a and S.B.17). During the stratified season, F_z ranged between -0.11 to $0.22 \,\mathrm{mmol/m^2/d}$ from 2016 to 2019 (Fig. S.B.17). In May, June and October the metalimnion acted as a source of CH₄ to the SML, and as a sink in August and September. In July of all years, F_z was negligible (Fig. S.B.18a). On a yearly average scale during the stratified period, the metalimnion was a small source of CH₄ to the SML except for the year 2017 (Fig. S.B.18b).

3.3.7 Surface mass balance

The P_{net} rates obtained with the full-scaled mass balance ($P_{net,fs}$), the wind-based ($P_{net,t}$) and the lateral transport ($P_{net,lt}$) models ranged between -140 to $1010 \,\mu\text{mol/m}^3/\text{d}$ (Fig. 3.3). Seasonally, a decrease of P_{net} rates was observed from the beginning towards the end of the summer where

the highest and the lowest averaged P_{net} were estimated for June $(580 \pm 329 \,\mu\text{mol/m}^3/\text{d})$ and September $(-44 \pm 76 \,\mu\text{mol/m}^3/\text{d})$ respectively (Fig. 3.4). On average, P_{net} rates were about 350, 500, 200 and $-35 \,\mu\text{mol/m}^3/\text{d}$ for 2016, 2017, 2018 and 2019, respectively (Fig. S.B.19). Note that only three campaigns were conducted in 2018 and two at the end of the stratified period of 2019. No simulation for $P_{net,t}$ was completed for 2019 given the lack of data in the temporal scale.

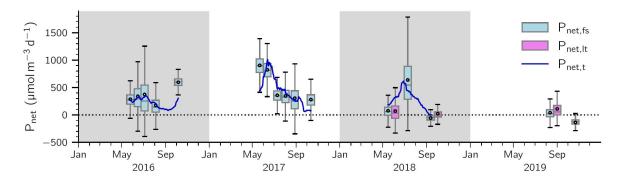


Figure 3.3. P_{net} estimation using the full-scale mass balance ($P_{net,fs}$, light blue box), the lateral transport model ($P_{net,lt}$, violet box) and the wind based model (two weeks moving average $P_{net,t}$, blue line). Each box shows the first and third quartiles with the median (line), whiskers extend to the most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of the distribution. Note that the date for $P_{net,lt}$ correspond to the date of $P_{net,fs}$ when both are calculated.

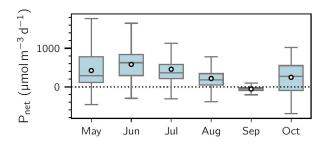


Figure 3.4. Monthly average $P_{\rm net}$ rates in the SML during the stratified season from 2016 to 2019 obtained with the full-scale mass balance model. Each box shows the first and third quartiles with the median (line), whiskers extend to the most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of the distribution.

3.3.8 Contribution of CH₄ sources to atmospheric diffusive emissions

Using the P_{net} obtained from the full-scaled mass balance ($P_{net,fs}$), we calculated the contribution of each CH_4 source to the diffusive CH_4 emissions to the atmosphere during the stratified season (Fig. 3.5). P_{net} and F_s were the major sources of CH_4 in the SML for most campaigns, with minor contributions from R_{dis} and F_z until the late summer campaigns of 2018 and 2019. On

average, P_{net} contributed about 64% of the emission from 2016 to 2019 reaching up to 90% in May 2017. Seasonally, P_{net} contributions were generally higher at the beginning of summer ($\sim 80\%$ in June) and decreased towards the end of summer ($\sim 40\%$ in September). P_{net} was the dominant source at each campaign except September 2018 and August and October 2019, when the dominant source was F_s (September 2018 and October 2019) or R_{dis} (August 2019). On average, F_s contributed about 20% from 2016 to 2019 reaching up to about $\sim 50\%$ in September 2018. The seasonal F_s contribution mirrored the seasonal P_{net} contribution trend, i.e., lower at the beginning of summer compared to the end. The contribution of bubble dissolution in the SML (R_{dis}) to diffusive CH_4 emissions was negligible (< 0.1%) in 2016 and 2018, but was $\sim 30\%$ in 2018 and 2019. The highest R_{dis} contribution was on August 2019 (38%). The F_z contribution was negligible during the entire period of analysis, except for May 2018 and October 2019 when it was 10 and 30%, respectively.

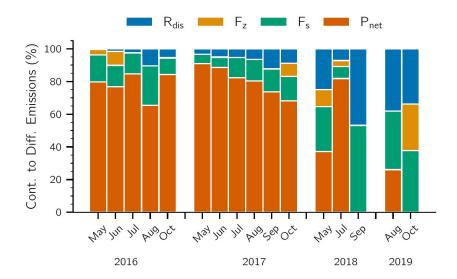


Figure 3.5. Contribution to diffusive atmospheric CH_4 emissions (F_a) from the sediment flux (F_s) , diffusive flux from hypolimnion (F_z) , bubble dissolution (R_{dis}) and net production (P_{net}) in the SML.

3.4 Discussion

In this study we quantified the rate of CH_4 net production in the oxic SML during four years in a hypereutrophic Swiss lake using a full-scaled mass balance, a wind-based mass balance, and a 1-D lateral transport model. The different approaches produced similar P_{net} rates and the same temporal pattern. P_{net} rates were positive during the stratified period, indicating that OMP needs to occur to close the mass balance in the SML (except for 2019). The average P_{net}

rates for the four years over the stratified season in Soppensee was $330 \pm 290 \,\mu\text{mol/m}^3/\text{d}$ and is in the range of previously reported values (0.2 - 1434 $\,\mu\text{mol/m}^3/\text{d}$; Günthel et al. 2021 and Chapter 2). In fact, a very similar P_{net} rate was estimated for the hypereutrophic lake Bretaye (316 \pm 2241 $\,\mu\text{mol/m}^3/\text{d}$; Chapter 2). Moreover, P_{net} was the major contributor to diffusive atmospheric CH₄ emissions from Soppensee (54 \pm 37%).

3.4.1 P_{net} seasonal variability

 $P_{\rm net}$ rates during the stratified period in Soppensee were temporally variable. The wind-based and the full-scaled mass balance models resulted in high $P_{\rm net}$ rates ($\sim 800 \, \mu {\rm mol/m^3/d}$) at the beginning of summer, indicating that OMP dominated over MOx in Soppensee surface waters. However, almost negligible rates were observed towards the end of the stratified season (Fig. 3.4), indicating that either the OMP rates decreased, the MOx increased or both. The same temporal trend was also observed in Chapter 2 and Günthel *et al.* 2019. The $\delta^{13}C_{\rm CH_4}$ signature also suggests that MOx dominated at the end of the stratified season as the residual CH₄ pool became highly $^{13}C_{\rm cerriched}$, due to methane oxidation (Reeburgh 2007). Although the mechanisms behind OMP and its impact on the $\delta^{13}C_{\rm CH_4}$ signature of CH₄ pools are not clear, our mass balance results suggest that MOx can dominate over $P_{\rm net}$ in the SML of Soppensee during the latter part of the stratified period (Fig. 3.1B).

3.4.2 Effect of methane oxidation and OMP on $\delta^{13}C_{CH_4}$ in the SML

The Keeling plot method was applied to estimate the $\delta^{13}C_{CH_4}$ of the CH₄ source in the SML (Keeling 1958). For the four years of measurements, the Keeling plot indicates that surface CH₄ might be a mixture of three sources (Fig. 3.6). For August 2019 and October 2016 and 2017, a $\delta^{13}C$ depleted source with $\delta^{13}C_{CH_4}$ of -59% whereas for September 2016 and 2018, and October 2019 the $\delta^{13}C_{CH_4}$ signature of the CH₄ source is less depleted in $\delta^{13}C$ (-55%). The positive correlation between $1/CH_4$ and $\delta^{13}C_{CH_4}$ in these two groups indicates that methane oxidation occurs during those time periods due to the CH₄ fractionation produced by methane oxidation. In contrast, OMP will not fractionate the CH₄ pool in the SML, and it will produce CH₄ with a $\delta^{13}C_{CH_4}$ that will not show a positive correlation with $1/CH_4$. However, if the isotopic signature of the carbon source of OMP changes, it will be reflected in the isotopic signature associated to OMP. These features were found in May, June and July for the four

years, where we do not observe a positive correlation between $1/\text{CH}_4$ and $\delta^{13}\text{C}_{\text{CH}_4}$. During those months higher P_{net} rates were estimated with no signal of methane oxidation found (Fig. 3.1B and 3.4), therefore the $\delta^{13}\text{C}_{\text{CH}_4}$ of the main source was estimated from the average $\delta^{13}\text{C}_{\text{CH}_4}$ signature in the SML (-55 %). This result is agreement with Hartmann *et al.* 2020 where they found that OMP likely produces CH₄ with $\delta^{13}\text{C}_{\text{CH}_4}$ values close or higher than -50 %.

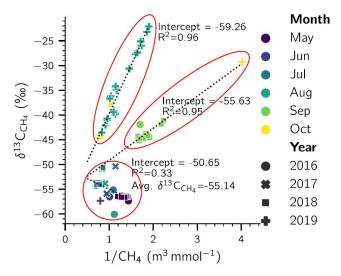


Figure 3.6. Keeling plot analysis of the surface CH_4 concentration during the stratified period from 2016 to 2019.

3.4.3 Oxic methane contribution to diffusive emissions

More evidence points to the occurrence of OMP across aquatic environments such as the ocean (Karl et al. 2008; Li et al. 2020), high altitude (Perez-Coronel & Beman 2020) and pre-alpine lakes (Chapter 2), as well as lakes in temperate and arctic regions (DelSontro et al. 2018b; Li et al. 2019). In addition, OMP has been observed using different methodologies, such as mesocosms (Bogard et al. 2014; Günthel et al. 2019), in-situ water incubations (Grossart et al. 2011; Bižić et al. 2020; Günthel et al. 2020), mass balances (Donis et al. 2017; Günthel et al. 2019; Chapter 2) and lateral transport models (Chapter 2). Despite the growing literature evidence, the OMP contribution to diffusive CH_4 emissions from inland waters remains hotly debated (Günthel et al. 2021; Peeters & Hofmann 2021). In our study, P_{net} was the greatest contributor to diffusive CH_4 fluxes to the atmosphere (F_a), reaching up $\sim 90\%$ in May 2017 with an average value of $\sim 55\%$. The temporal dynamic of the P_{net} contribution to diffusive emissions does not always follow the same trends as P_{net} rates. For example, in 2016, the

 $P_{\rm net}$ contribution to diffusive emissions was consistently around ~80 %, while in 2017 the $P_{\rm net}$ contribution decreased throughout the stratified season (Fig. 3.5). Unfortunately, we do not have enough data in 2018 and 2019 to investigate the temporal dynamic of $P_{\rm net}$ contribution.

3.4.4 OMP key drivers and upscaling approach evaluation

In Chapter 2 it was proposed an upscaling approach relating Chla, light climate (LC) and Secchi depth (SD) to the ratio of P_{net} and surface CH_4 concentration. We tested this relationship using the monthly average data from Soppensee (Fig. S.B.20) and observed a lower correlation coefficient ($R^2 = 0.56$) than in Chapter 2. Nevertheless, this positive relationship between Chla concentrations, light conditions and P_{net}/C_{CH_4} suggest that photosynthesis play a role in OMP, which is in agreement with the recent finding that photoautotrophs can produce CH_4 (Bižić et al. 2020; Günthel et al. 2020; Hartmann et al. 2020; Klintzsch et al. 2020). Moreover, when we include the data from Soppensee on Fig. 3.7, a good correlation ($R^2 = 0.83$) was observed (Fig. 3.7), implying that this approach could be useful as a first estimate of P_{net} rates on a global scale. As the current OMP dataset is still sparse, futures studies should be included to verify this trend on a larger scale.

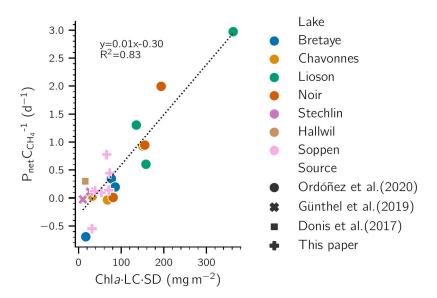


Figure 3.7. Relation between the ratio P_{net} (mmol m^{-3} d⁻¹) and the surface CH_4 concentration (C_{CH_4} , mmol/ m^3) and chlorophyll-a (Chla, mg m⁻³) × light climate ($LC = 2.5 \frac{SD}{H_{SML}}$, -) × Secchi depth (SD, m) proposed in Chapter 2. Chla is the average concentration obtained from CTD's profiles at the center of the lake corrected with Fig. S.B.2. Each point for Soppensee was calculated as the monthly average during 2016 to 2019.

3.4.5 Uncertainty assessment

P_{net} rates were estimated using three methodologies: a full-scale mass balance, a wind-based and a lateral transport models. The major advantage of the full-scaled mass balance is that it only depends on measured values for its boundary conditions but assumes that these measurements are representative of a general trend during the residence time of CH₄ in the SML (\sim 14 d). It is well known that F_a varies in time and space depending on the turbulence level at the water surface due to wind velocity and convection (MacIntyre et al. 2010; MacIntyre et al. 2021), as well as the CH₄ concentration in surface waters. To account for the effect of the temporal variability of F_a, we computed P_{net} with the wind-based model (P_{net,t}, Eqs. 3.1 and 3.2). In general, we observed a good agreement between the two methods, except for May and Oct 2016 and 2017 when P_{net,t} was lower than P_{net,fs} (Fig. 3.3). This discrepancy is likely because k_{CH4} estimated via the literature parameterization was lower than that estimated via chamber on those days (Fig. S.B.12). The chamber-based k_{CH4} may better reflect the local temporal conditions rather than the temporal average value provided by the literature parameterization (e.g., Oct 2016 data). However, this discrepancy may also point out a limitation of the k₆₀₀ literature parameterizations that neglects different phenomena, such as diffusive flux enhancement by microbubbles (McGinnis et al. 2015) or buoyancy effects due to heating at low wind speeds (MacIntyre et al. 2021). A good agreement was found between the full-scale mass balance P_{net,fs} and the lateral transport model P_{net,lt}. This latter approach accounts for the spatial variability of surface CH₄ concentrations and horizontal transport from littoral sediments using a constant k_{CH4}. Therefore, considering these spatial changes does not have a great impact on P_{net} estimations.

The other major input that can affect P_{net} estimations is F_s . In this study, we used a constant values for F_s for all four years that was the average from using three different methods on core data from six locations in the SML taken in 2016, 2017 and 2019 (Table 3.2, Fig. S.B.1). Our resulting value $(1.56 \pm 1.19 \, \text{mmol/m}^2/\text{d})$ was well within the range of those reported in the literature (0.001 - 8.8 mmol/m²/d, Huttunen et al. 2006; Bastviken et al. 2008; Peeters et al. 2019). Using six different locations to estimate F_s in the SML provided a good representation of the average littoral sediment in Soppensee. In addition, three methods were used to estimate the average flux, minimizing the bias related with each method. For example, bubbles can

be included when taking the sediment sample in the cut-syringe method or the influence of surrounding sediments in the Rizhons methods (Seeberg-Elverfeldt et~al.~2005). F_s can also vary by the increase of CH_4 production in the sediment with increasing temperature (Nozhevnikova et~al.~1997). However, our core measurements were conducted mostly in July and August when the highest water temperatures were measured, consequently the highest CH_4 production in littoral sediment would be expected. While this means that our F_s values were in the upper range of what is expected, this translates into our P_{net} estimates being conservative, as F_s is another source of CH_4 to the SML (Eqs. 3.1 and 3.3).

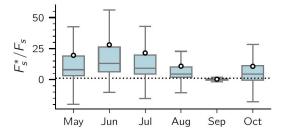


Figure 3.8. Monthly average of the proportion of littoral sediment needed (F_s^*) to compensate the surface diffusive emissions compared to measured littoral sediment flux (F_s) considering no P_{net} in the surface mixed layer. The boxes show the first and third quartiles with the median (line), whiskers extend to most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of the distributions.

As F_s has been shown to be highly influential on P_{net} estimations and a source of debate regarding OMP (Donis et al. 2017; Peeters et al. 2019; Günthel et al. 2021; Peeters & Hofmann 2021), we used the full-scaled mass balance ($P_{net,fs}$) to investigate the impact of F_s on our P_{net} estimations in Soppensee. If we assume that OMP is negligible (e.i $P_{net} = 0$), we would need up to 20 times higher sediment fluxes to account for the measured diffusive CH_4 flux to the atmosphere. This means that F_s would have to approach $F_s = \sim 30 \, \text{mmol/m}^2/\text{d}$ which exceeds the highest value reported for littoral sediment CH_4 fluxes by a factor of 3-4 (Peeters et al. 2019). We therefore conclude that the diffusive CH_4 flux from littoral sediment alone cannot account for the diffusive CH_4 flux to the atmosphere and that OMP needs to be included in the CH_4 budget of Soppensee.

3.4.6 Conclusions

Substantial evidence has been gathered over the past decade suggesting that OMP is a process that occurs in most aquatic environments, however, P_{net} rates have only been reported in a few

lakes and in a short timescale. Our study of OMP over four years in a hypereutrophic temperate lake illustrates that P_{net} (i.e., the balance between OMP and MOx) needs be included in the epilimnetic CH_4 lake budget. The good agreement between the three different models used to calculate P_{net} implies that there are no methodological issues with the models when the boundary conditions are properly determined.

We also showed in this study that P_{net} is the major contributor to the diffusive CH₄ emissions to the atmosphere from a stratified hypereutrophic lake, further supporting observations reported by Günthel et al. 2019, Donis et al. 2017 and in Chapter 2. We, however, are the first study to show this trend over multiple years and throughout the stratified period. Our temporal datasets over the stratified periods of multiple years also highlighted the tendency of P_{net} rates to begin high in the spring and decrease to almost negligible rates at the end of the stratified season, as has been observed elsewhere (Günthel et al. 2019, Chapter 2). Finally, our study also corroborates the findings that photoautotrophs may play an important role in CH₄ production in oxygenated waters and that the upscaling method proposed originally in Chapter 2 could be used as a first estimation of P_{net} for different lake environments. This estimation is thus useful as a first step in determining the potential for P_{net} in any system to impact surface $\mathrm{CH_4}$ emissions, which is important seeing as freshwater systems are responsible for about ~ 20 -40 % of natural CH₄ emissions to the atmosphere (Saunois et al. 2020; Rosentreter et al. 2021). Seeing as OMP is a CH₄ source near the air-water interface and associated with photoautotrophs, it is crucial to continue investigating the mechanisms, rates, and drivers of OMP from various aquatic systems to be able to accurately predict future aquatic CH₄ emissions in the face of global change.

Diurnal cycle of oxic methane production in a pre-alpine lake (Lac de Bretaye)

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Abstract

Breaking the traditional paradigms that methane (CH₄) is only produced by anoxic condition, evidence shows that CH₄ can also be produced in presence of oxygen (also called "oxic methane production" or OMP) by almost any living organism under oxidative stress. Recent studies show that different phytoplankton cultures are able to produce methane in oxic conditions and its production rates follow the dark-light cycle. In addition, it has been observed that OMP tends to be higher at spring/early summer and decrease at the end on the stratified season in lakes, however no research have been conducted to investigate OMP changes on a daily scale. In this work, we investigated the net production rates (P_{net}), i.e., the balance between OMP and methane oxidation, in a small eutrophic lake located in the pre-alpine region. Using day and night measurements over three days, we showed that P_{net} tends to be higher during the day while negative values during the night indicates a predominance of methane oxidation over OMP. This study highlights the importance of performing both day and night measurements to estimate the contribution of OMP in a global scale and further investigation is needed to understand the importance that photoautotrophs could play on the atmospheric CH₄ budget.

4.1 Introduction

Methane (CH₄) emissions from aquatic ecosystems contributes about 20 - 50% of the global CH₄ budget (Saunois *et al.* 2020; Rosentreter *et al.* 2021). Recent studies have shown high contribution of CH₄ produced in oxic condition (also called oxic CH₄ production or OMP) to diffusive CH₄ emissions from lakes (Bogard *et al.* 2014; Donis *et al.* 2017; Günthel *et al.* 2019; Chapter 2; Chapter 3), contrasting the assumption that CH₄ production was strictly an anaerobic process (Conrad 2009). It has been observed that net CH₄ production (P_{net}) follows a seasonal cycle, with higher contributions in spring/early summer than at the end of the stratified season (Günthel *et al.* 2019; Chapter 3). However, no investigations have been conducted to understand the diurnal OMP cycle in lakes.

Recent discoveries show that CH₄ can be produced by any organism under oxidative stress (Ernst *et al.* 2022). Several pathways have been proposed for OMP such as dissolved organic matter photoproduction (Xie *et al.* 2019; Li *et al.* 2020), biodegradation of methylphosphonate in

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phosphorus limited environments (Karl et al. 2008; Repeta et al. 2016; Wang et al. 2017; Khatun et al. 2019), archeal methanogenesis in anoxic microniches (Grossart et al. 2011) and bacteria conversion of methylamine to CH₄ (Wang et al. 2021). However, evidence that OMP follows the light-dark cycle in different phytoplankton cultures (Bižić et al. 2020; Hartmann et al. 2020), positive correlations between CH₄, oxygen and chlorophyll-a concentrations (Bižić et al. 2020; Hartmann et al. 2020; León-Palmero et al. 2020) and higher OMP rates by photoautotrophos compared with other organisms (Ernst et al. 2022), indicates the important and direct role of phytoplankton on OMP in lakes. Moreover, macrophytes, usually ignored in OMP studies, can also contribute to the oversaturation of CH₄ in the littoral zone in lakes (Hilt et al. 2022). Despite that the fact OMP can be related with different processes, an important fraction could be associated to photoautotrophos production during the day.

A key component to resolve the variability of the CH₄ dynamics in lakes is methane oxidation (MOx). It has been shown that MOx can be inhibited by light (Shelley et al. 2017) and elevated oxygen concentration (Thottathil et al. 2019). In fact, Thottathil et al. 2018 hypothesized that during the day the effect of light leads to higher CH₄ concentration with 13 C-CH₄ depleted and during the night lower CH₄ concentrations and enriched 13 C-CH₄ were observed product of MOx dynamics. Nevertheless, the impact of OMP on the variation of the CH₄ concentrations and its stable isotopic signature (δ^{13} C_{CH₄}) in the surface mixed layer (SML) remains unknown.

One of the most important sources of CH_4 in lakes is the diffusive flux from sediments, however in surface water its contribution to emissions remain controversial (Günthel et al. 2021; Peeters & Hofmann 2021). Higher temperatures could CH_4 production rates (Nozhevnikova et al. 1997), however using incubations with sediments at different temperatures, Liikanen et al. 2002 showed no significant increase on CH_4 sediment fluxes under oxic conditions. Liikanen & Martikainen 2003 further demonstrated that diffusive CH_4 fluxes from deep sediment in with oxic overlying water decrease by ~ 30 times compared to those with anoxic overlying water. Moreover, Damgaard et al. 1998 showed that sediment CH_4 front in rice paddy soils deepened several millimeters in presence of oxygen produced by photosynthesis, highlighting the idea that primary production occurring in the top layer of the sediment could create a CH_4 -oxidizing benthic filter. Therefore, we hypothesize that the contribution of CH_4 diffusive flux from littoral

sediment should be higher during the night than during the day.

It has been shown that bubble dissolution and turbulent diffusive flux from the hypolimnion are usually minor contributors of CH₄ to the SML (Donis *et al.* 2017; Günthel *et al.* 2019; Chapter 2; Chapter 3). However, deep convection due to the cooling of surface water during the night could increase the volume where bubbles can be dissolved, increase the sediment area included on the SML, as well as transport CH₄ from the metalimnion to the SML by advection. All these processes simultaneously can change the CH₄ mass balance in surface waters on a daily scale.

In this study, using a full-scale mass balance and a lateral transport model we investigate the diurnal cycle of P_{net} in a small pre-alpine eutrohpic lake. Here we report, for the first time, the diurnal cycle of P_{net} rates and its effect on the observed diurnal cycle of CH_4 concentrations and its $\delta^{13}C_{CH_4}$ in the SML over two days during the stratified season. We also investigate the diurnal cycle the different inputs of the CH_4 concentrations in the SML. Our study highlights the need to include night in-situ measurements to estimate OMP rates and its contribution to CH_4 emissions from freshwaters ecosystems on a temporal scale.

4.2 Methods

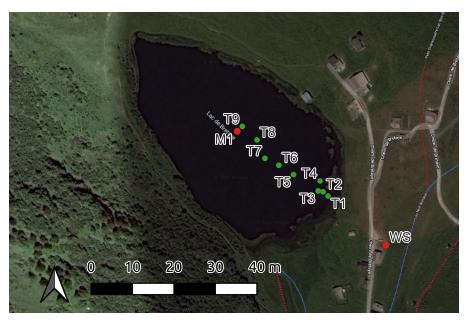
4.2.1 Study site

Lac de Bretaye is small pre-alpine lake located at 1785 m.a.s.l in the Swiss Alps (46.326 °N, 7.072 °E, Canton Vaud, Switzerland). Lac de Bretaye is a eutrophic lake, with no inflow and outflow, a maximum depth of ~ 9 m and a surface area of ~ 4 ha. Lac de Bretaye is surrounded by meadows that are used as pastures for animal grazing except for the south side that is covered mainly by *Alnus viridis* (Thöle *et al.* 2016) (Fig. 4.1c). The lake is located in a cirque with high steep slopes on north and west side of the lake (Fig. 4.1b). Abundant macrophytes are found in the north, south and west side of the lake (Fig. 4.1c).

4.2.2 Limnological measurements

The sampling campaign was conducted between the 5th to 9th July 2020 where Secchi disk depth (SD) and water column profiles were performed at the deepest point of Lac de Bretaye

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(a) Sampling locations

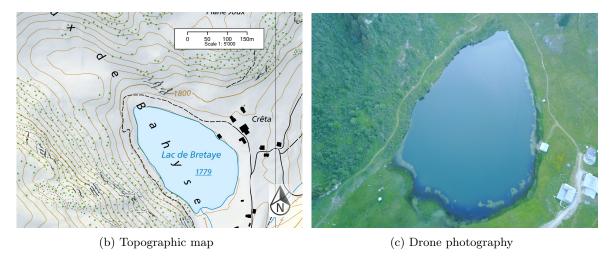


Figure 4.1. **a** Map of Lac de Bretaye. Location of transect sampling points (T1-T9), the deepest point of the lake (M1) and weather station (WS). **b** Topographic map around Lac de Bretaye © swisstopo. **c** Drone photography over the lake on July 2019.

(M1, Fig. 4.1). The water column profiles were conducted with a CTD profiler (Conductivity-Temperature-Depth, Seabird SBEplus19) equipped with conductivity, temperature, oxygen, PAR, pH, chlorophyll-a and turbidity sensors. At M1 a mooring was installed during the entire sampling campaign. Dissolved O₂ probes (Optode, miniDOT, Precision Measurements Engineering, range=0-150% saturation, accuracy=± 10 mmol m⁻³) were installed at 0, 5 and 8 m recording every 1min. Temperature loggers (Vemco Minilog-II-T loggers, Canada) were installed at 3, 4, 6, 7, 9 and 10 measuring every 1 min. Temperature/light loggers (Hobo Pendant Temperature/Light Data Logger, Bourne, Ma, USA) were deployed at 4, 6, 8 and 10 m measuring every minute and a temperature/pressure logger (Levellogger Edge M100, Solinst, Canada) was installed at the bottom of the lake.

4.2.3 Meteorological data

Shortwave radiation, air temperature, air pressure, water vapor, wind speed and direction were measured every 10 min with a weather station (Atmos41, METER Group, USA) installed at 46.325 °N, 7.074 °E (WS, Fig. 4.1a). Additionally, a portable weather station (Kestrel 3000) was deployed in the center of the lake when diffusive CH₄ emissions were measured.

4.2.4 Water CH₄ concentrations and its isotopic signature

CH₄ concentrations and carbon stable isotopic signature of CH₄ (δ^{13} C_{CH₄}) profile were conducted at M1 on July 5th with 1 m vertical resolution. Two 24 hours campaign were performed at M1 where measurements were performed every four hours. The first 24h campaign started at July 5th at 18:00 CET and ended July 6th at 23:00 CET. The second was conducted from July 7th at 18:45 CET to July 8th at 19:00 CET. Additionally, a longitudinal transect composed of 9 stations was performed on July 7th from the shore to M1 (Fig. 4.1).

For the CH₄ concentrations and $\delta^{13}C_{CH_4}$ water column profile, the water samples were taken with a 5-L Niskin bottle and the gently transferred into a 1 L glass bottle (Duran GmbH, Mainz, Germany), while for the transect, the water samples were taken directly with the glass bottle (Duran GmbH, Mainz, Germany). CH₄ and $\delta^{13}C_{CH_4}$ were measured using the headspace method following Donis *et al.* 2017. A Cavity Ring-Down Spectrometer analyzer (Picarro G220-i, Santa Clara, CA, USA) was used to measure the CH₄ concentration and its

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stable isotopic ratio in the gas phase. Water CH₄ concentrations were back calculated following Wiesenburg & Guinasso 1979 accounting for the headspace/water ratio, air concentration and water temperature in the bottle.

4.2.5 Diffusive CH_4 emissions

During the 24h campaigns, at each measuring point diffusive CH_4 emissions (F_a) were measured three times at station M1. At each station along the transect triplicates were also performed. F_a was measured using a floating chamber attached to a portable GHG analyzer (UGGA, Los Gatos Research, Inc.) (see McGinnis *et al.* 2015 for further details). The chamber was deployed from the boat and allowed to freely drift to minimize artificial turbulence disturbance. Fluxes were obtained by the slope of the CH_4 concentration curve over the first ~ 5 min when the curve was approximately linear ($R^2 > 0.97$).

The mass transfer coefficient for $\mathrm{CH_4}$ ($\mathrm{k_{CH_4}}$) was estimated from the chamber measurements following:

$$F_a = k_{\text{CH}_4} (C_w - H_{\text{cp}} p \text{CH}_4); \quad [\text{mmol m}^{-2} d^{-1}]$$
 (4.1)

where H_{cp} is the Henry's constant of CH₄ at in situ temperature obtained from Sander 2015. C_w and pCH₄ are the surface CH₄ concentration and the atmospheric partial pressure of CH₄, respectively.

4.2.6 Vertical diffusive flux to/from the SML

Vertical diffusive flux just below the SML was estimated using Fick's First Law as:

$$F_z = -K_z \frac{\partial C}{\partial z}; \qquad [\text{mmol m}^{-2} \, \text{d}^{-1}]$$
 (4.2)

where the vertical concentration gradient $(\frac{\partial C}{\partial z})$ was estimated from the water column profile just below the SML. The turbulent diffusivity (K_z) was calculated as the log-average of 13 microstructure profiles measured with a self-contained autonomous microstructure profiler

(SCAMP, PME, Inc.) considering:

$$K_z = \gamma_{mix} \frac{\epsilon}{N^2}; \qquad [\text{m}^2 \,\text{d}^{-1}]$$
 (4.3)

where the dissipation rate ϵ and N² were resolved after Kreling *et al.* 2014 and γ_{mix} assumed 0.15 (Wüest & Lorke 2003).

4.2.7 Mass balance in the SML

The net production rate (P_{net}) in the oxic SML was estimated using a full-scale mass balance every 4 hours during the two 24h sampling campaigns following Donis *et al.* 2017 (P_{net,fs}, Eq. 4.4). Moreover, a lateral transport model proposed in Chapter 2 (P_{net,lt}, Eq. 4.5, similar to Peeters *et al.* 2019) was used to estimate P_{net} rates fitting the simulated CH₄ concentration to the measured CH₄ concentrations along the transect using Eq. 4.5.

$$\frac{\partial C \forall_{\text{SML}}}{\partial t}(t) = A_s F_s - A_a F_a(t) + A_z F_z + R_{\text{dis}} \forall_{\text{SML}} + P_{\text{net,fs}}(t) \forall_{\text{SML}}; \quad [\text{mol d}^{-1}] \quad (4.4)$$

We consider a constant littoral sediment flux (F_s) and bubble dissolution rate (R_{dis}). R_{dis} was determined as the same rate obtained in Chapter 2 for July 2019 for the same lake. For the full-scale mass balance, we consider a constant hypolimnetic flux (F_z, Eq. 4.2) while for the lateral transport model we consider that $\frac{\partial C}{\partial z} \sim \frac{C_{\text{hyp}} - C}{\Delta z}$ where C_{hyp} is the concentration 1 m below the bottom of the SML and $\Delta z = 1$ m.

$$\frac{\partial C}{\partial t} = K_H \frac{1}{H(r)r} \frac{\partial}{\partial r} \left(H(r)r \frac{\partial C(r)}{\partial r} \right) + \frac{1}{H(r)} K_z \frac{C_{\text{hyp}} - C(r)}{\Delta z} - \frac{\overline{k}_{\text{CH}_4}}{H(r)} \left(C(r) - H_{\text{cp}} p \text{CH}_4 \right)
+ \frac{F_s}{H(r)} + R_{\text{dis}}(r) + P_{\text{net,lt}}; \quad [\text{mol m}^{-3} d^{-1}]$$
(4.5)

For each CTD, the seasonal SML depth ($H_{\rm SML}$) was defined when $\partial T/\partial z$ was smaller than $-1\,^{\circ}{\rm C}\,{\rm m}^{-1}$ (Read et al. 2011) and assumed constant during the entire sampling campaign. The SML depth for the lateral transport model (H(r)) was assumed equal to $H_{\rm SML}$ in the pelagic zone and decrease linearly to zero in the littoral region. The surface area (A_a), the sediment area (A_s), the planar area at the bottom of the SML (A_p) and the SML water volume ($V_{\rm SML}$) were obtained from Chapter 2. The horizontal dispersion coefficient (K_h) was estimated

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following Peeters & Hofmann 2015. The average mass transfer coefficient (\overline{k}_{CH_4}) was obtained from Eq. 4.1. A summary of the inputs values is shown in Table 4.1. Monte Carlo simulation were performed following Chapter 2 to calculate the uncertainty of the model.

4.2.8 Sources contribution to CH₄ diffusive fluxes to the atmosphere

Following Chapter 2, the contribution of each source (SC) to the diffusive surface flux was estimated as:

$$SC_i = \frac{S_i}{\sum_j S_j} \cdot 100;$$
 [%]

where S_i is each source term (mol d^{-1}) such as: sediment flux (F_sA_s) , bubble dissolution $(R_{\text{dis}}\forall_{\text{SML}})$, net production $(P_{\text{net}}\forall_{\text{SML}})$ and vertical diffusive fluxes (F_zA_z) . If $S_i \leq 0$ then $S_i = 0$ where i is each source term.

4.3 Results

4.3.1 Study site general features

The average Secchi disk depth during the campaign was 3.24 ± 0.25 m. The seasonal surface mixed layer (SML) depth (H_{SML}) define by the temperature profile was ~ 3.10 m. The SML remained oxic during the entire sampling period (Fig. S.C.6). A_a , A_s , A_p and $\forall_{\rm SML}$ were $45\,500$, $16\,100$ and $30\,080\,{\rm m}^2$, and $112\,350\,{\rm m}^3$, respectively. All the following results are presented in local time.

4.3.2 Meteorological data

The air temperature (T_a) ranged between 3.6 - 19 °C with an average value of 11 ± 3 °C (Fig. 4.5b). During the afternoon of July 6th a cold front decreased the air temperature and the wind speed increased over Lac de Bretaye (Fig. 4.5b). The average wind speed at 10 m (U₁₀) on July 7th was $1.74 \,\mathrm{m\,s^{-1}}$. After the cold front, the average temperature increased and wind speed decreased during July 8th (T_a =12.5 °C and U_{10} =0.94 m s⁻¹). The cold front also brought cloudiness producing lower and highly variable short wave solar radiation compared to the rest of the days. Significant cloud cover also occurred on the afternoon of July 8th (Fig. 4.5a).

Table 4.1. Measurements, modelled derived parameters and data sources analyzed during the sampling campaign.

Description	Symbol	Unit	Value	Source
Surface Area	A_a	m^2	45500	Chapter 2
Sediment Area	A_s	m^2	16100	Chapter 2
Planar Area at base of SML	A_p	m^2	30080	Chapter 2
SML Volume	$V_{ m SML}$	m^3	112350	Chapter 2
SML depth	${ m H}_{ m SML}$	m	3.10	This study
Transect average mass transfer coefficient	$\overline{k}_{\mathrm{CH_4}}$	$\rm md^{-1}$	1.51 ± 0.51	This study
Horizontal dispersion	K_H	$\mathrm{m}^2\mathrm{d}^{-1}$	2034	Chapter 2
CH ₄ concentration 1m below the bottom of the SML	C_{hyp}	${\rm mmolm^{-3}}$	1.59	This study
Surface CH ₄ concentration	C	${\rm mmolm^{-3}}$	Fig. 4.2a	This study
Surface CH ₄	$\delta^{13}C_{CH_4}$	‰	Fig. 4.2a	This study
Water column CH ₄	C	${\rm mmolm^{-3}}$	Fig. S.C.1	This study
Water column $\delta^{13}C_{CH_4}$	$\delta^{13}C_{CH_4}$	‰	Fig. S.C.1	This study
Littoral sediment Flux	F_s	$\mathrm{mmol}\mathrm{m}^{-2}\mathrm{d}^{-1}$	3.0 ± 6.7	This study
Diffusive CH ₄ emissions	F_{a}	$\mathrm{mmol}\mathrm{m}^{-2}\mathrm{d}^{-1}$	Fig. 4.2b	This study
Vertical diffusive flux	F_z	$\mathrm{mmol}\mathrm{m}^{-2}\mathrm{d}^{-1}$	0.02 ± 0.01	This study
Bubble dissolution at SML	R_{dis}	$\mu\mathrm{mol}\mathrm{m}^{-3}\mathrm{d}^{-1}$	0.04 ± 0.01	Chapter 2
Net CH ₄ Production	P _{net}	$\mu\mathrm{mol}\mathrm{m}^{-3}\mathrm{d}^{-1}$	Fig. 4.5	This study

4.3.3 Vertical profiles

The average water temperature in the SML was about $18\,^{\circ}$ C whereas at the bottom was about $9\,^{\circ}$ C (Figs. S.C.1a and S.C.1b). Surface water remained oxic during the day and night whereas below 6 m, the water column became anoxic (Figs. S.C.1a and S.C.1b). Chlorophyll-a (Chla) was observed along the entire water column with a peak at the base of the SML and at the bottom of the lake. During the day, lower Chla concentrations $(1.73\pm0.07\,\mathrm{mg\,m^{-3}},\mathrm{Fig.\,S.C.1a})$ were measured in the SML compare to the night period $(4.09\pm0.56\,\mathrm{mg\,m^{-3}},\mathrm{Fig.\,S.C.1b})$. This difference could be explained due to transport of phytoplankton from the thermocline caused by deep convection during the night or the influence of non-photochemical quenching in our measurements during the day (Harrison et al. 2018).

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From 0 - 4 m, the CH₄ concentrations remain fairly constant $(1.39 \pm 0.32 \,\mathrm{mmol/m^3})$ and then rapidly increased to $141.17 \,\mathrm{mmol/m^3}$ at the bottom of the lake (Fig. 4.4a). The average isotopic signature $(\delta^{13}\mathrm{C}_{\mathrm{CH_4}})$ in the SML was about $-49.8 \,\mathrm{\%}$, whereas in the oxic hypolimnion $\delta^{13}\mathrm{C}_{\mathrm{CH_4}}$ reached up to $-25 \,\mathrm{\%}$ at 6 m to then decreased to $-64 \,\mathrm{\%}$ at the anoxic bottom of the lake.

4.3.4 Surface CH₄ concentrations and $\delta^{13}C_{CH_4}$ signature

The CH₄ concentrations in the SML varied between $0.8 - 1.4 \,\mathrm{mmol/m^3}$ with an average value about $1.1 \pm 0.1 \,\mathrm{mmol/m^3}$ during the three days of measurements (Fig. 4.2a). A daily cycle was observed, where CH₄ concentrations reached up to $1.2 \,\mathrm{mmol/m^3}$ during the day followed by a decrease towards the night where CH₄ concentrations decreased to about $1.0 \,\mathrm{mmol/m^3}$ (Fig. 4.4). The average CH₄ concentration along the transect was about $1.23 \pm 0.01 \,\mathrm{mmol/m^3}$ with values near the shore $16 \,\%$ higher than in the center (Fig. 4.3).

The average isotopic signature $\delta^{13}C_{CH_4}$ in the SML were $-52.9 \pm 1.1 \%$ (Fig. 4.2a). Consistent $\delta^{13}C_{CH_4}$ of about -53 % was observed from 03:00 to 18:00, then enriched CH₄ was observed towards 24:00 reaching up to $\sim -51 \%$ (Fig. 4.4a). The average $\delta^{13}C_{CH_4}$ along the transect was about $-53.0 \pm 0.5 \%$ without a clear difference between near the shore versus the center of the lake (Fig. S.C.3).

4.3.5 Diffusive CH_4 emissions to the atmosphere

Diffusive CH₄ emissions (F_a) at the air-water interface (AWI) ranged between 0.43 - $3.28 \,\mathrm{mmol/m^2/d}$ with an average value of $1.52 \pm 0.85 \,\mathrm{mmol/m^2/d}$ (Fig. 4.2b). Between 00:00 to 6:00 F_a was about $1.14 \pm 0.75 \,\mathrm{mmol/m^2/d}$, then increased up to about $2.81 \pm 0.39 \,\mathrm{mmol/m^2/d}$ between 14:00 to 15:00, then finally decreased to about $0.79 \pm 0.31 \,\mathrm{mmol/m^2/d}$ from 19:00 to 00:00. The average F_a along the transect was $1.99 \pm 0.62 \,\mathrm{mmol/m^2/d}$ with an average mass transfer coefficient ($\overline{k}_{\mathrm{CH_4}}$) value of $1.51 \pm 0.51 \,\mathrm{m\,d^{-1}}$ and during the entire campaign was $1.29 \pm 0.68 \,\mathrm{m\,d^{-1}}$.

The normalized mass transfer coefficient (k_{600}) ranged 0.41 - 3.06 m d⁻¹ (Fig. 4.4c). We observed a daily cycle of k_{600} where the maximum values were measured around 14:00

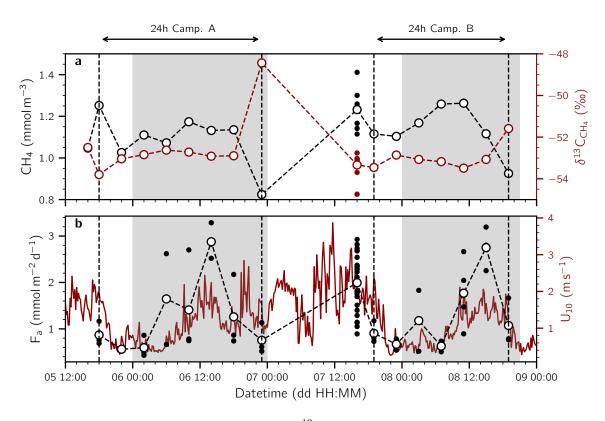


Figure 4.2. **a** Surface CH_4 concentrations and its $\delta^{13}C_{CH_4}$ isotopic signature. **b** Diffusive CH_4 fluxes to the atmosphere and the wind speed at $10 \,\mathrm{m}$ (U_{10}). The white dots represent the average value while the filled dot are each measurement.

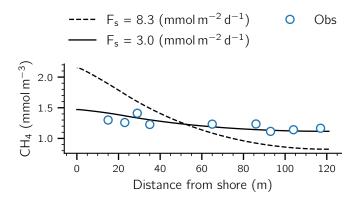


Figure 4.3. Surface CH_4 concentrations along the transect. Lines represent the CH_4 concentration simulated using the lateral transport model and dots are the measured values. For the dash line and the continuous line, 8.3 and $3.0\,\mathrm{mmol/m^2/d}$ was used as a CH_4 flux from littoral sediment (F_s) respectively.

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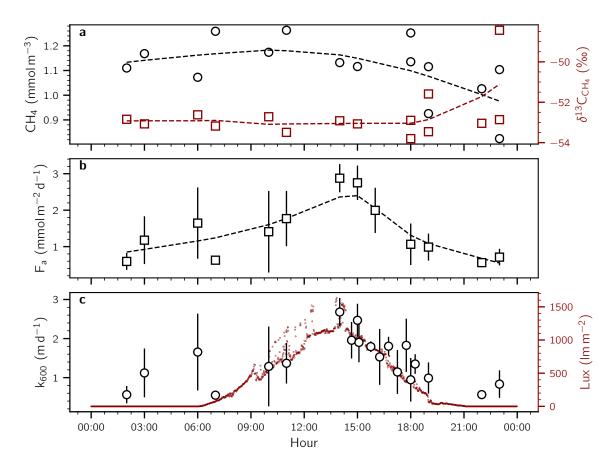


Figure 4.4. Daily cycles during the 24h campaigns for $\bf a$ surface CH_4 concentrations and $\delta^{13}C_{CH_4}$ isotopic signature and $\bf b$ diffusive CH_4 fluxes to the atmosphere (F_a) and $\bf c$ the normalized mass transfer coefficient (k₆₀₀) and light intensity at 3 m. The error bars represent the standard deviation of three surface diffusive fluxes performed each time. The dashed line is the smooth data obtained with locally weighted running line smoother (LOESS method).

with a clear decrease towards the night. The average k_{600} during the sampling campaign was $1.35\,\mathrm{m\,d^{-1}}$.

4.3.6 Vertical diffusive CH₄ flux through the bottom the SML

The vertical diffusive fluxes through the SML is driven by the vertical turbulent diffusivity (K_z) and the concentration gradient just below the bottom of the SML. K_z ranged between 0.26 - $4.30 \times 10^{-6} \,\mathrm{m^2 \, s^{-1}}$ and at the bottom of the SML was about $6.47 \times 10^{-6} \,\mathrm{m^2 \, s^{-1}}$ (Fig. S.C.1c). Using the gradient between the CH₄ concentration measured at 3 and 4 m (Fig. S.C.1d), the flux through the bottom of the SML (F_z) was $0.02 \pm 0.01 \,\mathrm{mmol/m^2/d}$ and assumed constant during the entire sampling campaign for the full-scale mass balance.

4.3.7 Mass balance in the SML

The net production rates ($P_{net} = OMP - MOx$) were obtained using the full-scaled mass balance ($P_{net,fs}$) and the lateral transport model ($P_{net,lt}$). The littoral sediment flux (8.3 mmol/m²/d) and the bubble dissolution rates obtained (0.04 µmol/m³/d) in July 2019 from Chapter 2 were used on both models to calculate P_{net} (Fig. S.C.2). A large discrepancy between the model and the measured CH_4 concentration near the shore was obtained using the measured F_s from Chapter 2 (Fig. 4.3). Therefore, we use the lateral transport model to estimate P_{net} and F_{net} that best fits the measured and simulated CH_4 concentrations (Fig. 4.3). Following this procedure we obtained that $F_s = 3.0 \, \text{mmol/m²/d}$ and $P_{net} = -99.7 \pm 390.0 \, \text{µmol/m³/d}$. Using the estimated F_s we obtained with the lateral transport model, the $P_{net,fs}$ ranged between -1636 to $750 \, \text{µmol/m³/d}$ with an average value of $-115 \, \text{µmol/m³/d}$ (Fig. 4.5a). On average P_{net} rates were about $534 \, \text{µmol/m³/d}$ from 00:00 to 6:00 to then increased to $575 \, \text{µmol/m³/d}$ at 10:00 to then decreased to $-994 \, \text{µmol/m³/d}$ at 23:00 (Fig. 4.6). The average P_{net} during the day (between 7:00 to 20:00) was $37 \pm 1119 \, \text{µmol/m³/d}$ and $-485 \pm 1306 \, \text{µmol/m³/d}$ during the night (between 21:00 to 6:00).

4.3.8 Sources contribution to diffusive CH₄ emissions

We calculated the daily cycle of the contribution of each CH₄ source to F_a (Fig. 4.7). The major contributor was the littoral sediment flux (F_s) with an average value about $69 \pm 19 \%$. During the night (from 21:00 to 6:00) F_s contribution was $76 \pm 18 \%$ and $64 \pm 19 \%$ during the day (from 7:00 to 20:00). On the other hand, P_{net} contribution was about $30 \pm 21 \%$ during the day and $16 \pm 20 \%$ during the night, with an average value of $24 \pm 21 \%$. The contribution of the bubble dissolution and vertical flux at the bottom of the SML was negligible, with an average value of $6.4 \pm 0.1 \%$ and $0.5 \pm 0.1 \%$ respectively.

4.4 Discussion

On average, we observed that methane oxidation (MOx) was predominant over OMP in the surface mixed layer of Lac de Bretaye (P_{net} <0). However, we observed that P_{net} rates tend towards a diurnal cycle with positive P_{net} values during the day that begin to decrease in the afternoon until they become negative around 18:00 reaching its minimum value around midnight

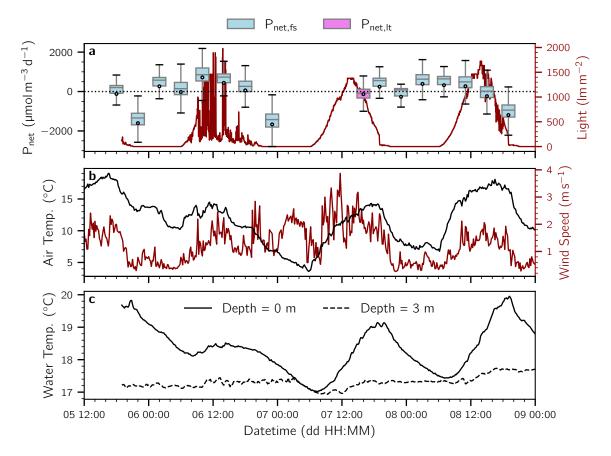


Figure 4.5. a $P_{\rm net}$ estimation using the full-scale mass balance ($P_{\rm net,fs}$, light blue box) and the lateral transport model ($P_{\rm net,fs}$, violet box) using $F_s{=}3.0\,{\rm mmol/m^2/d}$. Each box shows the first and third quartiles with the median (line), whiskers extend to the most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of the distribution. b Air temperature and wind speed measured with the Atmost 41 weather station. c Water temperature measured at 0 and 3 m depth at M1.

(Fig. 4.6). Here, we first investigate if P_{net} rates are an artefact of the methodology, then we analyze the impact of our model assumptions on our calculations and finally, we explore which are the different drivers to explain P_{net} temporal dynamics.

4.4.1 Is P_{net} an artefact of the methodology?

The diffusive CH₄ emissions to the atmosphere (F_a) and the change of mass in the SML ($\frac{\partial CV_{\text{SML}}}{\partial t}$) were the only two variable parameters to estimate P_{net} with the full-scale mass balance (Eq. 4.4). F_a depends on the surface CH₄ concentration and the mass transect coefficient (k_{CH_4}) (Eq. 4.1). Strong correlations between k₆₀₀ and the measured F_a indicate that the changes in F_a were mainly driven by k₆₀₀ and not by the surface CH₄ concentration changes (Fig. S.C.4). We

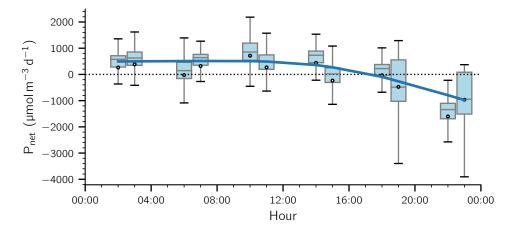


Figure 4.6. Daily cycle of $P_{\rm net}$ rates obtained with the full-scale mass balance model during the two 24h campaigns. Each box shows the first and third quartiles with the median (line), whiskers extend to the most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of the distribution. The solid blue line is the smooth data obtained with locally weighted running line smoother (LOESS method).

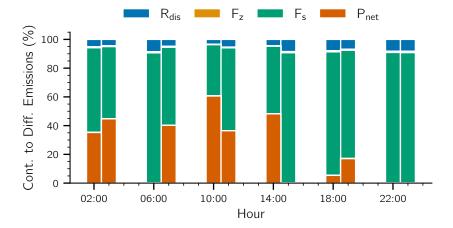


Figure 4.7. Daily cycle of contribution from bubble dissolution $(R_{\rm dis})$, littoral sediment flux $(F_{\rm s})$, vertical flux from hypolimnion $(F_{\rm z})$ and net production $(P_{\rm net})$ in the SML to the diffusive CH_4 emissions $(F_{\rm a})$

observed that k_{600} has a daily cycle and correlates much higher with solar radiation (r^2 =0.43) and in-situ light (r^2 =0.54) (Figs. S.C.4c, S.C.4d and 4.4c) rather than U_{10} (r^2 = 0.05, Fig. S.C.4b). We proposed that this could be because k_{600} is enhanced by microbubbles (McGinnis *et al.* 2015) produced by photosynthesis in the water column (Melack & Kilham 1974; Koschorreck *et al.* 2017), by surface heating (MacIntyre *et al.* 2021) or both. Nevertheless, weak correlation between k_{600} and P_{net} indicates that k_{600} is not the main driver of P_{net} (Fig. 4.8) and therefore derived P_{net} rates are mainly driven by $\frac{\partial C \forall_{\text{SML}}}{\partial t}$.

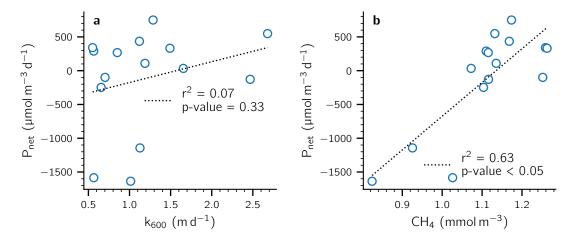


Figure 4.8. P_{net} correlations with **a** mass transfer coefficient (k_{600}) and **b** surface CH_4 concentrations.

4.4.2 Can we explain $\frac{\partial C \forall_{\text{SML}}}{\partial t}$ if we do not consider P_{net} ?.

Changes in F_a control how fast CH_4 in the SML will be released to the atmosphere. However, the rapid changes of CH_4 mass cannot be explained by F_a because, considering $k_{CH_4} = 1 \text{ m d}^{-1}$ and no other CH_4 sources or sinks occurring in the SML, it would take about 1.5 days to observe a 50% decrease on the surface CH_4 concentration. Moreover, under the same condition, F_a would only decrease the surface CH_4 concentration by $\sim 2\%$ in an hourly scale. Therefore, F_a does not drive the rapid changes of mass in the SML.

The other major source that can drive the change of mass in the SML is the diffusive CH₄ flux from littoral sediments (F_s). Since light can reach the littoral sediment, photosynthesis on the top layer of the sediment would potentially decrease F_s during the day (Damgaard *et al.* 1998; Liikanen *et al.* 2002). Therefore, higher CH₄ concentrations should be observed due to higher F_s during the night. In contrast, in Lac de Bretaye we observed an increase of CH₄ during

the day compare to the night. Moreover, conducting a similar analysis as F_a , we simulated the temporal change of F_s increasing the concentration at the shore of the lake by 100% compared to the initial condition. Without considering any other source/sink of CH_4 in the SML, the changes of F_s would need more than 10 h to change by 5% the CH_4 concentration in the center of the lake (Fig. S.C.7). These results indicate that a new source or sink needs to be included in the SML to close the mass balance on an hourly scale.

4.4.3 Modeled sediment flux estimation.

Accurate estimations of CH₄ flux from littoral sediment are difficult to obtain. F_s based on sediment cores such as benthic chambers and pore water measurements lack spatial coverage in the littoral zone. Due to coarse vertical resolution (\sim 1 cm) of pore water measurements, it might not able to capture the decrease of F_s due to oxidation caused by the presence of oxygen at the top of the sediment (Damgaard et al. 1998; Liikanen et al. 2002). In contrast, the transect data approach estimates the overall contribution from different sources of CH₄ from the littoral zone such as CH₄ production by macrophytes located at the shore of the lake (Hilt et al. 2022) and in littoral sediments. However, this method assumes no horizontal advection of CH₄ from other zones, and it could be only applicable to lakes with high horizontal CH₄ gradients.

In Chapter 2, F_s in Lac de Bretaye (8.3 \pm 6.7 mmol/m²/d) was estimated using benthic chambers and pore water measurements in three sediment cores taken in Sept 2018 and July 2019. This average flux, highly influenced by one high measurement of 13.5 mmol/m²/d, is one of the highest values ever reported in literature (Huttunen *et al.* 2006; Bastviken *et al.* 2008; Peeters *et al.* 2019). However, using that measured sediment flux in the lateral transport model, we observed a large discrepancy between the measured and model surface CH₄ concentration (Fig. 4.3) (note that P_{net} was obtained from the best fit the model to the observed surface CH₄ concentrations along the transect). To compensate for the high concentration at the shore due to the high F_s , the model finds a more negative P_{net} that underestimates the CH₄ concentration at the deepest point of the lake. Therefore, we resolved Eq. 4.5 using a combination of P_{net} and F_{net} that best fit the CH₄ transect concentrations at the surface. This new F_s estimated in July 2020 (3.0 mmol/m²/d) was lower compared to 2018 and 2019. The difference may be associated with more oligotrophic conditions, evidenced by the low concentrations of CH₄ and Chl*a* in the

SML observed in July 2020 during the day (Table 4.2).

Table 4.2. Summary of main drivers of net production rates (P_{net}) during July 2020 (this study), June and September 2018, and July 2019 (Chapter 2): Light climate (LC=2.5SD/ H_{SML}), Secchi depth (SD), average chlorophyll-a concentration in the SML (Chla), average CH₄ concentration and its isotopic signature ($\delta^{13}C_{CH_4}$) in the SML, and diffusive CH₄ flux from littoral sediments (F_s).

Date	$\mathbf{LC} \atop (\mathrm{m}\mathrm{m}^{-1})$	SD (m)	$\frac{\mathbf{Chl}\boldsymbol{a}}{(\mathrm{mg}\mathrm{m}^{-3})}$	$\delta^{13}C_{CH_4} \ (\%)$	$\begin{array}{c} \mathbf{CH_4} \\ \mathrm{(mmol/m^3)} \end{array}$	$\begin{array}{c} \mathbf{F_s} \\ (\mathrm{mmol/m^2/d}) \end{array}$	$\frac{\mathbf{P_{net}}}{(\mu mol/m^3/d)}$
Sept 2018 July 2019 July 2020 July 2020	Day 7.02 Day 1.42 Day 4.52 Day 2.61 Night - ampaign 2.61	3.65 2.95 4.70 3.24 - 3.24	3.01 4.08 4.05 1.73 4.09 3.12	-52.0 -38.0 -48.8 -53.0 -52.1 -52.7	6.7 ± 2.3 3.5 ± 0.5 2.8 ± 1.6 1.2 ± 0.1 1.1 ± 0.1 1.1 ± 0.1	$8.3 \pm 6.7 \\ 8.3 \pm 6.7 \\ 8.3 \pm 6.7 \\ 8.3 \pm 6.7 \\ 3.0 \pm 6.7 \\ 3.0 \pm 6.7 \\ 3.0 \pm 6.7$	$\begin{array}{c} 2313 \pm 1882 \\ -1941 \pm 1266 \\ 687 \pm 1210 \\ 37 \pm 1190 \\ -485 \pm 1306 \\ -169 \pm 1223 \end{array}$

4.4.4 Effect of choosing a constant or variable mixed layer depth on P_{net} .

Our study considers the seasonal SML depth to define the water volume for the CH₄ mass balance. We analyzed the impact of choosing a variable local H_{SML} ($P_{net,mod}$) instead of choosing a constant seasonal H_{SML}^* ($P_{net,0}$) on P_{net} calculations. For the simulation, we considered a variable H_{SML} ranged from 50 to 150% of H_{SML}^* (Fig. 4.9a). In addition, we considered a constant $F_a = 1.5 \, \text{mmol/m}^2/\text{d}$, $R_{dis} = 0.02 \, \mu \text{mol/m}^3/\text{d}$ and CH₄ concentration in the SML CH₄ = 1 mmol/m³. F_z changes were assumed negligible due to expected small changes on the CH₄ concentration inside the seasonal SML (see discussion below of F_z when $H_{SML} > H_{SML}^*$). When $H_{SML} < H_{SML}^*$, $P_{net,0}$ would be underestimated when $P_{net,0}$ is positive and overestimated when $P_{net,0}$ is negative (Fig. 4.9b, case $F_{net} = 3 \, \text{mmol/m}^3$ up to 300% underestimation). In contrast, when the local H_{SML} is higher than H_{SML}^* , $P_{net,0}$ would be overestimated when $P_{net,0}$ is positive and underestimated when $P_{net,0}$ is negative. Note that at higher $P_{net,0}$ rates, the impact of choosing a variable or constant H_{SML} is lower. Hence, including the change of H_{SML} in our calculations would tend to increase the difference of P_{net} during day and night considering $P_{net} > 0$ during the day and $P_{net} < 0$ during the night.

During our sampling campaigns we observed higher $P_{\rm net}$ rates during the day followed by a decrease during the evening (Fig. 4.6). However, between 00:00 to 6:00 on July 8th, we did not observe the negative $P_{\rm net}$ rates as we had observed during July 6th (Fig. 4.5a). To calculate $P_{\rm net}$ rates we assumed a constant F_z calculated from a vertical CH₄ concentration profile conducted during the day and average K_z during the sampling campaigns. This assumption is based on small changes on the seasonal $H_{\rm SML}$ estimated based on eight CTD profiles conducted

during day and night (STD_{H_{SML}} = 0.13 m). However, during the nights of the 7th and 8th of July (and not on the 6th), surface water temperature was similar to the temperature at 3 m at around 6:00 – 7:00 suggesting that the SML was deeper than the seasonal H_{SML} (3.1 m, Figs. S.C.5 and 4.5c). During those nights cold air temperatures produced convection that, enhanced by wind, deepened the SML locally (Figs. S.C.4b and 4.5b). This process increases F_z , as well as the total contribution of F_s and R_{dis} due to an increase of A_{sed} and \forall_{SML} respectively. While the latter two sources will be compensated by the increase of \forall_{SML} ($\frac{\partial A_s/\forall_{SML}}{\partial z}|_{H_{SML}} \sim 1$), the increase of F_z is independent of bathymetry of the lake. Therefore, deep convection could have transported CH₄ from below the SML by advection, not accounted by the model, and that in addition of the increase to \forall_{SML} (discussion above), would have produced an overestimation of P_{net} during the night of July 8th.

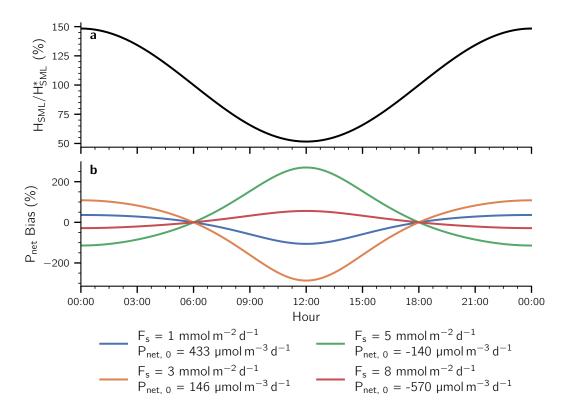


Figure 4.9. Evaluation of change on the surface mixed layer depth $(H_{\rm SML})$ on $P_{\rm net}$ calculations. a Simulated ratio between $H_{\rm SML}$ and the seasonal SML depth $H_{\rm SML}^*$ and surface CH_4 concentration applied to $P_{\rm net}$ simulations. b $P_{\rm net}$ Bias = 1- $P_{\rm net,mod}/P_{\rm net,0}$ where $P_{\rm net,mod}$ and $P_{\rm net,0}$ are the net production rate estimated with variable $H_{\rm net}$ and constant $H_{\rm SML}=3\,\rm m$, respectively.

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4.4.5 Drivers of P_{net} and isotopic analysis in the SML.

 P_{net} rates are the result of OMP and MOx that add and consume CH₄ respectively ($P_{net} = OMP - MOx$). Higher rates during the day than during the night point out that algae and cyanobacteria could be producing CH₄ following the light-dark cycle as shown by Bižić et al. 2020 and Hartmann et al. 2020. It is likely that these production rates are the result of the interaction between reactive oxygen species, iron and methyl donors, enhanced by oxidative stress (Ernst et al. 2022). On the other hand, MOx is inhibited in the presence on light (Thottathil et al. 2019) and elevated O_2 concentrations (Reis et al. 2022). Therefore, increased MOx would likely decrease P_{net} rates during the night rather than during the day. This hypothesis is supported by highly negative P_{net} rates observed during the night that were associated with heavier isotopic signature indicating that the CH₄ in the SML was oxidized during those time periods (Fig. 4.10 inset).

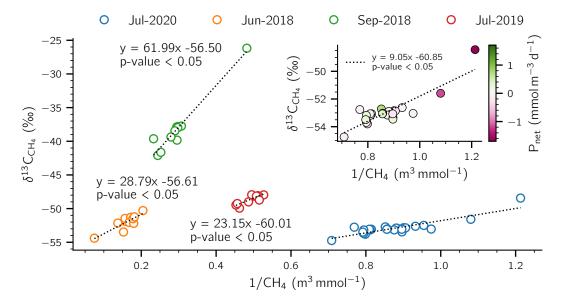


Figure 4.10. Keeling plot analysis of surface $\mathrm{CH_4}$ concentrations. Data from Chapter 2 was included (June and Sept 2018 and July 2019). The inset shows the Keeling plot only with the data of July 2020 where the color shows the $\mathrm{P_{net}}$ associated to the $\mathrm{CH_4}$ concentration.

To analyze the effect of methane oxidation and OMP on $\delta^{13}C_{CH_4}$ in the SML we applied the Keeling plot method to the data from this study and Chapter 2 (June and Sept. 2018 and July 2019). The average P_{net} rates in Chapter 2 in June and Sept 2018 and July 2019 were about 2300, -1940 and $690 \, \mu mol/m^3/d$ respectively (Table 4.2). When P_{net} was negative, enriched CH_4 was measured in the SML indicating the predominance of MOx over

OMP (Table 4.2). In addition, the CH₄ fractionation caused by MOx produced ¹³C enriched CH₄ at lower CH₄ concentrations (case Sep-2018, Fig. 4.10). In contrast, when $P_{net} > 0$, $\delta^{13}C_{CH_4}$ was about 50% and we do not observe clear differences of $\delta^{13}C_{CH_4}$ at different CH₄ concentrations (case June and July 2018 and July 2020). This is likely because OMP will not fractionate the CH₄ pool in the SML and it will produce CH₄ with a constant $\delta^{13}C_{CH_4}$.

We observed a good agreement with the upscaling method proposed in Chapter 2 which relates Chla, light climate (LC) and Secchi disk depth (SD) with the ratio between P_{net} and surface CH_4 concentration (Fig. 4.11), suggesting that OMP can be related with photosynthetic activity by phytoplankton. We observed that from 2018 to 2020, P_{net} can be mostly explained by the changes on LC, where higher LC leads to higher P_{net} rates (r^2 =0.9, Fig. S.C.8). This result shows that, despite different phytoplankton species are able to produce CH_4 in oxic condition (Bižić et al. 2020; Hartmann et al. 2020), the environmental conditions are key to control the CH_4 production rates by these organisms and highlight that CH_4 and Chla concentrations provide more an indication of the overall trophic condition of each lake rather than control the P_{net} rates.

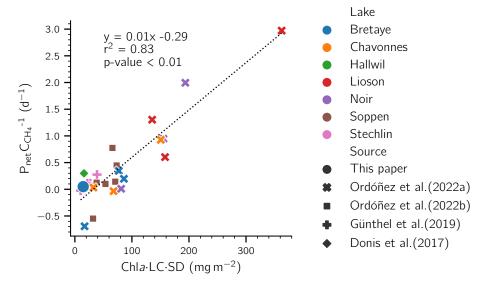


Figure 4.11. Relation between the ration of P_{net} (mmol m⁻³ d⁻¹) and the surface CH_4 concentration (C_{CH_4} , mmol m⁻³) versus chlorophylla (Chla, mg m⁻³) × light climate ($LC=2.5SD/H_{SML}$, -) × Secchi depth (SD, m) proposed in Chapter 2. Only the average data during the day was included.

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4.4.6 Conclusion and implications

In this study, we showed that $P_{\rm net}$ was highly variable on a daily basis, with higher rates during the day than during the night. We demonstrated that in-situ production/consumption needs to be included to explain the change of mass of CH_4 in the SML. On the other hand, the average contribution of $P_{\rm net}$ to the diffusive flux to the atmosphere (F_a) was about 30% (Fig. 4.7). However, in our study we consider the CH_4 production from macrophytes in the littoral zone (Hilt *et al.* 2022) as an anoxic source included in F_s and not as OMP in $P_{\rm net}$, hence the overall $P_{\rm net}$ contribution to F_a could be underestimated.

It is important to note that the potential contribution of CH₄ by macrophytes in the littoral zone, not accounted by the recent studies, could increase the OMP contribution to lake CH₄ emissions particularly in small ponds and coastal shallow zones (Hilt et al. 2022). While more evidence has been collected to show the high contribution of OMP to emissions in different lake environments (Bogard et al. 2014; Donis et al. 2017; Günthel et al. 2019; Hartmann et al. 2020; Chapter 2; Chapter 2), most of these studies were performed during the day when we observed the highest P_{net} rates and higher contribution to F_a (Figs. 4.6 and 4.7). As a first attempt, Bizic 2021 highlight the potential importance of OMP on the CH₄ atmospheric budget. Although it has been shown that OMP can be produced by any organism under oxidative stress and not only related to photosynthesis (Bižić et al. 2020; Hartmann et al. 2020; Ernst et al. 2022), we hypothesize that the OMP contribution to diffusive emissions from lakes may be overestimated, and night measurements have to be performed on aquatic ecosystems to further understand OMP dynamics on both the daily and seasonal scale.

Conclusions and Perspective

For the first time the Intergovernmental Panel on Climate Change (IPCC) established the unequivocal role of humanity on the warming of the atmosphere, land and ocean (IPCC 2021). It is also established that it is *very likely* that the main drivers of global warming are the greenhouse gases (GHG), methane (CH₄) and carbon dioxide (CO₂), and that the increase of these GHGs are unequivocally caused by human activity. The changes on the climate system are unprecedented over the last centuries, many of them are irreversible for centuries to millennia, and it will affect every region in the world. These projections mostly relied on numerical models that combine physical-chemical processes in the climatic system and GHG emissions inventories around the globe.

The interaction between the GHG inventories and the climate system depends on the physical and biochemical processes that produce those emissions. The understanding of these processes and the GHG inventories are constantly updated in the climatic models. For example, in 2007 Cole et al. 2007 demonstrate that inland waters were not passive conducts that simply transported carbon downstream but are active carbon transformers producing GHG emissions. That study resulted in the carbon emissions associated to inland waters to be included in the atmospheric carbon budget. These emissions have been constantly updated since then (Cole et al. 2007; Tranvik et al. 2009; DelSontro et al. 2018b).

After carbon dioxide (CO₂), methane (CH₄) is the second most important carbon based GHG contributing about 20 % to the global radiative forcing (Myhre *et al.* 2013; Saunois

et al. 2020; Rosentreter et al. 2021). A quarter of those CH₄ emissions comes from inland waters, but high uncertainties remain on these estimations (Saunois et al. 2020; Rosentreter et al. 2021). In lakes, CH₄ can is produced in anoxic sediment and emitted through bubbles or by diffusion to the atmosphere (Bastviken et al. 2004). Donis et al. 2017 and Günthel et al. 2019 showed, for the first time, that oxic methane production (OMP) was the most important source of CH₄ emissions in Lake Hallwil contrasting the results from Peeters et al. 2019 where they showed that CH₄ produced in anoxic littoral sediment was the main contributor of the CH₄ emissions in that lake.

Under this context the research goals stated in Chapter 1 have been addressed as follows. In a study on four pre-alpine lakes, Chapter 2 demonstrated that the OMP controversy can be resolved when the correct boundary conditions are utilized. Furthermore, OMP occurred in three of the four studied lakes and $P_{\rm net}$ rates were higher in eutrophic than oligotrophic lakes. Nevertheless, no clear trend was observed regarding the $P_{\rm net}$ contribution to diffusive CH₄ emissions at different trophic states. Chapter 3 took advantage of a four-year data set on a small eutrophic Swiss lake and illustrated that $P_{\rm net}$ rates and its contribution were not constant during the stratified season, but had a tendency to be higher at late spring and then decrease towards the end of the summer. Chapter 4 analyzed the diel variation of $P_{\rm net}$ in a pre-alpine lake, and showed that OMP tended to dominate MOx during the day, and that MOx was higher than OMP during the night. So far, all OMP studies in the literature were conducted during the day, therefore Chapter 4 concludes that, due to the lack of night measurements, the OMP contribution to diffusive CH₄ emissions could be overestimated. Moreover, Chapters 2 to 4 highlight the role of phytoplankton on OMP and identifies light climate, Secchi disk depth, and chlorophyll-a as main drivers.

The following section discusses the contribution of this doctoral thesis to the understanding of the interaction between the changes in physical and biochemical processes due to climate change and their role on the CH₄ cycle in freshwaters. Then, the open research questions based on the findings and limitations obtained during the development of this doctoral thesis are proposed. And finally, this work concludes with a discussion on the estimations of the OMP contribution globally and final remarks.

5.1 Interaction between OMP, eutrophication and climate change

Lakes are sentinels for climate change due to their sensitivity to alterations in their surrounding landscape and climate (Adrian et al. 2009). Lakes are already experiencing rapid changes caused by climate change such as alteration on evaporation, loss of ice cover, warming surface water temperatures and variations in mixing regimes (Woolway et al. 2020). These changes interact with each other, making it difficult to predict their future response to climatic variations and to the direct anthropogenic pressures.

This doctoral thesis identifies light climate (i.e, light availability and SML depth) and Chla as main drivers of P_{net} . As discussed in Chapter 2, P_{net} rates at each lake is driven by light climate (LC), however the different P_{net} rates between lakes are mostly controlled by is trophic states reflected by the surface CH_4 and Chla concentrations. This argument was complemented in Chapter 4 where light climate explained $\sim 90\%$ of P_{net} variation in Lac de Bretaye. Changes in surface water temperature and mixing regimes will affect the SML differently depending on the trophic state of the lake. Oligotrophic lakes will experience a deepening of the SML (Flaim et al. 2016) that, considering no change in light availability, will decrease its light climate. However, since P_{net} becomes independent of LC in oligotrophic lakes (Fig. 5.1), no variation of P_{net} is expected under future climate scenarios. In contrast, eutrophic lakes would experience shallower SMLs with climate change that, considering no change in light conditions, will increase its LC. Since P_{net} strongly increases in eutrophic lakes as function of LC (Fig. 5.1) an increase of P_{net} would be expected in the future for these lakes.

In addition to the effect on the SML, climate warming will increase the ice-free duration in lakes allowing CH₄ to be emitted for a longer period (Guo et al. 2020). The increase of surface water temperature will also enhance littoral production rates of CH₄ in littoral sediments (Bastviken et al. 2008) and OMP rates (Klintzsch et al. 2020). Moreover, longer stratification periods (Woolway et al. 2021) would also allow the occurrence of OMP in the SML for a longer period. Unfortunately, the alteration of the climatic system will stay the same or increase in the next century (IPCC 2021), producing at least the changes on the CH₄ cycle discussed here. As OMP it is a newly recognize source of CH₄ in freshwaters in lakes and oceans, its dynamics

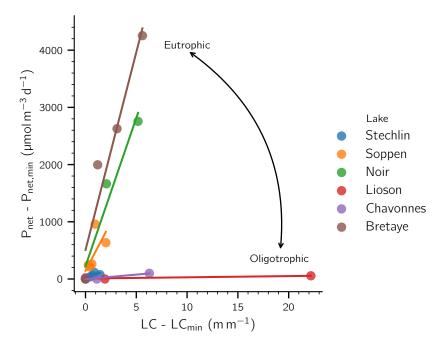


Figure 5.1. Interaction between P_{net} , light climate and eutrophication. Per each lake the minimum P_{net} rate $(P_{net,min})$ and the minimum light climate (LC_{min}) was subtracted to be able to compare the slope of each curve. P_{net} becomes more independent of LC for more oligotrophic lakes.

have not been included in the IPCC model. Despite these new insights of the OMP dynamics in lakes on different temporal scales presented in this doctoral thesis, further investigation needs to be conducted to understand the OMP associated mechanisms and the OMP dynamics in other ecosystems.

5.2 Determination of the boundary conditions in the SML

The methodology proposed in this thesis uses modeling approaches based on the Law of Conservation of Mass. P_{net} rates and their accuracy depend on the precision of the measurements of the sources and sinks of CH_4 in the SML. The high uncertainties found on P_{net} rates calculated in this thesis mostly result from the surface diffusive emissions (F_a) and the diffusive CH_4 flux from the littoral sediments (F_s) . The excellent agreement between the two modelling approaches, the 0-D full-scale mass balance and the 1-D lateral transport model in Chapters 2 to 4, illustrate that the inclusion of the lateral transport did not substantially change the P_{net} estimations and adds robustness to our calculations. Nevertheless, the estimation of F_s remains highly uncertain due to the methodological constraints and spatial heterogeneity of the sediments.

Moreover, very little is known about how the diffusive CH_4 fluxes from the sediment are affected by temperature and the presence of oxygen at the top of the sediment (Damgaard et al. 1998; Liikanen et al. 2002). To decrease the uncertainty of F_s , three different methodologies were used in Chapters 2 and 3 in different locations in each lake, but the temperature and oxygen effects were analyzed but not included in P_{net} calculations due to methodological limitations (see Chapter 4). Although these changes will be not reflected on a daily scale (Chapter 4), laboratory experiments using sediment cores incubations can be used to further investigate the effect of temperature and oxygen on CH_4 fluxes and better estimate P_{net} on a seasonal scale.

Surface diffusive CH_4 emissions to the atmosphere are spatially and temporally variable. Chapter 4 demonstrates that the short-term temporal variability of F_a did not affect the P_{net} calculations, however the average on a daily scale or longer is critical to perform the mass balance in the SML. For this PhD thesis, F_a was measured with the flux chambers approach (McGinnis et al. 2015). Vachon et al. 2010 showed that, using turbulence measurements inside and outside of the chambers, this methodology may overestimate F_a . Nevertheless, good agreement was found with the eddy covariance method (Schubert et al. 2012), which is a non-invasive methodology to estimate F_a . Moreover, flux chambers experiments on streams showed that drifting chambers had very small impact on the turbulence level below the chamber and produce reliable fluxes (Lorke et al. 2015). Hence, turbulence measurements in the surface layer and eddy covariance measurements need to be compared with the flux chamber method to identify key drivers for modelling the mass transfer coefficient (k_{600}) and understand the possible bias produced by the chamber method.

Bizic 2021 highlighted the potential significance that photosynthesis-derived OMP has on global CH₄ emissions. In this thesis, OMP contributions to diffusive emissions were studied across a trophic gradient and at seasonal and diurnal time scales. Chapter 4 shows that OMP did not contribute to F_a during the night. Seasonally, the OMP contribution tended to decrease at the end of the summer in eutrophic lakes, whereas in oligotrophic lakes tended to stay consistent during the stratified season (Chapter 2). Considering the temporal tendency observed in the OMP contribution during this work and that eutrophic lakes tend to emit more CH₄ than the oligotrophic lakes, the OMP contribution from lakes suggested by Bizic 2021 may be overestimated. Nevertheless, most of OMP studies have been conducted in lakes or the

ocean, but the role of OMP in rivers and wetlands is still unknown. Finally, including the role of macrophytes on CH₄ production in oxic conditions can increase the significance of OMP in the atmospheric CH₄ budget.

5.3 Final remarks

This doctoral thesis resolves the OMP dynamics in lakes across trophic state gradients and different temporal scales using physical and biochemical principles. The findings of this study are of significance for resolving the "Methane paradox" globally, however some key points need to be addressed to be able to predict CH₄ emissions from lakes:

- The upscaling method proposed here is a step forward to be able to predict P_{net} rates in different lake ecosystems. In this study LC climate is recognized as the main driver to predict P_{net} in each lake, however the role of trophic variables such as nutrients and algae composition play on OMP is still poorly constrained.
- To understand the P_{net} dynamics of in the SML from diurnal to seasonal scales, it is necessary to accurately isolate each OMP and MOx. This knowledge will also provide a better understanding on the isotopic signature of OMP which could help to determine the sources of CH₄ in surface waters.
- Further investigation is needed regarding the spatial and temporal variations of both CH₄ diffusive fluxes and ebullition from sediments. The variations caused by temperature, oxygen and carbon burial on these fluxes are key to determine its contribution to total CH₄ emissions from inland waters.
- In this thesis I observed a disagreement between the flux chambers gas transfer coefficient and the literature parameterizations. Further research needs to be conducted to understand the impact of solar radiation and microbubbles to improve the parameterizations and therefore accurately estimate diffusive CH₄ emissions to the atmosphere.

Resolving the issues mentioned above and including them to the existing or new numerical models of CH₄ in lakes will improve the estimations of CH₄ emissions from inland waters and it will be useful to predict its response to climate change and eutrophication.

CHAPTER **A**

Supplementary Information Chapter 2

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Table S.A.1. Study sites

Lake	Latitude (°N)	Longitude (°E	Altitude (m.a.s.l)	Max. Depth (m)	Surf. Area (ha)	Trophic State
Bretaye	46.326	7.072	1785	8	4	Hypereutrophic
Chavonnes	46.333	7.085	1692	25	5	Mesotrophic
Lioson	46.386	7.128	1848	28	7	Oligotrophic
Noir	46.327	7.079	1715	9	1	Eutrophic

Table S.A.2. Total (TP) and dissolved phosphorus (DP), dissolved silica (DSIL), dissolved inorganic nitrogen as nitrate plus nitrite (DIN) and total carbon (TC) measurements in the hypolimnion and epilimnion.

Lake	Date	Region	Depth (m)	$ \text{TP} \pmod{\text{mg m}^{-3}}$	$\begin{array}{ c c }\hline DP \\ (mg m^{-3})\end{array}$	$\begin{array}{c c} DIN \\ (g m^{-3}) \end{array}$	$\begin{array}{c c} \text{DSIL} \\ (\text{g m}^{-3}) \end{array}$	$ TC $ $ (g m^{-3})$
	Jun-18	Epilimnion Hypolimnion	0 to 4 5 to 8	20.7 66.7	9.0 18.7	0.02 0.01	0.39 1.89	19.73 26.72
Bretaye	Sept-18	Epilimnion Hypolimnion	0 to 5 5.5 to 7.5	10.0 150.0	7.3 39.3	$\begin{vmatrix} 0.03 \\ 0.02 \end{vmatrix}$	2.77 5.29	20.16 21.00
	Jul-19	Epilimnion Hypolimnion	0 to 5.5 6 to 8.5	33.0 672.0	191.0	$\begin{vmatrix} 0.00 \\ 0.02 \end{vmatrix}$	$\begin{vmatrix} 0.83 \\ 7.36 \end{vmatrix}$	29.13 42.52
	Jun-18	Epilimnion Hypolimnion	0 to 5 6 to 9.2	$\begin{vmatrix} 4.0 \\ 29.7 \end{vmatrix}$	2.3	$\begin{vmatrix} 0.02 \\ 0.05 \end{vmatrix}$	$\begin{vmatrix} 1.02 \\ 3.99 \end{vmatrix}$	29.52 42.43
Noir	Sept-18	Epilimnion Hypolimnion	0 to 5 6 to 9	3.7 13.0	2.7 2.3	$\begin{vmatrix} 0.03 \\ 0.05 \end{vmatrix}$	$\begin{vmatrix} 3.21 \\ 7.60 \end{vmatrix}$	19.03 51.69
Jul-	Jul-19	Epilimnion Hypolimnion	0 to 5 6 to 10	8.0 57.0	BD 5.0	$\begin{array}{ c c } & BD \\ 0.02 \end{array}$	$\begin{vmatrix} 1.22 \\ 8.21 \end{vmatrix}$	33.11 56.48
	Jun-18	Epilimnion Hypolimnion	0 to 15 18 to 26	4.3 26.3	$\begin{vmatrix} 2.0 \\ 3.0 \end{vmatrix}$	$\begin{vmatrix} 0.24 \\ 0.18 \end{vmatrix}$	$\begin{vmatrix} 1.10 \\ 2.61 \end{vmatrix}$	21.22 23.99
Chavonnes Sept-18	Sept-18	Epilimnion Hypolimnion	0 to 10 12 to 22	2.7 10.3	1.0 1.7	$\begin{bmatrix} 0.17 \\ 0.06 \end{bmatrix}$	$\begin{vmatrix} 1.70 \\ 4.06 \end{vmatrix}$	15.91 28.40
	Jul-19	Epilimnion Hypolimnion	0 to 8 10 to 25	BD 9.0	BD BD	$\begin{vmatrix} 0.19 \\ 0.15 \end{vmatrix}$	$\begin{vmatrix} 0.67 \\ 3.37 \end{vmatrix}$	$\begin{vmatrix} 20.77 \\ 25.10 \end{vmatrix}$
	Jun-18	Epilimnion Hypolimnion	0 to 9 13 to 25	3.3 5.0	$\begin{vmatrix} 2.0 \\ 2.7 \end{vmatrix}$	$\begin{vmatrix} 0.13 \\ 0.16 \end{vmatrix}$	$\begin{vmatrix} 1.82 \\ 2.55 \end{vmatrix}$	$\begin{vmatrix} 23.03 \\ 24.05 \end{vmatrix}$
Lioson	Sept-18	Epilimnion Hypolimnion	0 to 14 16 to 26	3.3 2.7	1.0 2.0	$\begin{vmatrix} 0.05 \\ 0.08 \end{vmatrix}$	$\begin{vmatrix} 1.42 \\ 3.26 \end{vmatrix}$	7.66 10.33
	Jul-19	Epilimnion Hypolimnion	0 to 14 16 to 28	BD BD	BD BD	$\begin{vmatrix} 0.07 \\ 0.10 \end{vmatrix}$	$\begin{vmatrix} 1.63 \\ 2.71 \end{vmatrix}$	22.62 24.47

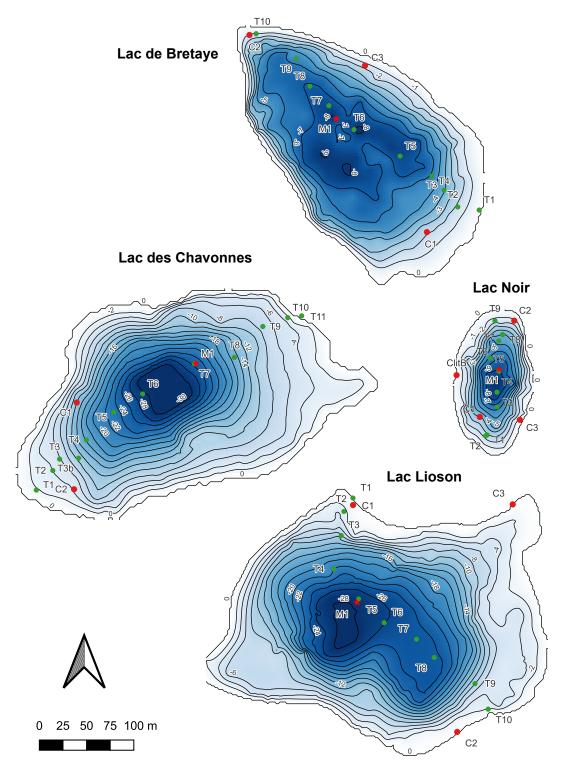


Figure S.A.1. Lake bathymetries obtained from sonar survey, sampling locations for each transect (T#; green points) and sediment sampling points (C#; red dots). M1 (red star) is the location were CTD's, nutrients sampling and CH_4 profiles were performed.

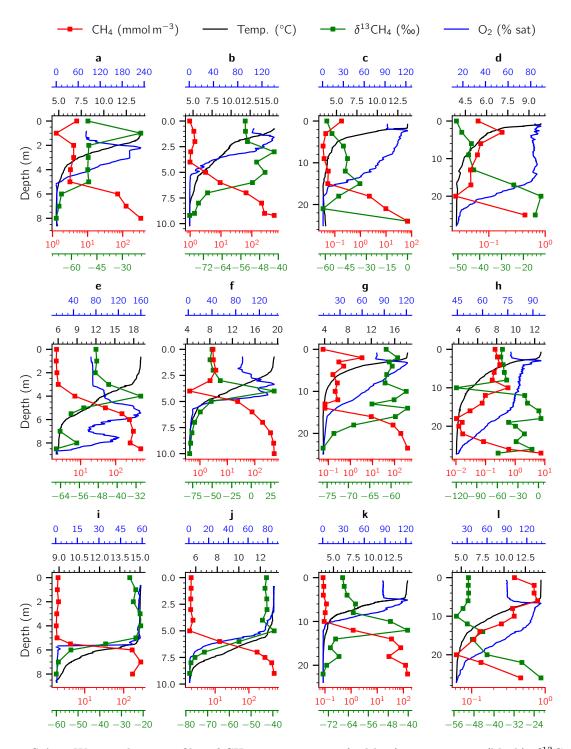


Figure S.A.2. Water column profiles of CH_4 concentrations (red line), temperature (black), $\delta^{13}C_{CH_4}$ (green line) and O_2 saturation (blue). **a-d** June 2018, **e-h** July 2019 and **i-l** Sept 2018 for Lac de Bretaye (BRE), Lac Noir (NOI), Lac des Chavonnes (CHA) and Lac Lioson (LIO).

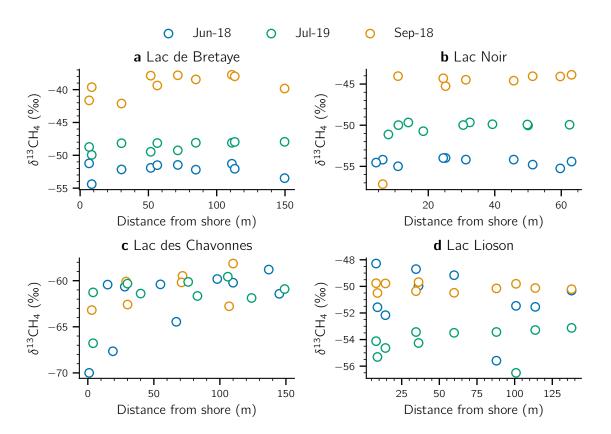


Figure S.A.3. Surface $\delta^{13}C_{CH_4}$ along the transects from shore to the center each lake.

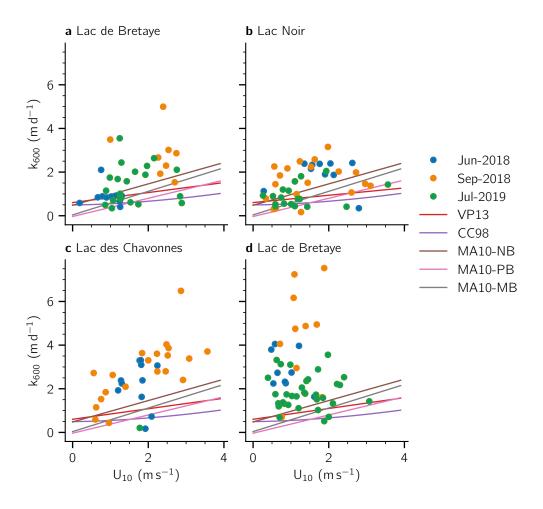


Figure S.A.4. Comparison between chamber-based k_{600} with different wind based parametrizations (CC98: Cole & Caraco 1998; MA10-NB, MA10-PB and MA10-MB: MacIntyre *et al.* 2010 Negative Buoyancy, Positive Buoyancy and Mixed Model; VP13: Vachon & Prairie 2013).

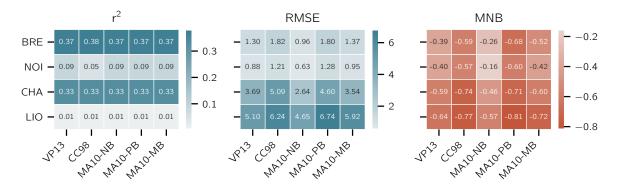


Figure S.A.5. Statistics between chamber based mass transfer coefficient and literature parameterizations. R-square (r²), root mean square error (RMSE) and mean normalized bias (MNB).

Table S.A.3. Summary of diffusive sediment CH_4 fluxes and its associated sampling depth, temperature of overlying water, type of measurement (PW: Porewater and BC: benthic chamber) and zone classification.

Lake	Date	Depth (m)	Zone	T _w (°C)	Core	Type	$F_{\rm s} \pmod{m^{-2}d^{-1}}$
	16/06/18	9	Deep	4.8	M1	PW	5.25
	16/06/18	5.2	Deep	4.8	C1	PW	1.53
	02/09/18	9	Deep	4.8	M1	PW	5.23
D.,	02/09/18	3.5	Shallow	15.3	C1	PW	5.76
Bretaye	02/09/18	8.4	Deep	8.8	M1	PW	12.6
	20/07/19	2.7	Shallow	19.7	C1	PW	0.50
	20/07/19	1.7	Shallow	19.7	C3	BC	12.65
	20/07/19	2.6	Shallow	19.7	C2	BC	12.48
	20/06/18	10.9	Deep	4.5	M1	PW	8.54
	20/06/18	5.5	Deep	5.6	C1	PW	0.22
	03/09/18	10.5	Deep	5.4	M1	PW	6.54
Noir	03/09/18	4.7	Shallow	13.5	C1	PW	2.18
	24/07/19	0.75	Shallow	22.3	ClitB	PW	2.50
	24/07/19	0.75	Shallow	22.3	ClitB	BC	1.62
	24/07/19	0.5	Shallow	22.3	C2	BC	1.33
	24/07/19	1	Shallow	22.3	C3	BC	1.33
	25/07/19	8	Deep	4.4	M1	PW	5.62
	23/07/19	0.5	Shallow	21.2	C1	PW	0.01
Chavonnes	23/07/19	0.6	Shallow	21.2	C1	BC	0.82
	23/07/19	0.5	Shallow	21.2	C2	BC	0.39
	23/07/19	0.2	Shallow	14.7	C1	ВС	0.31
Lioson	23/07/19	0.3	Shallow	14.7	C2	BC	0.16
	23/07/19	0.45	Shallow	14.7	C3	BC	0.27

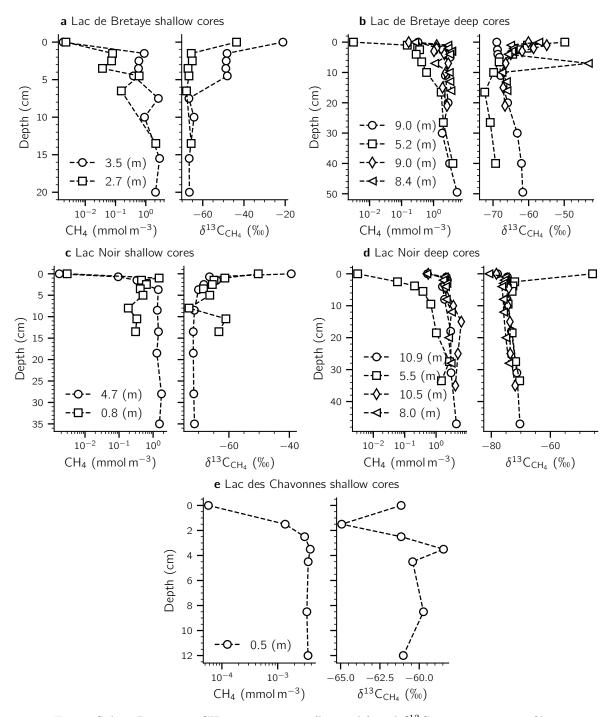


Figure S.A.6. Porewater CH_4 concentration (log scale) and $\delta^{13}C_{CH_4}$ signature profiles.

Table S.A.4. Comparison between average $\delta^{13}C_{CH_4}$ at the surface water and $\delta^{13}C_{CH_4}$ at the top and bottom of the porewater measurement at each lake.

Lake	Date	Surface Water $\delta^{13}C_{CH_4}$ (‰)	Sec $\delta^{13}C_{CH_4 \text{ top }}(\%)$	$\begin{array}{c} \text{diment} \\ \delta^{13} C_{\text{CH}_4 \text{ bottom}} \ (\%) \end{array}$
Bretaye	June 2018 July 2019 Sept 2018	-52.0 -48.8 -38.8	-66.0 -48.4	-66.5 -65.7
Noir	June 2018 July 2019 Sept 2018	-54.5 -49.9 -45.5	-63.9 -66.2	-65.9 -71.6
Chavonnes	June 2018 July 2019 Sept 2018	-62.3 -61.2 -62.4	-62.0	-60.0
Lioson	June 2018 July 2019 Sept 2018	-50.9 -54.0 -50.1		

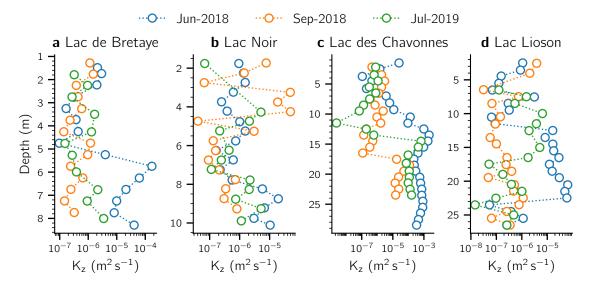


Figure S.A.7. Vertical diffusivity (K_z) for each lake and sampling campaign.

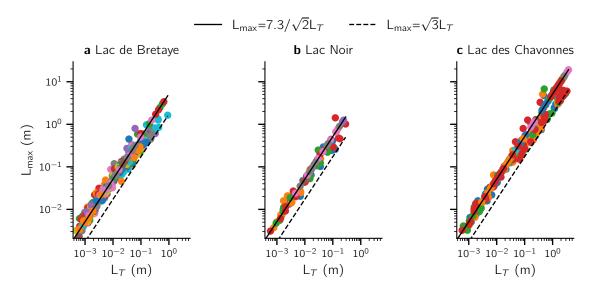


Figure S.A.8. Maximum (L_{max}) versus root-mean square (L_T) displacement length scales within segments of temperatures microstructure profiles from a) Lac de Bretaye, b) Lac Noir and c) Lac des Chavonnes. Each color represent one profile performed on July 2020.

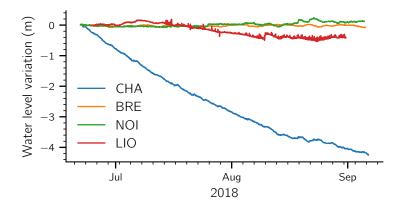


Figure S.A.9. Water level variations from the beginning of 2018 sampling campaign of 2018 for Lac de Bretaye (BRE), Lac Noir (NOI), Lac des Chavonnes (CHA), Lac Lioson (LIO).

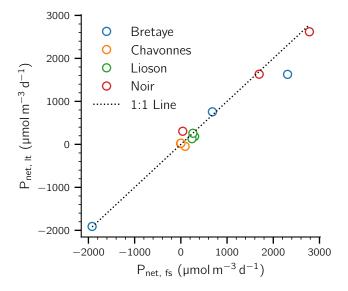


Figure S.A.10. Comparison between $P_{\rm net}$ rates calculated for the full scale mass balance $(P_{\rm net,fs})$ and the lateral transport model $(P_{\rm net,lt})$ for each lake at every sampling date.

Table S.A.5. Volumes and areas used in the full-scale mass balance model. A_z , A_a , A_s are the planar area at the bottom of the surface mixed layer, the surface area and sediment area respectively. \forall_{SML} is the volumne of the surface mixed layer (SML). L is the length scale used to estimate the horizontal dispersion coefficient for the lateral transport model.

Lake	Date	$\begin{array}{c} A_z \\ (m^2) \end{array}$	$\begin{array}{c} A_a \\ (m^2) \end{array}$	$\begin{array}{c} A_s \\ (m^2) \end{array}$	$\forall_{ m SML} \ ({ m m}^3)$	L (m)
Bretaye	Jun 2018	36443	45449	8688	49380	120.28
	Sept 2018	20609	45449	24820	165770	120.28
	July 2019	32832	45449	12552	83425	120.28
Noir	Jun 2018	8012	9960	2026	7509	56.31
	Sept 2018	3057	9960	7139	30778	56.21
	July 2019	6566	9960	3511	14063	56.31
Chavonnes	Jun 2018	54267	60311	5700	71935	138.55
	Sept 2018	28949	44233	14958	166037	118.65
	July 2019	47681	54716	8236	97588	131.97
Lioson	Jun 2018	64659	70052	5129	31557	149.32
	Sept 2018	43222	70052	27800	295316	149.32
	July 2019	57722	70052	12385	89918	149.32

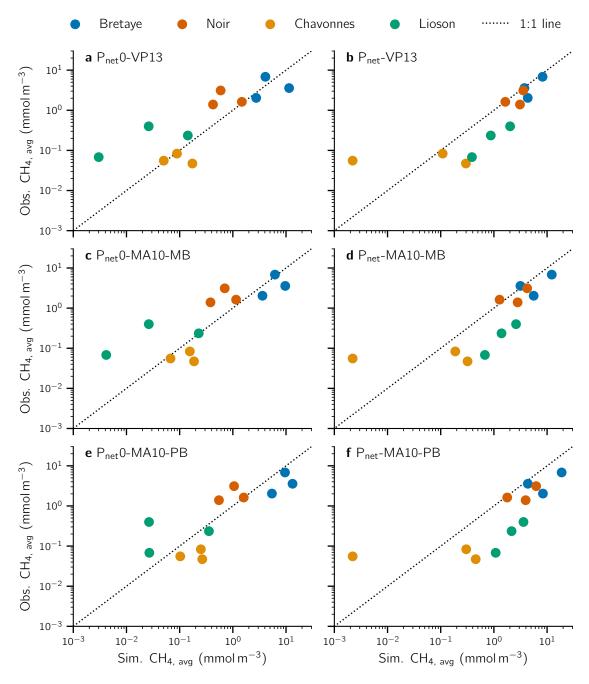


Figure S.A.11. Comparison of observed and simulated average surface CH_4 concentration along transects for each campaign. Simulated CH_4 concentrations were obtained with the lateral transport model using k_{600} for diffusive emissions either with (panels **b**, **d** and **f**) or without $\operatorname{P}_{\rm net}$ (panels **a**, **c** and **e**). The k_{600} was either estimated by Vachon & Prairie 2013 (VP13, panels **a** and **b**), MacIntyre *et al.* 2010 mixed buoyancy (MA10-MB, panels **c** and **d**) and MacIntyre *et al.* 2010 positive buoyancy (MA10-PB, panels **e** and **f**). The statistical results of each panel are shown in Table 2.3.

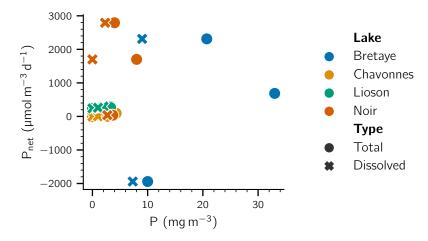


Figure S.A.12. P_{net} rates versus total and dissolved phosphorus concentration (P) at the SML for each lake and sampling campaign.

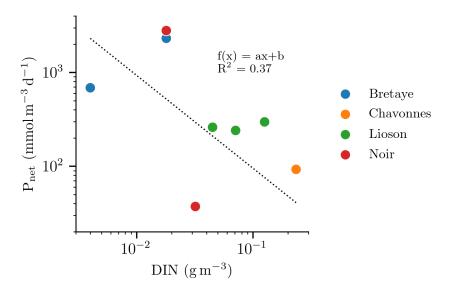


Figure S.A.13. $P_{\rm net}$ versus dissolved inorganic nitrogen (DIN) concentration in the SML for each lake and sampling campaign.

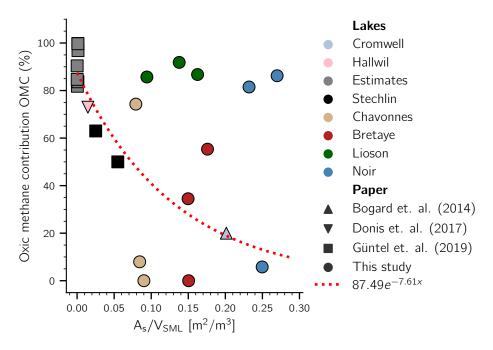


Figure S.A.14. P_{net} contribution to diffusive versus the ratio between the sediment area (A_s) and the epilimnetic volume (V_{SML}) . The trend line $y=87.49e^{7.61x}$ was proposed by Günthel *et al.* 2019.

Supplementary Information Chapter 3

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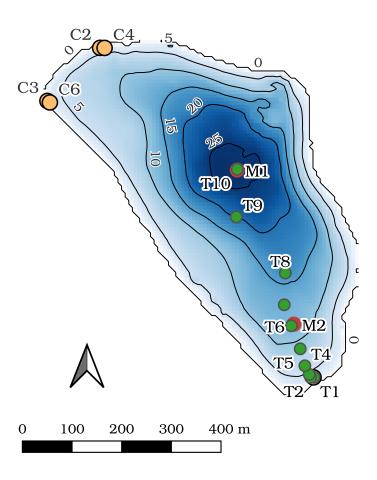


Figure S.B.1. Bathymetry of Soppensee and sampling locations.

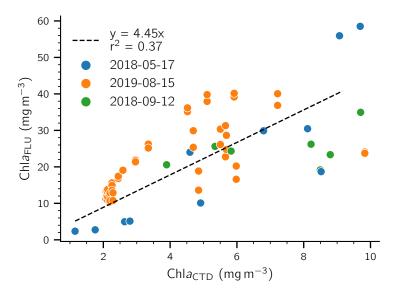


Figure S.B.2. Linear relationship between the chlorophyll-a concentration measured by the fluoroprobe $(Chla_{FLU})$ and the Seabird SBE $(Chla_{CTD})$.

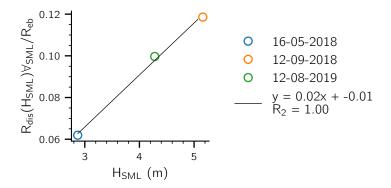


Figure S.B.3. Proportion of total bubble dissolution at the SML of the total ebullition flux versus surface mixed layer depth.

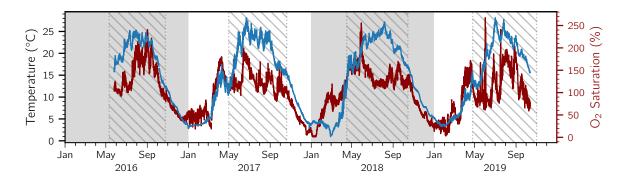


Figure S.B.4. Surface temperature and dissolved oxygen saturation. The shaded areas represent the determined the stratified period for every year.

Table S.B.1. Surface mixed layer depth $(H_{\rm SML})$, Secchi disk depth, littoral sediment area (A_s) , planar area at the bottom of the SML (A_p) and SML volume $(V_{\rm SML})$ during the stratified period. The shaded areas represent the determined stratified period for every year.

Dates	H _{SML} (m)	Secchi depth (m)	$\frac{\mathrm{Chl}a}{(\mathrm{mg}\mathrm{m}^{-3})}$	$\begin{array}{c} A_s \\ (m^2) \end{array}$	$\begin{array}{c} A_p \\ (m^2) \end{array}$	$ m V_{SML} \ (m^3)$
25-May-16	3.52			31,891	222,137	854,361
15-Jun-16	3.44			31,079	222,992	836,993
4-Jul-16	3.03			27,044	227,191	746,529
4-Aug-16	4.86	1.92		46,800	206,417	1,131,748
6-Oct-16	5.98	1.57		61,123	191,768	$1,\!344,\!499$
22-May-17	2.93	4.25		26,091	228,167	724,093
12-Jun-17	1.41	0.95	9.25	12,648	240,236	364,807
$10 ext{-Jul-}17$	2.90	2.10	2.24	25,808	228,456	717,334
2-Aug-17	3.52	3.50		31,891	222,137	854,361
28-Aug-17	4.00	2.70	1.75	36,976	216,759	957,270
12-Oct-17	6.66	2.40		$70,\!520$	$182,\!474$	$1,\!465,\!454$
16-May-18	2.87	1.30	8.73	25,478	228,791	709,432
10-Jul-18	3.12	4.50	1.40	27,912	226,296	766,596
12-Sep- 18	5.16	2.30	7.11	50,479	202,598	1,190,411
12-Aug-19	4.28	1.79	5.86	40,039	213,519	1,014,792
22-Oct-19	7.48	2.20	4.43	82,371	171,066	1,603,283

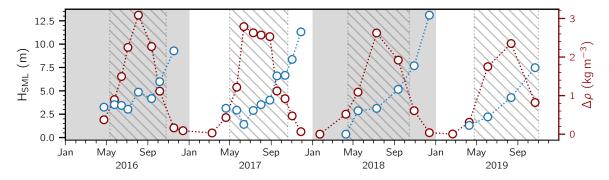


Figure S.B.5. Time series of the surface mixed layer depth $(H_{\rm SML})$ and the difference of average density between the SML and hypolimnion. The shaded areas represent the determined stratified period for every year.

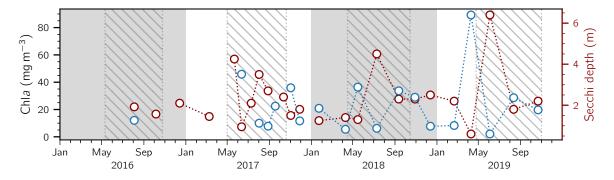


Figure S.B.6. Secchi depths and average chlorophyll-a concentrations (Chla) in the SML. The shaded areas represent the determined stratified period for every year.

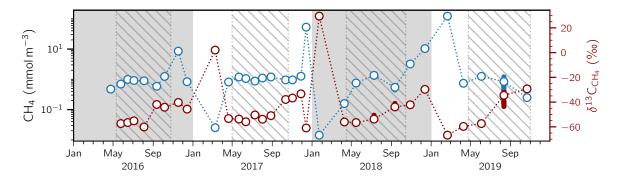


Figure S.B.7. Surface CH₄ concentrations and its isotopic signature ($\delta^{13}C_{CH_4}$) from 2016 to 2019. The shaded areas represent the determined stratified period for every year.

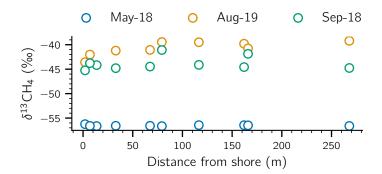


Figure S.B.8. Surface $\delta^{13}C_{CH_4}$ along the transect from shore to center of the lake.

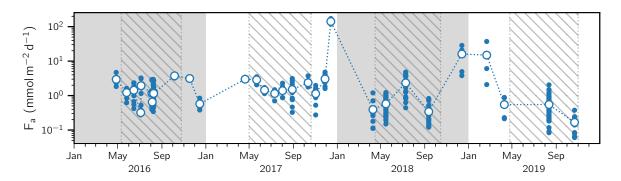


Figure S.B.9. Surface CH₄ diffusive fluxes (F_a) to the atmosphere from 2016 to 2019. The shaded areas represent the determined stratified period for every year.

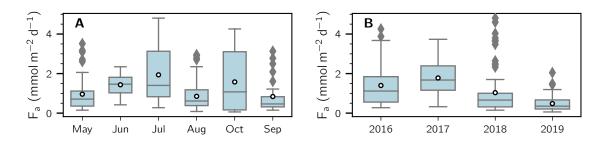


Figure S.B.10. (A) Monthly and (B) yearly average surface CH₄ diffusive fluxes to the atmosphere (F_a) during the stratified season from 2016 to 2019.

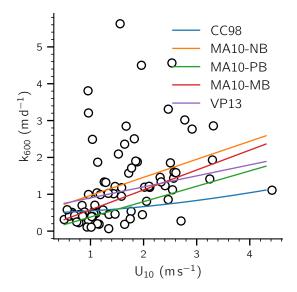


Figure S.B.11. Comparison between chambers based k_{600} (dots) with different wind based literature parameterization (CC98:Cole & Caraco 1998; MA10-NB, MA10-PB and MA-MB: MacIntyre *et al.* 2010 Negative Buoyancy, Positive Buoyancy and Mixed Model; VP13: Vachon & Prairie 2013.

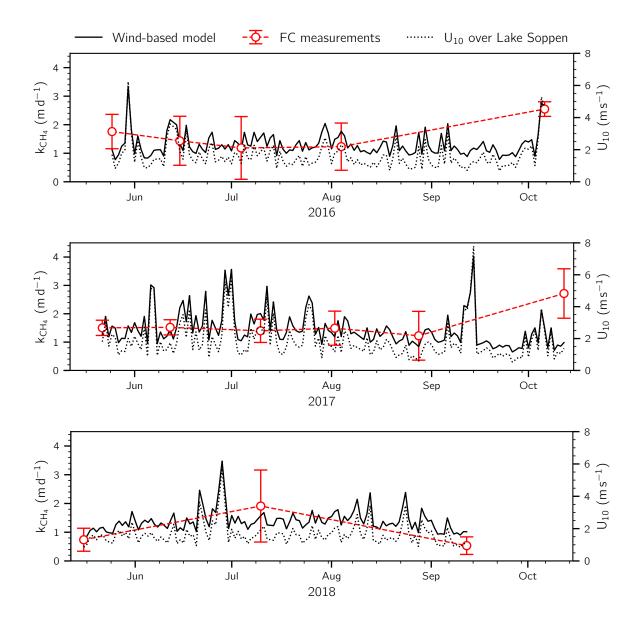


Figure S.B.12. Daily average wind-speed over the lake ($\rm U_{10}$, black dotted line) and CH₄ mass transfer coefficient ($k_{\rm CH_4}$) estimated from wind-based model (black line, MacIntyre *et al.* 2010 negative buoyancy) and chamber measurements estimations (red line) for the stratified period for 2016, 2017 and 2018.

Table S.B.2. Root mean square error (RMSE), determination coefficient (\mathbb{R}^2) and mean normalized bias (MNB) from the comparison between k_{600} parameterization and chamber-based mass transfer coefficient ($k_{600.cb}$).

Parameterization	\mathbb{R}^2	RMSE	MNB	Reference
CC98	-0.26	1.59	0.50	Cole & Caraco 1998
MA10-NB	0.14	1.08	0.0	MacIntyre et al. 2010 Negative buoyancy
MA10-PB	-0.18	1.48	0.50	MacIntyre et al. 2010 Positive buoyancy
MA10-MB	0.04	1.21	0.28	MacIntyre et al. 2010 Mixed buoyancy
VP13	0.09	1.15	0.14	Vachon & Prairie 2013

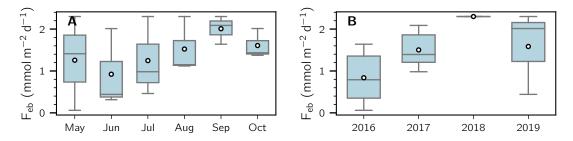


Figure S.B.13. (A) Monthly and (B) yearly average total ebullition flux of CH_4 ($F_{\rm ebul}$) during the stratified season from 2016 to 2019.

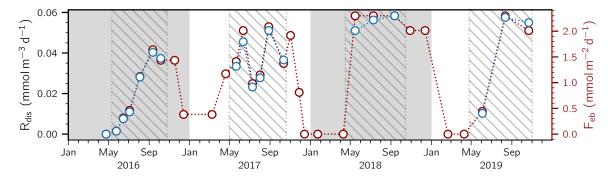


Figure S.B.14. Total ebullition flux $(F_{\rm eb})$ and bubble CH_4 dissolution $(R_{\rm dis})$ in the SML from 2016 to 2019. The shaded areas represent the determined stratified period for every year.

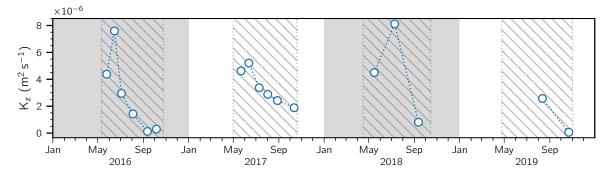


Figure S.B.15. Turbulent diffusion coefficient (K_z) just below the bottom of the SML during the stratified period from 2016 to 2019. The shaded areas represent the determined stratified period for every year.

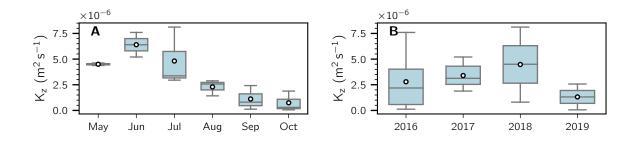


Figure S.B.16. (A) Monthly and (B) yearly average turbulent diffusive coefficient (K_z) just below the bottom of the SML during the stratified season from 2016 to 2019.

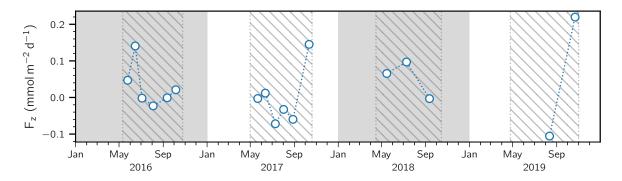


Figure S.B.17. Turbulent diffusive $\mathrm{CH_4}$ flux $(\mathrm{F_z})$ through the bottom of the SML during the stratified period from 2016 to 2019. The shaded areas represent the determined stratified period for every year.

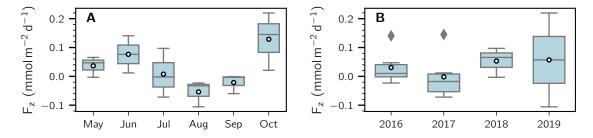


Figure S.B.18. (A) Monthly and (B) yearly average turbulent diffusive CH₄ flux through the bottom of the SML during the stratified season from 2016 to 2019.

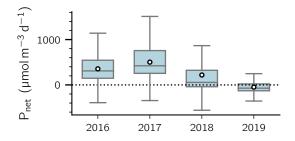


Figure S.B.19. Yearly average $P_{\rm net}$ rates in the SML during the stratified season from 2016 to 2019. The number of day averaged are 6, 7, 3 and 2 for 2016, 2017, 2018 and 2019 respectively.

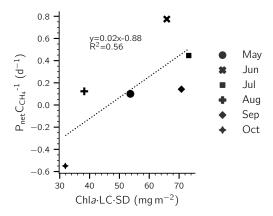


Figure S.B.20. Relation between the ration of P_{net} (mmol m⁻³ d⁻¹) and the surface CH₄ concentration (C_{CH_4} , mmol/m³) versus chlorophyll- $a \times$ light climate ($LC = 2.5 \frac{SD}{H_{SML}}$,-) \times Secchi disk depth (SD, m) proposed by Ordóñez *et al.* 2022b. Each point for were calculated as the monthly average during 2016 to 2019.

Supplementary Information Chapter 4

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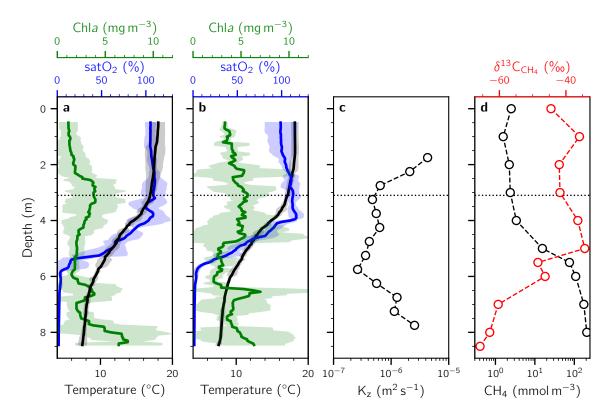


Figure S.C.1. **a** and **b** Water column profile for temperature, chlorophyll-a (Chla) concentration and oxygen saturation (satO₂) for day (n=3, CTDs conducted at 4:50 PM, 11:30 AM at 2:00 PM) and night (n=3, CTDs conducted at 11:15 PM, 6:10 AM and 6:10 PM) conditions, respectively. **c** Average vertical turbulent diffusive coefficient (K_z). **d** Water column profile of CH₄ concentrations and its isotopic signature (δ^{13} C_{CH₄}). The dashed line represents the SML depth (H_{SML}).

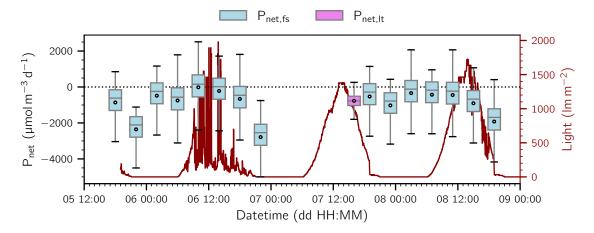


Figure S.C.2. $P_{\rm net}$ estimation using the full-scale mass balance ($P_{\rm net,fs}$, light blue box) and the lateral transport model ($P_{\rm net,fs}$, violet box) using $F_{\rm sed}=8.3\,\rm mmol/m^2/d$. Each box shows the first and third quartiles with the median (line), whiskers extend to the most extreme data point within 1.5 times the interquartile range from the box. The white dot represents the average of the distribution.

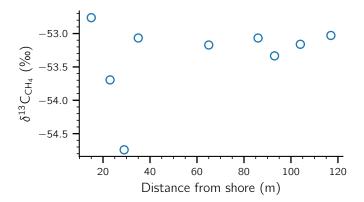


Figure S.C.3. Surface CH_4 isotopic signature along the transect.

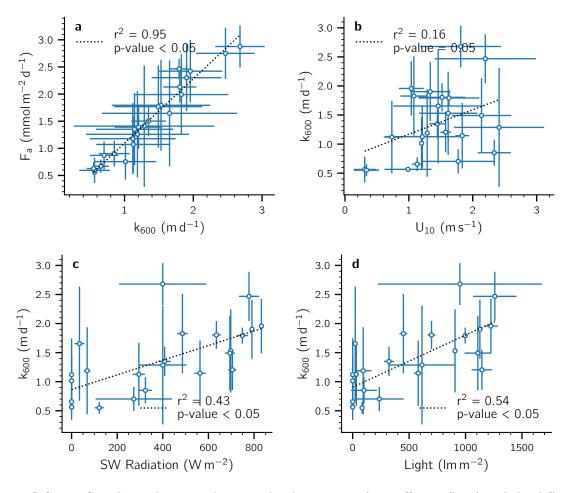


Figure S.C.4. **a** Correlation between the normalized mass transfer coefficient (k_{600}) and the diffusive CH₄ emissions (F_a). **b** Correlation between k_{600} versus **b** wind velocity at 10 m (U_{10}), **c** short wave solar radiation and **d** light intensity at 4 m deep. The data was averaged at each sampling point and whiskers represent the standard deviation.

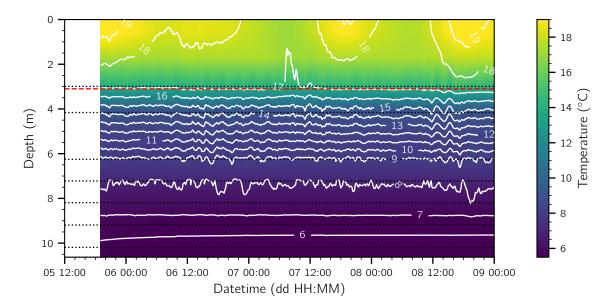


Figure S.C.5. Evolution of the temperature of the water column from 5th to 9th July 2020 at the deepest point of the lake (M1).

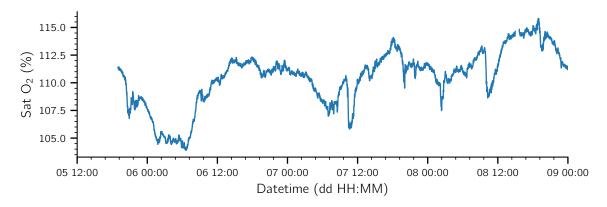


Figure S.C.6. Evolution of surface \mathcal{O}_2 saturation at M1.

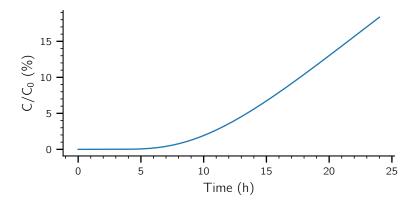


Figure S.C.7. Temporal variation of the percentage of change of the concentration at the center of the lake.

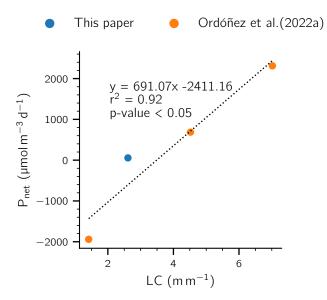


Figure S.C.8. Correlation between average net production rates in the SML (P_{net}) during the day and light climate for the sampling campaigns on July 2020 (this study), June and Sept 2018 and July 2019 (Ordóñez *et al.* 2022b).

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César Ordóñez V.

Education

- 2017 2022 **Ph.D. in Environmental Science**, F.-A. Forel for Environmental and Aquatic Science, University of Geneva, Aquatic Physics Group.
 - Thesis: "Oxic Methane Production Dynamics in Lakes". **Ph.D. supervisor: Daniel F. McGinnis**
- 2012 2014 Civil and Environmental Engineering Diploma, Faculty of Physical and Mathematical Sciences, University of Chile.
 - Professional Engineer diploma in water resources engineering. Graduated Summa Cum Laude,
- 2008 2012 **B.Sc. in Civil and Environmental Engineering**, Faculty of Physical and Mathematical Sciences, University of Chile.

The main focus was on the creation of mathematical skills and the acquisition of basic knowledge in fluid mechanics.

Working Experience

Research Experience

- 2015 2017 Research Assistant, Center for Climate and Resilience Research (CR)2, University of Chile.
 - The goal of this project is to study the transport of urban pollutants flowing into the Andean Cryosphere.
- 2012 2014 Research Assistant, Civil Engineering Dept., Environment and Water Resources Division, Faculty of Physical and Mathematical Sciences, University of Chile.

The aim on this project was to characterize physical and biochemical processes in salty lagoons.

Teaching

- 2021 **Teacher**, Alpine Env. and Society, Faculty of Science, University of Geneva. This course is part of the Master of Sciences in Environmental Sciences program.
- 2020 **Teaching Assistant**, Field Course, Faculty of Science, University of Geneva. Greenhouse gases exchange in alpine lakes.
- 2018 **Teaching Assistant**, Alpine Env. and Society, Faculty of Science, University of Geneva.
 - Greenhouse gases exchange in alpine rivers.

2018 **Teaching Assistant**, Cadagno Field Course, Faculty of Science, University of Geneva.

Greenhouse gases exchange and water column characterization in lakes

2012 Teaching Assistant, Fluid Mechanics, Faculty of Physical and Mathematical Sciences, University of Chile.

Supervision of students projects

- 2019-2020 Co-advisor, Methane fluxes from littoral sediment of a eutrophic swiss lake, Master thesis, Ms. Aurora Pinto.
- 2019-2020 Co-advisor, Sources and fate of aerobic methane production: a study on a small eutrophic lake, Master thesis, Ms. Alexandrine Massot.

Grants

2021 Polar Access Fund, Arctic Change: Resolving methane and CO2 concentrations and their fluxes in Baffin Bay, Swiss Polar Institute, Switzerland.

Languages

Spanish Native

English Advance

Proficient reading, writing, and speaking

Computer skills

Operative Windows, Linux Programming Java (basic), Bash (basic), Git

Systems

Scientific MATLAB, Python, Scilab, R (ba-

Typography Microsoft Office, LATEX

Programming

Aquatic Dyrems (Basic), Caedym (Basic)

Atmospherics WRF Models

Models

Relevant Computing I, Probability and

courses taken Statistics, Numerical Modeling

Publications

Journal Papers

- A. de la Fuente, C. Ordóñez, and R. Pérez. "Diffusional mass transfer coefficient at the water-sediment interface for wind-induced flow in very shallow lagoons". In: Environmental Fluid Mechanics 15.6 (2015). DOI: 10.1007/s10652-015-9437-9.
- C. Ordóñez, A. de la Fuente, and P. Díaz-Palma. "Modeling the influence of benthic primary production on oxygen transport throught the water-sediment interface". In: Ecological Modelling 311 (2015), pp. 1-10. DOI: 10.1016/j. ecolmodel.2015.05.007.
- A. Mazzeo, N. Huneeus, C. Ordóñez, A. Orfanoz-Cheuquelaf, L. Menut, S. Mailler, M. Valari, H. Denier van der Gon, L. Gallardo, R. Muñoz, R. Donoso, M. Galleguillos, M. Osses, and S. Tolvett. "Impact of residential combustion and

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