UNIVERSITÉ DU QUÉBEC À MONTRÉAL

LA DYNAMIQUE DU DIOXYDE DE CARBONE ET DU MÉTHANE DANS UN RÉSERVOIR TROPICAL

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RÉSUMÉ

Les réservoirs sont reconnus comme étant des émetteurs de gaz à effet de serre (GES) vers l'atmosphère, comme le dioxyde de carbone (CO₂) et le méthane (CH₄), pourtant, l'ampleur globale de ces émissions est encore incertaine et débattue. Les GES sont émis à travers plusieurs voies dans les réservoirs (la diffusion en surface, l'ébullition, le dégazage, et les émissions dans l'effluent), cependant, elles ne sont généralement pas toutes prises en compte dans les bilans de carbone des réservoirs. Les flux de GES varient énormément entre les réservoirs, mais aussi au sein d'un même système en fonction de du temps, de l'espace, et de la voie d'émission. Prédire les émissions de carbone à l'échelle de l'écosystème des réservoirs actuels et futurs est donc un défi, et nécessite une bonne compréhension des processus qui sous-tendent les flux de CO_2 et de CH₄. Les flux de GES des réservoirs résultent de la somme des apports horizontaux, des échanges gazeux à l'interface sédiment-eau, du transport vertical des gaz, et du métabolisme dans la colonne d'eau. Ces processus sont rarement quantifiés dans les réservoirs, et sont souvent étudiés séparément. Par conséquent, la connaissance actuelle de la régulation du CO_2 et du CH_4 est fragmentée, n'offrant pas de portrait intégral de l'importance relative des processus et de leurs interactions à l'échelle de l'écosystème. Le but général de cette thèse est de construire un portrait complet des flux de CO_2 et CH4 et des processus qui les sous-tendent ces émissions dans le réservoir Batang Ai situé sur l'ile tropicale de Bornéo. Cet objectif a été réalisé grâce à une combinaison de techniques : échantillonnage sur le terrain, incubations expérimentales, et modélisation.

Les résultats montrent que les flux diffusifs et ebullitifs sont relativement faibles dans le réservoir, dont l'empreinte carbone est principalement due au dégazage et aux émissions dans l'effluent, de CH₄ surtout (Chapitre I). Les flux mesurés contrastent avec les prédictions des modèles actuels qui surestiment la diffusion de surface et l'ébullition et sous-estiment les flux en aval du barrage. La composition du sol, les patrons de sédimentation, le métabolisme net du CO_2 , et la dynamique des GES en eaux profondes ont été identifiés comme éléments clés à investiguer pour améliorer la modélisation des flux de GES des réservoirs. La contribution relative des sources qui soutiennent la diffusion de surface change le long du continuum hydrologique, avec une influence décroissante des apports horizontaux, et diffèrent entre les deux gaz, avec un plus grand apport des sédiments pour le CH₄ versus de la couche profonde pour le CO_2 (Chapitre II). Le métabolisme dans la colonne d'eau joue un rôle majeur, dans le cas du CO_2 et du CH₄, mais est aussi la composante la plus variable et incertaine des budgets épilimnetiques de ces gaz. La modélisation du CO_2 et du CH₄ dans la couche

profonde, qui alimente les émissions en aval, suggère que l'accumulation de gaz s'est stabilisée environ 10 ans après la construction du barrage (Chapitre III). La comparaison entre les données modélisées et mesurées a révélé que la production anaérobique de CO_2 et la méthanogenèse dans la colonne d'eau sont des processus peu connus, mais importants dans la dynamique des GES en profondeur. Les patrons de CO_2 et CH₄ en eaux profondes sont étroitement liés à la concentration en oxygène (O₂), qui a un grand impact sur la respiration aérobique et sur l'oxydation du CH₄. Parmi les paramètres du modèle, la diffusivité verticale des gaz joue un rôle fondamental à travers son effet direct sur le transport des gaz, et son impact sur le profile d'O₂, modulant indirectement les processus métaboliques.

La combinaison des résultats des trois chapitres présente un portrait exhaustif permettant une meilleure compréhension du rôle et de l'interaction des voies d'émissions, des processus, et des facteurs qui régissent la dynamique du CO_2 et du CH_4 à l'échelle de l'écosystème. À travers un échantillonnage spatiotemporel détaillé, l'étude montre la variabilité des composantes mesurées, et révèle leurs chagements le long de la transition hydrologique d'une rivière à un réservoir. Cette recherche fournie des informations manquantes sur le cycle des GES dans un réservoir stratifié en permanence, un trait commun dans les régions tropicales sous étudiées et où la construction de barrages est en forte expansion. Les connaissances acquises ici, particulièrement à travers la modélisation, sont aussi en partie transférables aux réservoirs à plus grande échelle. Finalement, cette thèse identifie les failles dans les connaissances actuelles à combler pour améliorer les modèles et les grandes sources de variabilité et d'incertitude, guidant ainsi les recherches futures sur les GES dans les réservoirs.

Mots clés : dioxyde de carbone, méthane, gaz à effet de serre, réservoir, métabolisme, modélisation, flux.

ABSTRACT

Reservoirs are known to emit greenhouse gases (GHG) to the atmosphere, namely carbon dioxide (CO_2) and methane (CH_4) , yet, the global extent of their emissions is still uncertain and currently debated. In reservoirs, GHG are emitted through various pathways (surface diffusion, ebullition, degassing, and downtream emissions), however, they are usually not all sampled in reservoirs carbon footprint assessments. GHG fluxes vary tremendously among reservoirs, but also within a system as a function of space, time, and flux pathway. This makes it challenging to predict ecosystem scale carbon emissions of current and future reservoirs, a task which requires a good understanding of the processes underlying CO₂ and CH₄ fluxes. Reservoir CO₂ and CH₄ fluxes are shaped as the sum of external horizontal inputs, sediment-water fluxes, vertical gas transport, and water column metabolism. These processes are rarely quantified in reservoirs, and are mostly studied seperately. Hence, current information on CO₂ and CH₄ regulation is fragmented, with no integrated picture of the relative importance of each process and how they vary and interact at an ecosystem scale. The general objective of this thesis is to build a complete portrait of CO₂ and CH₄ fluxes and the processes underlying them in the case study of Batang Ai, a tropical reservoir located on the Borneo Island. This goal was addressed using a combination of field sampling, experimental incubations, and modeling approaches.

Results showed that the reservoir had low to moderate diffusive and ebullitive GHG fluxes, and that its carbon footprint was mainly due to degassing and downstream emissions, especially of CH₄ (Chapter I). Measured flux data contrasted with current predictive models, which overestimated surface diffusion and ebullition and underestimated emissions below the dam. Soil composition, sedimentation patterns, net CO₂ metabolism, and deep water GHG dynamics, were identified as key elements to investigate in order to improve modeling of reservoir GHG emissions. The relative contribution of sources sustaining surface diffusion changed along the hydrological continuum, with a decreasing influence of horizonral inputs, and differed among the two gases, with higher inputs from sediments for CH₄ versus the bottom layer for CO₂ (Chapter II). Water column metabolism played a major role for both CO₂ and CH₄, although it was also the most variable and uncertain component of the epilimnetic gas budgets. Modeling of CO₂ and CH₄ in the deep layer, fueling downstream emissions, suggested that the accumulation of gases had stabilized around 10 years after impoundment (Chapter III). Comparing modeled results to empirical data revealed that anaerobic CO₂ production and water column methanogenesis, despite being poorly known processes, likely play an important role in deep GHG dynamics. Deep CO_2 and CH_4 patterns were tightly linked to oxygen (O₂) concentration, which had a large impact on aerobic respiration and on CH_4 oxidation. Among all model parameters, vertical gas diffusivity had an overarching role via its direct effect on CO_2 and CH_4 transport, and its impact on the O_2 profile, indirectly modulating metabolic processes.

Combining the results of the three chapters provides a comprehensive picture allowing a better understanding of the role and interplay between each flux pathway, process, and factor in shaping CO_2 and CH_4 dynamics at the ecosystem scale. Through an extensive spatial and temporal sampling, the study grasps the variability of measured components, and reveals how transitioning from a river to a reservoir system impacts them. Overall this research provides lacking information on reservoir GHG cycling in a permenently stratified system, a common feature in the understudied tropical regions where reservoir construction is rapidely expanding. Insights derived here, especially from the mechanistic modeling framework, are also partly transferable to reservoirs at large. Finally, the thesis identifies knwoledge gaps to address for improving the modeling framework and tackling the large sources of variability and uncertainty, thus guiding future research on GHG in reservoirs.

Keywords : carbon dioxide, methane, greenhouse gases, reservoir, metabolism, modeling, flux.

INTRODUCTION

0.1 Le contexte actuel de l'empreinte carbone des réservoirs

La construction de réservoirs d'eau douce date de 3000 avant J.-C. et a plusieurs utilités : l'irrigation, l'approvisionnement en eau, le contrôle des inondations, la navigation, et la production d'hydroélectricité. Il y a présentement plus de 58 000 grands barrages (plus de 15 m de haut) recensés à travers le monde, dont 20 % produisent de l'hydroélectricité qui fournit 24 % de la demande globale en électricité (ICOLD, 2019). Le secteur hydroélectrique est en expansion rapide, avec une croissance attendue de 50 à 100 % d'ici 2050 (International Hydropower Association (IHA), 2015). L'hydroélectricité a longtemps été considérée comme une énergie verte, ne produisant pas de gaz à effet de serre (GES). Cependant, au début des années 90, Rudd et al. (1993) suggèrent que la décomposition biologique de la matière organique inondée lors de la construction de réservoirs peut être une source importante de dioxyde de carbone (CO_2) et de méthane (CH_4) vers l'atmosphère. Cette affirmation a déclenché un débat au sein de la communauté scientifique (dos Santos et al., 2006; Fearnside et Pueyo, 2012 ; Fearnside, 2006 ; Rosa et al., 2004), et a mené à la collecte de données sur des réservoirs à travers le monde et à la publication de nombreuses études sur le sujet. Après deux décennies de recherche, la contribution des réservoirs aux émissions de GES fait consensus, mais l'amplitude de ces émissions à l'échelle globale reste incertaine (Deemer et al., 2016). Cette incertitude est due d'une part à un manque de données sur certaines voies d'émissions (ex. émissions en aval) et dans certaines régions (ex. Afrique et Asie du Sud-Est), et d'autre part à une compréhension limitée des mécanismes régulant les émissions. Étant au croisement entre la crise climatique et la demande croissante en eau et en énergie, le sujet de l'empreinte carbone (C) des réservoirs reçoit présentement beaucoup d'attention, non seulement des chercheurs, mais aussi des acteurs du milieu et des médias. Dans ce contexte, il y a actuellement un intérêt marqué à combler le manque d'information concernant la dynamique des GES dans les réservoirs, afin de pouvoir mieux estimer et prédire leur impact sur climat actuel et futur.

0.2 Les sources de CO₂ et CH₄ dans les réservoirs

0.2.1 Le métabolisme aérobique

La production et la consommation de CO_2 dans la colonne d'eau se font principalement à travers le métabolisme aérobique de respiration et production primaire. La respiration aérobique est la transformation de la matière organique en CO₂ via la consommation d'oxygène (O₂). Dans les lacs, les taux de respiration aérobique varient grandement, allant de 0.029 à 4800 mmolO₂.m⁻³.d⁻¹ (Pace et Prairie, 2005 ; Pollard, 2013). La respiration dépend de la disponibilité et de la labilité de son substrat : la matière organique (Amon et Benner, 1996; Cammack et al., 2004). En effet, les taux de respiration aquatiques varient selon un gradient de statut trophique et de concentration en C organique dissout; des indicateurs de la productivité interne et des apports externes de matière organique (Pace et Prairie, 2005). La labilité de la matière organique peut être influencée par le ratio de ces deux sources et par la photooxydation, processus par lequel les radiations solaires dégradent la matière organique en molécules plus labiles et faciles à respirer (Cory et Kling, 2018). La respiration est sensible à la température (Carignan et al., 2000), avec des taux généralement plus élevés dans les climats chauds (Lewis, 2010). Dans les réservoirs, la respiration est un processus majeur de décomposition de la matière inondée surtout dans les premiers temps après la construction des barrages (Abril et al., 2005 ; Teodoru et al., 2012 ; Thérien et Morrison, 2005). La consommation d' O_2 par la respiration est aussi responsable de la création de conditions hypoxiques et anoxiques surtout dans les eaux profondes des lacs et réservoirs.

A l'inverse, la production primaire (principalement la photosynthèse) produit de l'O₂ et de la matière organique, et constitue un puit de CO_2 majeur dans les systèmes aquatiques. La photosynthèse dépend de la lumière, de la température, et de la disponibilité des nutriments (surtout l'azote et le phosphore). Ainsi, les eaux tropicales ont en moyenne des taux de productivité primaire plus élevés (Amarasinghe et Vijverberg, 2002 ; Lemoalle, 1981 ; Likens, 1975). Bien que les facteurs contrôlant la production primaire soient les mêmes dans les lacs et les réservoirs, des différences ont été notées entre ces deux types de systèmes (Jones et Knowlton, 1993 ; Lind *et al.*, 1993 ; Walker, 1984). Les réservoirs sont généralement caractérisés par une forte influence des apports riverains et par des zones de transition hydrologiques d'un système lotique à lentique (Kimmel et Groeger, 1984). Cette configuration crée un gradient horizontal de nutriments, de turbidité, et d'atténuation de la lumière qui favorise une grande hétérogénéité spatiale de la productivité primaire dans les réservoirs comparés aux lacs (Kimmel, Bruce L. *et al.*, 1990 ; Knoll *et al.*, 2003).

0.2.2 La production et consommation de CH₄

La production de CH₄ se fait par méthanogenèse, une forme de respiration anaérobique performée par des microorganismes méthanogènes du domaine des archées. Deux types de méthanogenèse existent : la fermentation des acétates qui produit du CO₂ en plus du CH₄, et l'oxydation du CO₂ qui le consomme (Demirel et Scherer, 2008). Bien qu'il n'y a pas de consensus sur le sujet, la fermentation des acétates semble être la voie de méthanogenèse dominante dans les eaux douces (Conrad, 2005). La méthanogenèse est une réaction très couteuse en énergie, elle est donc principalement active dans des conditions biochimiques réduites, où d'autres accepteurs d'électrons comme l'O₂ ne sont pas disponibles (Ferry, 2012). La méthanogenèse se produit donc surtout dans les couches anoxiques des lacs et réservoirs, majoritairement dans les sédiments, mais a aussi été observé dans les eaux profondes (Iversen *et al.*, 1987 ; Lliros *et al.*, 2010 ; Musenze *et al.*, 2016). Cependant, il y a de plus en plus de preuves que la méthanogenèse se produit aussi dans les eaux oxygénées (dans des niches microanoxiques), mais son importance à l'échelle écosystémique reste peu connue (Bogard *et al.*, 2014 ; Donis *et al.*, 2017 ; Grossart *et al.*, 2011 ; Tang *et al.*, 2014). Le taux de méthanogenèse augmente avec la quantité et la labilité de la matière organique (Kelly et Chynoweth, 1981 ; West *et al.*, 2012), la production de CH₄ est donc très active dans les réservoirs nouvellement inondés (Matthews *et al.*, 2005). La méthanogenèse est aussi très sensible à la température (Yvon-Durocher *et al.*, 2014), ce qui en fait un processus particulièrement important dans les réservoirs tropicaux (Marotta *et al.*, 2014).

L'oxydation du CH₄ est un processus microbien qui consomme le CH₄ en tant que source de C et relâche du CO₂. Cette réaction est particulièrement importante en termes de GES car elle transforme le CH₄, un GES puissant, en CO₂ qui a un potentiel de réchauffement atmosphérique 32 fois moins élevé sur un horizon de 100 ans (Ciais *et al.*, 2013). L'oxydation du CH₄ se produit en grande majorité dans les environnements oxiques et hypoxiques, mais peut dans une moindre mesure se produire dans les environnements anoxiques (Valentine, 2002). Les taux maximaux d'oxydation du CH₄ sont retrouvés dans les zones hypoxiques et riches en CH₄, comme à l'interface eausédiment ou dans la zone metalimnétique (Bastviken *et al.*, 2002 ; Kankaala *et al.*, 2006 ; Thottathil *et al.*, 2019 ; Zigah *et al.*, 2015). Cependant, ce n'est pas clair si les organismes méthanotrophes préfèrent les zones hypoxiques à cause de l'effet direct de l'O₂ sur leur métabolisme (Rudd et Hamilton, 1978) ou dû à l'abondance de CH₄ dans ces zones (Borrel *et al.*, 2011). L'oxydation du CH₄ est optimale à 25 °C mais est active dans une large gamme de températures (Chowdhury et Dick, 2013). L'oxydation du CH₄ est moins active dans les environnements photiques dus à un effet d'inhibition de

la lumière (Murase et Sugimoto, 2005 ; Shelley *et al.*, 2017). La proportion de CH₄ oxydé diffère d'un système à l'autre, dépendamment de l'interaction entre les facteurs, dont la communauté méthanotrophique (Reis *et al.*, 2020b) et la structure physicochimique de la colonne d'eau en termes de luminosité, stratification thermale et profile d'oxygène (Thottathil *et al.*, 2018 ; Vachon *et al.*, 2019).

0.2.3 Le bilan à l'échelle de l'écosystème

Les concentration de CO₂ et CH₄ dans l'eau sont déterminées par la balance entre les processus métaboliques internes, et par les apports externes. Dans le cas du CO₂, la balance entre la respiration aérobique et la production primaire détermine si un système est autotrophe ou hétérotrophe (net consommateur ou producteur de CO₂). La tendance globale suggère que l'hétérotrophie est plus fréquente dans les systèmes moins productifs (oligotrophes) (Duarte et Agustí, 1998) et plus riches en matière organique dissoute (Bogard et al., 2020), notamment d'origine externe, qui favorisent la respiration bactérienne tout en limitant la disponibilité de la lumière et des nutriments pour la production primaire (Giorgio et al., 1999). À l'inverse, la production primaire excède généralement la respiration dans les systèmes très productifs, où les bactéries sont plus sujettes à la prédation et à la lyse (Cotner et Biddanda, 2002 ; Roberts et Howarth, 2006). Bien que les systèmes autotrophes consomment du CO₂, leur haute productivité a tendance à promouvoir la sédimentation de C organique (Evans et al., 1997) favorisant la production de CH_4 dans les zones benthiques (Adams, 2005; Almeida et al., 2016; Beaulieu et al., 2019; Sobek et al., 2012). Le CH₄ produit peut être oxydé en CO₂ avant d'être émis vers l'atmosphère. La balance entre la production et la consommation du CH4, bien que très peu étudiée (DelSontro et al., 2018b), dépend d'une part de l'O₂ qui est fortement associé au métabolisme du CO₂, et d'autre part de facteurs physiques comme la lumière (Thottathil et al., 2018) et la stratification thermale (Vachon *et al.*, 2019). Ainsi, les dynamiques du CO₂, du CH₄, et de l'O₂ sont étroitement liées, et sont modulées par des les traits physiques des lacs et réservoirs.

Les flux de CO_2 et CH_4 des systèmes aquatiques vers l'atmosphère ne reflètent pas seulement le métabolisme interne du C, mais aussi les apports externes de gaz. L'amplitude de ces apports dépend de la biogéochimie et de l'hydrologie du bassin versant, soit de la quantité de C présente dans les sols et de sa capacité à être lixiviée vers les systèmes aquatiques. La contribution des apports externes aux émissions aquatiques de CO_2 et CH_4 varie de négligeable à majeure (Bogard *et al.*, 2019 ; Rasilo *et al.*, 2017 ; Wilkinson *et al.*, 2016), et a tendance à augmenter avec la connectivité sol - eau, donc a diminué avec la taille du système (Hotchkiss *et al.*, 2015 ; Humborg *et al.*, 2010 ; Lapierre et del Giorgio, 2012). Les apports externes de CO_2 et CH_4 sont souvent liés, non seulement dû à leur source commune, mais aussi dû au fait que l'oxydation du CH_4 à l'interface sol - eau peut alimenter les émissions de CO_2 dans systèmes aquatiques (Lupon *et al.*, 2019 ; Rasilo *et al.*, 2014).

Les sources de CO_2 et CH_4 dans les réservoirs, tout comme dans les lacs, sont régulées par l'interaction entre les propriétés physiques, chimiques, et biologiques du système et de son bassin versant. À l'échelle de l'écosystème, ces interactions restent mal connues dans les lacs, et encore moins dans les réservoirs. En ce sens, l'impact de la construction de réservoirs sur la dynamique aquatique des GES est encore un sujet de recherche actif (Maavara *et al.*, 2017).

0.3 Les émissions de GES des réservoirs

0.3.1 Les voies d'émissions

Les réservoirs émettent des GES vers l'atmosphère à travers plusieurs voies: la diffusion de surface, l'ébullition, le dégazage, et l'émission en aval. La diffusion de surface est l'échange de gaz entre l'air et l'eau à la surface du réservoir. Ce flux dépend d'une part du gradient de concentration entre les deux fluides (air et eau) qui détermine la direction et l'amplitude du flux de gaz, et d'autre part de la turbulence à l'interface air-eau qui détermine la vitesse de l'échange gazeux. La diffusion de surface est donc influencée par les sources et puits des GES dans la couche d'eau de surface. Cette voie d'émission est la plus documentée dans les réservoirs (et dans les lacs naturels) car elle est relativement facile à mesurer.

L'émission de GES par ébullition résulte de la formation de bulles de gaz dans les sédiments. Ces bulles sont composées principalement de CH₄, un gaz beaucoup plus enclin à former des bulles comparativement au CO₂ et à l'oxyde nitreux (N₂O) dû à sa faible solubilité dans l'eau. La formation et l'ascension de bulles de GES des sédiments vers l'atmosphère sont contrôlées par la quantité de gaz produite et par la pression hydrostatique. L'ébullition est donc généralement restreinte aux zones littorales où la température est plus élevée (plus grande production) et où la pression hydrostatique est plus faible (DelSontro *et al.*, 2016a ; Wik *et al.*, 2014). L'ébullition est particulièrement importante en termes de GES car elle permet au CH₄ d'atteindre directement l'atmosphère sans être oxydé en CO₂ dans la colonne d'eau (ce qui réduirait son potentiel de réchauffement climatique). L'ébullition est connue pour son extrême hétérogénéité spatiale et temporelle (Grinham *et al.*, 2011), ce qui rend le flux ebullitif de GES difficile à estimer à l'échelle d'un écosystème. Néanmoins, vu son importance dans les bilans aquatiques de GES (Deemer *et al.*, 2016), l'ébullition est de plus en plus mesurée dans les réservoirs.

Le dégazage représente le flux de gaz vers l'atmosphère dû au relâchement de pression lors de la décharge d'eaux profondes vers la surface. Ce cas se retrouve particulièrement dans les réservoirs hydroélectriques où les turbines sont alimentées par des eaux profondes souvent riches en CO₂ et CH₄. A cause de à sa faible solubilité, le CH₄ est particulièrement propice au dégazage. L'accumulation de GES dans les eaux profondes est surtout liée à la présence d'une stratification thermale, qui limite les échanges entre la couche profonde et la surface du réservoir. Le dégazage est principalement associé à la dynamique des GES dans les eaux profondes. C'est aussi le cas pour les émissions de GES en aval du réservoir. En effet, les GES produits à l'intérieur du réservoir (souvent dans les eaux profondes) peuvent être émis à la surface de la rivière effluente sur plusieurs dizaines de kilomètres en aval du barrage (Abril et al., 2005). Les émissions de CH₄ en aval peuvent être atténuées par l'oxydation en CO₂ (DelSontro et al., 2016b; Guérin et Abril, 2007; Kemenes et al., 2007). Les processus en aval du barrage (dégazage, émission de la rivière, et oxydation du CH₄) sont beaucoup moins documentés que les émissions par diffusion en surface et par ébullition (Deemer et al., 2016). Pourtant, le dégazage et les émissions en aval sont dans plusieurs cas des sources majeures d'émission de GES des réservoirs (Galy-Lacaux et al., 1997; Kemenes et al., 2007).

0.3.2 La variabilité temporelle

À l'échelle journalière, la concentration et les flux de GES des réservoirs varient principalement en fonction de la lumière, de la température, et de la vitesse du vent. Ces facteurs physiques ont un impact direct sur l'échange gazeux diffusif et ebullitif (Grinham *et al.*, 2011 ; Keller et Stallard, 1994 ; Sellers *et al.*, 1995), et indirect sur la balance métabolique, ce qui cause des variations à court terme de CO₂ et d'O₂ particulièrement dans les systèmes productifs (Demarty *et al.*, 2009 ; Maberly, 2008). Des variations journalières de CH₄ ont aussi été observées dû à l'apport d'O₂ de la photosynthèse stimulant la méthanotrophie durant le jour (Bergier *et al.*, 2011). Les variations saisonnières des flux de GES des réservoirs sont influencées par la température, la stabilité de la colonne d'eau, et l'hydrologie. Dans les réservoirs tropicaux, la température est relativement stable dans le temps, mais les précipitations peuvent induire une forte augmentation du ruissèlement et du niveau des eaux durant la saison des pluies. Cette saison est généralement associée à des émissions moins élevées de CO₂ et CH₄ des réservoirs dû à un temps de résidence plus court (Abril *et* al., 2005; Bastien et Demarty, 2013; Galy-Lacaux et al., 1999; Kemenes et al., 2007; Roland et al., 2010). Cependant, Pacheco et al., (2015) a rapporté des émissions plus élevées lors de périodes de rétention courtes dues à l'augmentation des apports riverains et du mélange des eaux. Ceci souligne l'importance des patrons hydrologiques spécifiques à chaque système et de leur interaction avec les facteurs biochimiques. Dans les régions tempérées et boréales, l'été est caractérisé par la stratification thermale des eaux, qui tend à réduire les flux diffusifs de GES en empêchant le CO₂ et le CH₄ de l'hypolimnion d'atteindre la surface (Encinas Fernández et al., 2014). D'un autre côté, les hautes températures estivales stimulent la production de CH4 dans les sédiments littoraux (DelSontro et al., 2010) et changent la balance métabolique souvent vers une diminution ou même une absorption du CO₂ (Adamczyk et Shurin, 2015 ; Wang et al., 2011). Lors du mélange des eaux, au printemps et en automne, les GES stockés sous la glace ou dans l'hypolimnion sont relâchés produisant un pic d'émission de CO₂ et CH₄ vers l'atmosphère (Encinas Fernández et al., 2014 ; Huotari et al., 2009; Riera et al., 1999; Rudd et Hamilton, 1978). Dans les réservoirs, la stratification saisonnière influence la concentration de GES dans la colonne d'eau, ce qui affecte le dégazage et les émissions en aval (Kemenes et al., 2007).

Sur le long terme, les émissions de CO_2 des réservoirs suivent une tendance généralisée avec des émissions maximales durant les premières années du réservoir dû à la décomposition de la matière nouvellement inondée, suivie d'une décroissance exponentiellement des flux avec le temps (Abril *et al.*, 2005 ; Barros *et al.*, 2011 ; Ometto *et al.*, 2013 ; St. Louis *et al.*, 2000 ; Teodoru *et al.*, 2012). Les émissions de CO₂ se stabilisent généralement dix an après l'inondation, devenant comparables aux émissions lacustres dans certains cas, mais restant plus élevées dans d'autres (Åberg *et al.*, 2004 ; Abril *et al.*, 2005 ; St. Louis *et al.*, 2000 ; Teodoru *et al.*, 2012 ; Tremblay *et al.*, 2005). L'évolution temporelle des émissions de CH₄ est plus variable étant parfois très élevée dans des réservoirs âgés (DelSontro *et al.*, 2010 ; Demarty *et al.*, 2011 ; Rosa *et al.*, 2003 ; Teodoru *et al.*, 2012).

Indépendamment de l'âge, les émissions des réservoirs varient aussi d'une année à l'autre selon les conditions climatiques (Lima, 2005; Musenze *et al.*, 2014). Globalement, les données de suivi à long terme sont rares dans les réservoirs, ce qui limite la compréhension de la dynamique des GES sur de longues échelles temporelles dans ces systèmes.

0.3.3 La variabilité spatiale

Les profils verticaux de GES dans les lacs et réservoirs sont le résultat de l'interaction entre les processus biologiques et physiques dans la colonne d'eau. La concentration du CO_2 dépend de la balance métabolique entre la respiration et la photosynthèse (Giorgio *et al.*, 1999). Cependant, ces deux processus peuvent être découplés dans l'espace, la photosynthèse étant contrainte à la zone phototique (couche de surface) et la respiration étant stimulée par l'abondance de matière organique dans les eaux profondes et les sédiments (Pace et Prairie, 2005). Ce patron est typiquement observé dans les réservoirs stratifiés, où l'hypolimnion est hautement concentré en CO_2 et CH₄ provenant de la décomposition de la matière organique inondée (Abril *et al.*, 2005 ; Chanudet *et al.*, 2011 ; Kemenes *et al.*, 2007). La présence d'une stratification des eaux restreint les échanges entre la couche profonde et celle de surface, ce qui amplifie les gradients verticaux d'O₂ et de GES. Le métalimnion est une couche de transition où les concentrations intermédiaires d'O₂ sont propices à l'oxydation du CH₄. Cette couche agit donc comme un filtre qui empêche le CH₄ d'atteindre la couche de surface (Abril *et al.*, 2005 ; Chanudet *et al.*, 2011 ; Galy-Lacaux *et al.*, 1997 ; Rudd et Hamilton, 1978). La stratification thermique module les échanges gazeux, ce qui influence l'oxydation du CH₄ (Vachon *et al.*, 2019), et les profils verticaux de CH₄ et CO₂.

Dans les réservoirs, les concentrations et flux de GES à la surface présentent une grande hétérogénéité spatiale, surtout dans le cas du CH₄ (Grinham *et al.*, 2011). Les concentrations et émissions de CH₄ sont généralement plus élevées en zone littorale due à la présence de sédiments plus chauds et moins profonds, ce qui stimule la production et l'ébullition de CH₄ (Hofmann *et al.*, 2010 ; Joyce, 2003 ; Sherman et Ford, 2011). Les zones riches en matière organique, comme les embouchures d'affluents, les rejets d'égouts, les aquacultures, et les zones fertiles nouvellement inondées, sont souvent associées à une forte production et émission de CH₄ et de CO₂ (Chen *et al.*, 2011 ; da Silva *et al.*, 2018 ; DelSontro *et al.*, 2011 ; Musenze *et al.*, 2014 ; Roland *et al.*, 2010 ; Sherman et Ford, 2011 ; Teodoru *et al.*, 2011 ; Zhao *et al.*, 2013). La variabilité spatiale du CO₂ en surface est aussi dépendante du métabolisme interne du système, ainsi, les zones de forte production primaire coïncident souvent avec des flux de CO₂ bas voir négatifs (Abril *et al.*, 2005 ; Chanudet *et al.*, 2011 ; Demarty *et al.*, 2009 ; Pacheco *et al.*, 2015 ; Rudorff *et al.*, 2011).

0.3.4 Les patrons à l'échelle globale

Les flux de GES mesurés dans les réservoirs varient globalement entre -623 et 7206 mgC.m⁻².d⁻¹, allant de l'absorption de C à des émissions comparables à celles des énergies fossiles (Demarty *et al.*, 2009 ; Giles, 2006 ; Tamooh *et al.*, 2013). Cette variabilité est causée par des facteurs agissant à différentes échelles, locales à globales. La disponibilité de la matière organique est un facteur central dans ce contexte, étant le substrat pour la production de CO₂ et CH₄. Ainsi, plusieurs études rapportent une corrélation directe entre la concentration en matière organique et les émissions de C à larges échelles (Barros *et al.*, 2011 ; Bastien et Demarty, 2013 ; Bergström *et al.*, 2004).

D'autres études ont identifié l'effet de variables comme l'âge du réservoir, la topographie du bassin versant, et l'utilisation du territoire comme modulateurs des flux de surface à travers leur effet indirect sur la quantité et la qualité de la matière organique dans les réservoirs (Adamczyk et Shurin, 2015; Barros et al., 2011; Bastien et Demarty, 2013 ; Bergström et al., 2004 ; Demarty et al., 2009, 2011 ; Musenze et al., 2014 ; Ometto et al., 2013 ; Roland et al., 2010 ; St. Louis et al., 2000). La température est un autre facteur régional influençant la production et les flux de GES, ce qui se traduit par des émissions en moyenne plus élevée dans les réservoirs tropicaux par rapport aux régions tempérées et boréales (Barros et al., 2011; Bastien et Demarty, 2013 ; Musenze et al., 2014 ; St. Louis et al., 2000 ; Tremblay et al., 2005). À l'échelle locale, la morphométrie joue un rôle particulièrement important, avec de plus fortes émissions de CH₄ dans les systèmes de petite taille et peu profonds, une configuration qui favorise l'ébullition et restreint le potentiel d'oxydation du CH₄ (Barros *et al.*, 2011; Kankaala et al., 2013a ; Lima, 2005). Une analyse globale a montré que 40 et 54 % de la variabilité des flux diffusifs de CO₂ et CH₄ respectivement était expliquée par la latitude, la concentration en matière organique, l'âge, et la profondeur des réservoirs (Barros et al., 2011). Un modèle empirique a récemment été construit incluant l'effet du climat, de la morphométrie, et des caractéristiques du bassin versant, et prédisant les flux ebullitifs et le dégazage du CH_4 en plus des flux diffusifs (CO_2 et CH_4) (UNESCO/IHA, 2017). Cependant, les données et connaissances actuelles ne sont pas suffisantes pour intégrer toutes les émissions en aval des barrages, et pour tester et réduire l'incertitude des modèles. Ainsi, les estimés globaux d'émissions de C des réservoirs varient entre 0.5 et 2.3 PgCO₂eq.yr⁻¹ (Barros et al., 2011 ; Bastviken et al., 2011 ; Deemer et al., 2016 ; St. Louis et al., 2000).

0.4 Les objectifs de la thèse

L'objectif général de cette thèse est d'approfondir les connaissances sur la dynamique du CO₂ et du CH₄ dans les réservoirs tropicaux. La recherche présentée vise à établir un portrait exhaustif des flux de C vers l'atmosphère et des sources et processus qui les sous-tendent. La thèse cherche aussi à comprendre comment évoluent ces dynamiques à travers le continuum aquatique en étudiant les patrons spatiaux du CO₂, du CH₄, et de leurs sources le long de la transition d'une rivière vers un réservoir. L'accent est mis sur la compréhension de l'interaction entre les différentes composantes du cycle du C à l'échelle de l'écosystème. Plus spécifiquement, la thèse est divisée en trois chapitres (Figure 0.1) qui répondent aux objectifs suivants :

- Chapitre 1 : Évaluer l'empreinte C d'un réservoir tropical en prenant en compte toutes les voies d'émission, et comparer les données mesurées aux prédictions des modèles actuels afin de tester leur performance et guider leur amélioration.
- Chapitre 2 : Quantifier et établir un bilan des sources de CO₂ et CH₄ dans l'épilimnion afin d'évaluer la contribution de chaque source aux flux diffusifs de surface le long d'un gradient hydrologique.
- Chapitre 3 : Examiner l'évolution spatiale et temporelle du CO₂ et CH₄ dans la couche profonde, et modéliser leurs concentrations selon un profil vertical en aval du barrage pour comprendre l'influence et l'interaction des facteurs qui façonnent les émissions en aval.



Figure 0.1: Schéma de la structure de la thèse illustrant les trois chapitres et les liens entre eux.

0.5 L'approche générale

Pour répondre aux questions énoncées, une étude de cas a été entreprise dans le réservoir Batang Ai situé sur l'ile de Bornéo dans la province de Sarawak en Malaisie. Le réservoir est géré par la compagnie hydroélectrique malaisienne Sarawak Energy Berhad avec laquelle une étroite collaboration a été établie pour rendre le projet possible. Le réservoir Batang Ai (68.4 km²) a été inondé en 1985 suite à la construction d'un barrage de 85 m pour la production d'hydroélectricité. Il est situé dans le parc national Batang Ai, le bassin versant est donc principalement constitué de forêt tropicale (forêt mixte de diptérocarpes et forêt secondaire), avec peu d'activités humaines (quelques habitations, sites d'aquacultures, et plantation à petite échelle). Les sols du bassin versant sont acides et pauvres en nutriments, formés de roche sédimentaire dominée par du schiste argileux et du grès à gros grains (Wasli *et al.*, 2011). Le réservoir est oligotrophe dû au type de sol et au territoire forestier du bassin

versant. L'eau du réservoir a un temps de résidence de 0.9 an, et est alimentée principalement par deux rivières en amont (Batang Ai et Engkari) qui se déversent dans les deux branches du réservoir, qui se croisent pour former le bassin principal du réservoir. Le bassin principal présente une stratification thermique permanente due au climat qui reste chaud toute l'année.

L'échantillonnage a été effectué durant 4 campagnes de terrain (durée d'environ 3 semaines), au cours de différentes années et mois : 1) novembre-décembre 2016, 2) avril-mai 2017, 3) février-mars 2018, 4) aout 2018. Durant les 4 campagnes d'échantillonnage, le climat était relativement similaire en termes de température et précipitations, mais le débit affluent était plus bas à la première et dernière campagne (Figure S3.1). Ainsi, l'échantillonnage couvre une variabilité temporelle et aussi spatiale vu qu'à chaque campagne, environ 39 sites ont été échantillonnés. Ces sites sont distribués à travers un gradient hydrologique allant des rivières affluentes, aux branches du réservoir, jusqu'au bassin principal, et à l'effluent plusieurs kilomètres en aval du barrage. Cette distribution spatiale a permis d'évaluer les changements dans la dynamique du CO_2 et du CH₄ lors de la transition d'un système hydrologique riverain à lacustre, reflétant l'impact de la construction d'un barrage.

Les conditions physicochimiques générales de l'eau (température, pH, concentrations en azote, phosphore, et matière organique dissoute) ont été mesurées dans tous les sites. L'étude des flux de GES vers l'atmosphère (Chapitre I) est basée principalement sur des mesures de gaz prises en surface (concentration, signature isotopique, flux). Afin d'examiner la dynamique du CO₂ et du CH₄ dans l'épilimnion (Chapitre II) et la couche profonde (Chapitre III), un échantillonnage plus approfondi a été effectué sur un sousensemble de 9 sites comprenant un profil vertical des gaz dans la colonne d'eau et dans les sédiments, ainsi que des mesures de métabolisme via des senseurs autonomes à haute fréquence et des incubations *in situ*. Les données collectées sur le terrain ont permis de construire un budget à l'échelle de l'écosystème des émissions vers l'atmosphère (Chapitre I) et des sources de CO_2 et CH_4 dans l'épilimnion, qui soustendent les flux diffusifs de surface (Chapitre II). Aussi, ces données ont servi à paramétriser la modélisation du CO_2 et CH_4 dans la couche profonde (Chapitre III) en lien avec les émissions en aval.

CHAPITRE I

THE CARBON FOOTPRINT OF A MALAYSIAN TROPICAL RESERVOIR: MEASURED VERSUS MODELED ESTIMATES HIGHLIGHT THE UNDERESTIMATED KEY ROLE OF DOWNSTREAM PROCESSES

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N.B. References cited in this chapter are presented at the end of the thesis.
1.1 Abstract

Reservoirs are important sources of greenhouse gases (GHGs) to the atmosphere and their number is rapidly increasing, especially in tropical regions. Accurately predicting their current and future emissions is essential but hindered by fragmented data on the subject, which often fail to include all emission pathways (surface diffusion, ebullition, degassing, and downstream emissions) and the high spatial and temporal flux variability. Here we conducted a comprehensive sampling of Batang Ai reservoir (Malaysia), and compared field-based versus modeled estimates of its annual carbon footprint for each emission pathway. Carbon dioxide (CO_2) and methane (CH_4) surface diffusion were higher in upstream reaches. Reducing spatial and temporal sampling resolution resulted in up to 64 and 33 % change in flux estimate, respectively. Most GHGs present in discharged water were degassed at the turbines, and the remainder were gradually emitted along the outflow river, leaving time for CH₄ to be partly oxidized to CO₂. Overall, the reservoir emitted 2475 gCO₂eq m⁻² yr⁻¹, with 89 % occurring downstream of the dam, mostly in the form of CH₄. These emissions, largely underestimated by predictions, are mitigated by CH₄ oxidation upstream and downstream of the dam, but could have been drastically reduced by slightly raising the water intake elevation depth. CO₂ surface diffusion and CH₄ ebullition were lower than predicted, whereas modeled CH₄ surface diffusion was accurate. Investigating latter discrepancies, we conclude that exploring morphometry, soil type, and stratification patterns as predictors can improve modeling of reservoir GHG emissions at local and global scales.

1.2 Introduction

Reservoirs provide a variety of services to humans (water supply, navigation, flood control, hydropower) and cover an estimated area exceeding 0.3 million km^2 globally (Lehner *et al.*, 2011). This area is increasing, with an expected rapid growth of the hydroelectric sector in the next two decades (International Hydropower Association (IHA), 2015), mainly in tropical and subtropical regions (Zarfl et al., 2015). The flooding of terrestrial landscapes can transform them into significant greenhouse gas (GHG) sources to the atmosphere (Prairie et al., 2018; Rudd et al., 1993; Teodoru et al., 2012). While part of reservoir GHG emissions would occur naturally (not legitimately attributable to damming), the remainder results from newly created environments favoring carbon (C) mineralization, particularly methane (CH₄) production (flooded organic-rich anoxic soils) (Prairie et al., 2018). Field studies have revealed a wide range in measured fluxes, with spatial and temporal variability sometime spanning several orders of magnitude within a single reservoir (Paranaíba et al., 2018; Sherman et Ford, 2011). Moreover, reservoirs can emit GHG through several pathways: diffusion of gas at the air-water interface (surface diffusion), release of gas bubbles formed in the sediments (ebullition), and for some reservoirs (mostly hydroelectric) through gas release following pressure drop upon water discharge (degassing), and through evasion of the remaining excess gas in the outflow river (downstream emissions). The relative contribution of these flux pathways to total emissions is extremely variable. While surface diffusion is the most frequently sampled, it is often not the main emission pathway (Demarty et Bastien, 2011). Indeed, measured ebullition, degassing, and downstream emissions range from negligible to several order of magnitude higher than surface diffusion in different reservoirs (Bastien et Demarty, 2013; DelSontro et al., 2010; Galy-Lacaux et al., 1997; Keller et Stallard, 1994; Kemenes et al., 2007; Teodoru et al., 2012; Venkiteswaran et al., 2013), making it a challenge to model total reservoirs GHG emissions.

Literature syntheses over the past 20 years have yielded highly variable global estimates of reservoirs GHG footprint, ranging from 0.5 to 2.3 PgCO₂eq yr⁻¹ (Barros et al., 2011; Bastviken et al., 2011; Deemer et al., 2016; St. Louis et al., 2000). These estimates are based on global extrapolations of averages of sampled systems, representing an uneven spatial distribution biased toward North America and Europe, and an uneven mixture of emission pathways. Recent studies have highlighted the lack of spatial and temporal resolution as well as the frequent absence of some flux pathways (especially degassing, downstream, and N_2O emissions) in most reservoir GHG assessments (Beaulieu et al., 2016; Deemer et al., 2016). More recently, studies have focused on identifying drivers of reservoir GHG flux variability. Using global empirical data, Barros et al. (2011) proposed the first quantitative models for reservoir carbon dioxide (CO_2) and CH_4 surface diffusion as a negative function of reservoir age, latitude, and mean depth (for CO_2 only), and a positive function of dissolved organic carbon (DOC) inputs (Barros et al., 2011). An online tool (G-res) for predicting reservoir CO₂ and CH₄ emissions was later developed on the basis of a similar empirical modeling approach of measured reservoir fluxes with globally available environmental data (UNESCO/IHA, 2017). Modeling frameworks to predict GHG emissions from existing and future reservoirs are essential tools for reservoir management. However, their accuracy is directly related to available information and inherently affected by gaps and biases of the published literature. For example, while the G-res model predicts reservoir CO₂ and CH₄ surface diffusion as well as CH₄ ebullition and degassing on the basis of temperature, age, % littoral zone and soil organic C, it does not consider N₂O emissions, CO₂ degassing, and downstream emissions due to scarcity of data. Overall, the paucity of comprehensive empirical studies limits our knowledge of reservoir GHG dynamics at a local scale, introducing uncertainties in large scale estimates and hindering model development.

The research reported here focuses on building a comprehensive assessment of GHG fluxes of Batang Ai, a tropical reservoir in South-east Asia (Malaysia), over four

sampling campaigns spanning two years with an extensive spatial coverage. The main objective of this study is to provide a comprehensive account of CO_2 , CH_4 and N_2O fluxes from surface diffusion, ebullition, degassing, and downstream emissions (accounting for riverine CH_4 oxidation) to better understand what shapes their relative contributions and their potential mitigation. The second objective is to compare our measured values with modeled estimates from each pathway and gas species to locate where the largest discrepancies are, and thereby identify research avenues for improving the current modeling framework.

1.3 Materials and methods

1.3.1 Study site and sampling campaigns

Batang Ai is a hydroelectric reservoir located on the Borneo Island in the Sarawak province of Malaysia (latitude 1.16° and longitude 111.9°). The regional climate is tropical equatorial with a relatively constant temperature throughout the year, on average 23°C in the morning to 32°C during the day. Annual rainfall varies between 3300 and 4600 mm with two monsoon seasons: November to February (northeast monsoon), and June to October (southwest) (Sarawak Government, 2019). Batang Ai reservoir was impounded in 1985 with no prior clearing of the vegetation, and has a dam wall of 85 m in height, a mean depth of 34 m, and a total area of 68.4 km². The reservoir catchment consists of 1149 km² of mostly forested land where human activities are limited to a few traditional habitations and associated croplands, and localized aquaculture sites within the reservoir main basin. The reservoir has two major inflows: the Batang Ai and Engkari rivers, which flow into two reservoir branches merging upstream of the main reservoir basin (Figure 1). Four sampling campaigns were conducted: 1) November 14th to December 5th 2016 (Nov-Dec 2016), 2) April 19th to May 3th 2017 (Apr-May 2017), 3) February 28th to March 13th 2018 (Feb-Mar 2018), and 4) August 12th to 29th 2018 (Aug 2018).



Figure 1.1: Map of Batang Ai showing the location of sampled sites and reservoir sections. * Represents the reservoir inflow sites.

1.3.2 Water chemistry

Samples for DOC, total phosphorus (TP), total nitrogen (TN), and chlorophyll a (Chla) analyses were collected from the water surface (<0.5 m) at all surface diffusion sampling sites shown in Figure 1 and during each campaign. For TP and TN, we collected non-filtered water in acid-washed glass vials stored at 4°C until analysis. TP was measured by spectrophotometry using the standard molybdenum blue method after persulphate digestion at 121°C for 20 min, and a calibration with standard solutions from 10 to 100 μ g L⁻¹ with a 5 % precision (Wetzel et Likens, 2000). TN analyses were performed by alkaline persulphate digestion to NO₃, subsequently measured on a flow

Alpkem analyzer (OI Analytical Flow Solution 3100) calibrated with standard solutions from 0.05 to 2 mg L⁻¹ with a 5 % precision (Patton et Kryskalla, 2003). Water filtered at 0.45 μ m was used for DOC analysis with a Total Organic Carbon analyser 1010-OI following sodium persulphate digestion, and calibrated with standard solutions from 1 to 20 mg L⁻¹ with a 5 % precision (detection limit of 0.1 mg L⁻¹). Chla was analysed through spectrophotometry following filtration on Whatman (GF/F) filters and extraction by hot 90 % ethanol solution (Sartory et Grobbelaar, 1984).

1.3.3 Surface diffusion

Surface diffusion is the flux of gas between the water surface and the air driven by a gradient in gas partial pressure. Surface diffusion of CO_2 and CH_4 to the atmosphere were measured at 36 sites in the reservoir and 3 sites in the inflow rivers (Figure 1.1), and sampling of the same sites was repeated each campaign (with a few exceptions). Fluxes were measured using a static air tight floating chamber connected in a closed loop to an Ultraportable gas Analyser (UGGA from LGR). Surface diffusion rates (F_{gas}) were derived from the linear change in CO_2 and CH_4 partial pressures (continuously monitored at 1 Hz for a minimum of 5 min) through time inside the chamber using the following Eq. (1.1):

$$F_{\text{gas}} = \frac{s \, V}{m \, A} \tag{1.1}$$

where *s* is the gas accumulation rate in the chamber, V = 25 L the chamber volume, A = 0.184 m² the chamber surface area, and *m* the gas molar volume at current atmospheric pressure.

 N_2O surface diffusion was estimated at 7 of the sampled sites (Figure 1.1) using the following Eq. (1.2) (Lide, 2005):

$$F_{\rm gas} = k_{\rm gas} \left(C_{\rm gas} - C_{\rm eq} \right) \tag{1.2}$$

where k_{gas} is the gas exchange coefficient, C_{gas} is the gas concentration in the water and C_{eq} is the theoretical gas concentration at equilibrium given measured water temperature, atmospheric pressure and ambient gas concentration. C_{N2O} was measured using the headspace technique, with a 1.12 L sealed glass serum bottle containing surface water and a 0.12 L headspace of ambient air. After shaking the bottle for two minutes to achieve air-water equilibrium, the headspace gas was extracted from the bottle with an airtight syringe and injected in previously evacuated 9 mL glass vial capped with an air tight butyl stopper and aluminium seal. Three analytical replicates and a local sample of ambient air were taken at each site and analysed by gas chromatography using a Shimadzu GC-2040, with a Poropaq Q column to separate gases and an ECD detector calibrated with 0.3, 1, and 3 ppm of N₂O certified standard gas. After analysis the original N₂O concentration of the water was back-calculated based on the water temperature before and after shaking (for gas solubility), the ambient atmospheric pressure, the ratio of water to air in the sampling bottle, and the headspace N₂O concentration before shaking. k_{N2O} was derived from measured k_{CH4} values obtained by rearranging Eq. (1.2) for CH₄, with known values of F_{gas} , C_{gas} , and C_{eq} . The k_{CH4} to k_{N2O} transformation was done using the following Eq. (1.3) (Cole et Caraco, 1998 ; Ledwell, 1984):

$$k_{\rm N2O} = \left(\frac{sc_{\rm N2O}}{sc_{\rm CH4}}\right)^{-0.67} k_{\rm CH4}$$
(1.3)

where Sc is the gas Schmidt number (Wanninkhof, 1992).

CH₄ and CO₂ concentrations in the water were measured using the headspace technique. Surface water was collected in a 60 mL gas-tight plastic syringe in which a 30 mL headspace was created (using either ambient air or carbon free air). The syringe was shaken for 2 min to achieve air-water gas equilibrium. The gas phase was then injected in a 12 mL air-tight pre-evacuated vial and subsequently analysed through manual injection on a Shimadzu GC-8A Gas chromatograph with flame ionization detector following a calibration curve with certified gas standards (0-10000 ppm for CO₂ and 0-50000 ppm for CH₄). The samples were also analyzed for isotopic δ^{13} CO₂ and δ^{13} CH₄ signatures by manually injecting 18 mL of gas in a Cavity Ring Down Spectrometer (CRDS) equipped with a Small Sample Isotopic Module (SSIM A0314, Picarro G2201-*i* Analyzer) set in a non-continuous mode with a three point calibration curve based on certified gas standards (-40 -3.9, and 25.3 ‰ for δ^{13} CO₂, and -66.5, -38.3, -23.9 ‰ for δ^{13} CH₄).

1.3.4 Ebullition

Ebullition is the process through which gas bubbles formed in the sediments rise through the water column and are released to the atmosphere. Sediment gas ebullition was measured at four sites in the reservoir and two sites in the inflows (Figure 1.1) by deploying 0.785 m^2 underwater inverted funnel traps at 2 to 3 m deep for approximately 20 days in the reservoir and 1h in the inflows. The top part of a closed plastic syringe was fixed to the narrow end of the funnel trap where the emerging bubbles accumulated. Upon recovery, bubble gas volume was measured, collected from the syringe, and injected in 12 mL pre-evacuated air tight vials for CO₂ and CH₄ concentration analyses (using the aforementioned method). Ebullition rate was calculated assuming the original bubble composition was similar to bubbles collected almost right after ascent in the inflows sites, which was 100 % CH₄. Hence we considered CO₂ and N₂O ebullition to be null.

In order to estimate the potential for sediment accumulation fueling ebullition in the littoral zone, we calculated the mud energy boundary depth (EBD in m, below which fine grained sediments accumulation occurs) using the reservoir surface area (E in km²) as the exposure parameter in the following Eq. (1.4) (Rowan *et al.*, 1992):

1.3.5 Degassing, downstream emissions and CH₄ oxidation

Degassing of CO₂ and CH₄ right after water discharge (F_{deg}), and downstream emissions of the remaining reservoir-derived GHGs in the outflow river (F_{dwn}) were calculated using the following Eq. (1.5) and Eq. (1.6):

$$F_{\rm deg} = Q \left(C_{\rm up} - C_0 \right) \tag{1.5}$$

$$F_{\rm dwn} = Q \left(C_0 - C_{19} + C_{\rm ox} \right) \tag{1.6}$$

where Q is the water discharge, and C_{up} , C_0 and C_{19} the measured gas concentrations upstream of the dam at the water withdrawal depth, at the powerhouse right after water discharge, and in the outflow 19 km downstream of the dam respectively. C_{ox} is the net change in gas concentration due to oxidation (loss for CH₄ and gain for CO₂). For downstream emissions, we considered that, after a river stretch of 19 km, all excess gas originating from the reservoir was evaded and gas concentration was representative of the outflow river baseline. This assumption potentially underestimates actual downstream emissions (in case of remaining excess gas after 19 km). However, given the observed exponential decrease of gas concentration along the outflow (Figure 1.3), emissions after 19 km are expected to be small compared to those in the 0 to 19 km river stretch, consistent with observations in other reservoirs (Guérin *et al.*, 2006 ; Kemenes *et al.*, 2007).

Gas concentrations upstream and downstream of the dam were obtained by measuring, in each campaign, CO_2 and CH_4 concentrations in a vertical profile right upstream of the dam at a 1 to 3 m interval from 0 to 32 m, and at four locations in the outflow: at 0 (power house), 0.6, 2.7, and 19 km downstream of the dam (Figure 1.1). Sampling was done using a multi-parameter probe equipped with depth, oxygen, and temperature

(1.4)

sensors (Yellow Spring Instruments, YSI model 600XLM-M) attached to a 12 Volt submersible Tornado pump (Proactive Environmental Products) for water collection. Gas concentration and δ^{13} C were measured as described in section 1.3.3. Water withdrawal depth ranged from 20 to 23 m and was estimated based on known values of elevations of water intake and water level compared to sea level. Gas concentration in the water exiting the reservoir was defined as the average measured gas concentrations in the ± 1 m range of the withdrawal depth.

Estimates of downstream CH₄ oxidation were obtained, for each sampling campaign, by calculating the fraction of CH₄ oxidized (F_{ox}) using the following Eq. (1.7):

$$F_{\rm ox} = \frac{-\left(\ln(\delta^{13}CH_{\rm 4resid} + 1000) - \ln(\delta^{13}CH_{\rm 4source} + 1000)\right) \left(1 - \frac{[CH_4]_{\rm resid}}{[CH_4]_{\rm source}}\right)}{\left(1 - \frac{1}{\alpha}\right) \ln\left(\frac{[CH_4]_{\rm resid}}{[CH_4]_{\rm source}}\right)}$$
(1.7)

Eq. (1.7) is based on a non-steady state isotopic model developed considering evasion (emission to the atmosphere) and oxidation as the two loss processes for CH₄ in the outflow river, assuming negligible isotopic fractionation for evasion (Knox *et al.*, 1992) and a fractionation of $\alpha = 1.02$ for oxidation (Coleman *et al.*, 1981) (see derivation in Supplementary Information). [CH₄]_{source}, [CH₄]_{resid}, δ^{13} CH₄_{source}, and δ^{13} CH₄_{resid} are the concentrations of CH₄ and their corresponding isotopic signatures at the beginning of the outflow (km 0) and 19 km downstream, representing the source and residual pools of CH₄ respectively. The amount of CH₄ oxidized to CO₂ along the 19 km of river stretch for each sampling campaign was calculated as the product of F_{ox} and [CH₄]_{source}. The resulting loss of CH₄ and gain of CO₂ in the outflow were accounted for in downstream emissions (C_{ox} in Eq. (1.6)). Note that downstream N₂O emissions were considered null since N₂O concentrations measured in the deep reservoir layer were lower than concentrations in the outflow.

1.3.6 Ecosystem scale C footprint

Batang Ai annual C footprint was calculated as the sum of surface diffusion, ebullition, degassing, and downstream emissions of CO₂, CH₄, and N₂O considering a greenhouse warming potential of 1, 34, and 298 respectively over a 100 years lifetime period (Myhre *et al.*, 2013). For each flux pathway, annual flux was estimated as the average of the sampling campaigns. Ecosystem scale estimate of surface diffusion was calculated for each campaign as the average of measured flux rates applied to the reservoir area for N₂O, and for CO₂ and CH₄ it was obtained by spatial interpolation of measured fluxes over the reservoir area based on inverse distance weighting with a power of two (a power of one yields similar averages, CV < 11 %) using package gstat version 1.1-6 in the R version 3.4.1 software (Pebesma, 2004 ; R Core Team, 2017). Ebullition at the reservoir scale was calculated as the average of measured reservoir ebullition rates applied to the littoral area (< 3 m deep).

The estimated GHG emissions of Batang Ai based on measured data was compared to values derived from the G-res model (UNESCO/IHA, 2017) and the model presented in Barros et al. (2011). Both models predict surface CO_2 and CH_4 diffusion as a function of age and account for the effect of temperature using however different proxies: the G-res uses effective temperature while Barros et al. model uses latitude (an indirect proxy that integrates other spatial differences). In terms of CO_2 surface diffusion, the G-res uses reservoir area, soil C content, and TP to quantify the effect of C inputs fueling CO_2 production, while Barros et al. model uses directly DOC inputs (based on in situ DOC concentration). For CH_4 surface diffusion, both models account for morphometry using the fraction of littoral area (G-res) or the mean depth (Barros et al. model). Overall, both models predict surface diffusion based on the same conceptual framework but use different proxies. CH_4 ebullition and degassing are modeled only by the G-res, being the sole model available to this date. Details on

models equations and input variables are presented in the Supplementary Information (Table S1.2 and S1.3).

1.4 Results and discussion

1.4.1 Water chemistry

The reservoir is stratified throughout the year with a thermocline at a depth around 13 m and mostly anoxic conditions in the hypolimnion of the main basin (Figure S1.1). The system is oligotrophic, with very low concentrations of DOC, TP, TN, and Chla averaging 0.9 (SD \pm 0.2) mg L⁻¹, 5.9 (SD \pm 2.4) µg L⁻¹, 0.11 (SD \pm 0.04) mg L⁻¹, and 1.3 (SD \pm 0.7) µg L⁻¹ respectively (Table S1.1), and high water transparency (Secchi depth > 5 m). In the reservoir inflows, concentrations of measured chemical species are slightly higher but still in the oligotrophic range (Table S1.1), however the transparency is much lower due to turbidity (Secchi < 0.5 m). The oligotrophic status of the reservoir likely results from nutrient poor soils (Wasli *et al.*, 2011) and a largely undisturbed forested catchment in the protected Batang Ai National Park. The reservoir's low Chla concentrations are comparable to the neighboring Bakun reservoir (Ling *et al.*, 2017), and its DOC concentrations are on the low end of the wide range of measured values in nearby rivers (Martin *et al.*, 2018).

1.4.2 Surface diffusion

Measured CO₂ diffusion in the reservoir averaged 7.7 (SD \pm 18.2) mmol m⁻² d⁻¹ (Table S1.1), which is on the low end compared to other reservoirs (Deemer *et al.*, 2016) and even to natural lakes (Sobek *et al.*, 2005), but similar to CO₂ fluxes measured in two reservoirs in Lao PDR (Chanudet *et al.*, 2011). CO₂ diffusion across all sites ranged from substantial uptake to high emissions (from -30.8 to 593.9 mmol m⁻² d⁻¹, Table

S1.1) reflecting a large spatial and temporal variability. Spatially, CO₂ fluxes measured in the main basin and branches had similar averages of 7.9 and 7.3 mmol m⁻² d⁻¹ respectively (overall SD \pm 18.2), contrasting with higher and more variable values in the inflows with a mean of 137.3 (SD \pm 192.4) mmol.m⁻².d⁻¹ (Figure 1.2). Within the reservoir, CO₂ fluxes varied (SD \pm 18.2 mmol m⁻² d⁻¹) but did not follow a consistent pattern, and might reflect pre-flooding landscape heterogeneity (Teodoru *et al.*, 2011). Temporally, highest average reservoir CO₂ fluxes were measured in Apr-May 2017, when no CO₂ uptake was observed, contrary to other campaigns, especially Feb-Mar and Aug 2018, when CO₂ uptake was common (Figure S1.2) and average Chla concentrations were the highest. This reflects the important role of metabolism (namely CO₂ consumption by primary production) in modulating surface CO₂ fluxes in Batang Ai.

All CH₄ surface diffusion measurements were positive and ranged from 0.03 to 113.4 mmol m⁻² d⁻¹ (Table S1.1). Spatially, CH₄ fluxes were progressively higher moving further upstream (Figure 1.2 and S1.3) with decreasing water depth and increasing connection to the littoral. This gradient in morphometry induces an increasingly greater contact of the water with bottom and littoral sediments, where CH₄ is produced, explaining the spatial pattern of CH₄ fluxes. CH₄ surface diffusion also varied temporally, but to a lesser extent than CO₂, being on average highest in Aug 2018 in the reservoir and in Nov-Dec 2016 in the inflows.

Reservoir N₂O surface diffusion (measured with a limited spatial resolution) averaged -0.2 (SD \pm 2.1) nmol m⁻² d⁻¹ (Table S1.1). The negative value indicates that the system acts as a slight net sink of N₂O, absorbing an estimated 2.1 gCO₂eq m⁻² yr⁻¹ (Table 1.2). Atmospheric N₂O uptake have previously been reported in aquatic systems and linked to low oxygen and nitrogen content conducive to complete denitrification which consumes N₂O (Soued *et al.*, 2016; Webb *et al.*, 2019). These environmental

conditions match observations in Batang Ai, with a low average TN concentration of 0.11 (0.04) mg L^{-1} (Table S1.1) and anoxic deep waters (Figure S1.1).



Figure 1.2: Boxplots of measured CH_4 (on a log axis) and CO_2 fluxes grouped according to spatial position. Boxes are bounded by the 25th and 75th percentile and show medians (solid lines), and whiskers show 10th and 90th percentiles. Gray circles show single data points.

1.4.3 Ebullution

We calculated that CH₄ ebullition rates in Batang Ai's littoral area ranged from 0.02 to $0.84 \text{ mmol m}^{-2} \text{ d}^{-1}$, which contrasts with rates measured in its inflows that are several orders of magnitude higher (52 to 103 mmol m⁻² d⁻¹). Similar patterns were observed in other reservoirs, where inflow arms where bubbling hot spots due to a higher organic C supply driven by terrestrial matter deposition (DelSontro et al., 2011; Grinham et al., 2018). Since ebullition rates are notoriously heterogeneous and were measured at only four sites in the reservoir, they may not reflect ecosystem-scale rates. However, our attempt to manually provoke ebullition at several other sites (by physically disturbing the sediments) did not result in any bubble release, confirming the low potential for ebullition in the reservoir littoral zone. Moreover, we calculated that fine grained sediment accumulation is unlikely at depths shallower than 9.7 m (estimated EBD) in Batang Ai. This, combined with the reservoir steep slope, prevents the sustained accumulation of organic material in littoral zones (Blais et Kalff, 1995), hence decreasing the potential for CH₄ production and bubbling there. Also, apparent littoral sediment composition in the reservoir; dense clay with low porosity, may further hinder bubble formation and emission (de Mello et al., 2018).

1.4.4 Degassing and downstream emissions

Emissions downstream of the dam, expressed on a reservoir-wide areal basis, ranged from 19.3 to 30.0 mmol m⁻² d⁻¹ for CO₂ and from 5.9 to 13.8 mmol m⁻² d⁻¹ for CH₄ (Table 1.1). The amount of CO₂ exiting the reservoir varied little between sampling campaigns (CV = 3 %) contrary to CH₄ (CV = 28 %, Table 1.1 and Figure 1.3). Higher temporal variability of CH₄ concentration in discharged water is likely modulated by microbial CH₄ oxidation in the reservoir water column upstream of the dam. Evidence of high CH₄ concentration and increase of δ^{13} CH₄ right around the water withdrawal depth (Figure S1.1). This vertical pattern results from higher oxygen availability when

moving up in the hypolimnion (Figure S1.1), promoting CH₄ oxidation at shallower depths.

Once GHGs have exited the reservoir, a large fraction (40 and 65 % for CO₂ and CH₄ respectively) is immediately lost to the atmosphere as degassing emissions (Table 1.1), which is in line with previous literature reports (Kemenes et al., 2016). Along the outflow river, CO₂ and CH₄ concentrations gradually decreased, δ^{13} CO₂ remained stable, whereas δ^{13} CH₄ steadily increased (Figure 1.3). Given the very small isotopic fractionation (0.9992) of CH₄ during gas evasion (Knox et al., 1992), the only process that can explain the observed δ^{13} CH₄ increase is CH₄ oxidation (Bastviken *et al.*, 2002; Thottathil et al., 2018). We estimated that riverine CH₄ oxidation ranged from 0.38 to 1.80 mmol $m^{-2} d^{-1}$ (expressed per m^2 of reservoir area for comparison), transforming 18 to 32 % (depending on the sampling campaign) of the CH₄ to CO₂ within the first 19 km of the outflow. Riverine oxidation rates did not co-vary temporally with water temperature, oxygen availability, or CH₄ concentrations (known as typical drivers (Thottathil et al., 2019)), hence they might be regulated by other factors like light and microbial assemblages (Murase et Sugimoto, 2005; Oswald et al., 2015). Overall, riverine oxidation of CH₄ to CO₂ (which has a 34 times lower warming potential) reduced radiative forcing of downstream emissions (excluding degassing) by, on average, 21 %, and the total annual reservoir C footprint by 7 %. Despite having a measurable impact on reservoir GHG emissions, CH₄ oxidation downstream of dams was only considered in three other reservoirs to our knowledge (DelSontro et al., 2016b; Guérin et Abril, 2007; Kemenes et al., 2007). Accounting for this process is particularly important in systems where downstream emissions are large, a common situation in tropical reservoirs (Demarty et Bastien, 2011). While additional data on the subject is needed, our results provide one of the first basis for understanding CH4 oxidation downstream of dams, and eventually integrating this component to global models (from which it is currently absent).



Figure 1.3: Concentrations (black symbols and solid line) and δ^{13} C (gray symbols and dotted lines) of CO₂ and CH₄ from right upstream of the dam (gray band) to 19 km downstream in the outflow river. Circles, squares, diamonds, and triangles represent values from Nov-Dec 2016, Apr-May 2017, Feb-Mar 2018, and Aug 2018 respectively.

Table 1.1: CO_2 and CH_4 dynamics downstream of the dam: gas export rate from upstream to downstream of the dam, degassing, result of CH_4 oxidation (CO_2 production and CH_4 consumption), downstream emissions, and total emissions to the atmosphere below the dam. Uncertainties based on variation coefficients are reported in parentheses. Units are in mmol m⁻² d⁻¹ of reservoir surface area.

	GHG downstream of the dam (mmol m ⁻² d ⁻¹)				
	Exported	Degassed	Gain / loss. by oxidation	Downstream emiss.	Total emiss.
CO ₂					
Nov-Dec 2016	40.62 (±2.27)	15.26 (±0.85)	0.90 (±0.13)	12.67 (±0.71)	27.93 (±1.56)
Apr-May 2017	37.80 (±2.11)	14.91 (±0.83)	0.59 (±0.08)	9.83 (±0.55)	24.70 (±1.38)
Feb-Mar 2018	37.98 (±2.12)	9.58 (±0.54)	1.80 (±0.26)	9.70 (±0.54)	19.30 (±1.08)
Aug 2018	38.07 (±2.13)	21.67 (±1.21)	0.38 (±0.05)	8.31 (±0.46)	30.00 (±1.68)
CH ₄					
Nov-Dec 2016	14.84 (±2.10)	11.56 (±1.64)	0.90 (±0.13)	2.19 (±0.31)	13.76 (±1.95)
Apr-May 2017	7.32 (±1.04)	4.00 (±0.57)	0.59 (±0.08)	1.90 (±0.27)	5.90 (±0.84)
Feb-Mar 2018	12.47 (±1.77)	4.92 (±0.70)	1.80 (±0.26)	3.99 (±0.57)	8.91 (±1.26)
Aug 2018	10.71 (±1.52)	9.54 (±1.35)	0.38 (±0.05)	0.51 (±0.07)	10.05 (±1.42)

1.4.5 Importance of sampling resolution

High spatial and temporal sampling resolution have been recently highlighted as an important but often lacking aspect of reservoir C footprint assessments (Deemer *et al.*, 2016; Paranaíba *et al.*, 2018). Reservoir scale fluxes are usually derived from applying an average of limited flux measurements to the entire reservoir area. For Batang Ai, this method overestimates by 14 % (130 gCO₂eq m⁻² yr⁻¹) and 64 % (251 gCO₂eq m⁻² yr⁻¹) CO₂ and CH₄ surface diffusion respectively compared to spatial interpolation. This is due to the effect of extreme values that are very constrained in space but have a disproportionate effect on the overall flux average. Also, reducing temporal sampling resolution to one campaign instead of four changes the reservoir C footprint estimate by up to 33 %. An additional source of uncertainty in reservoir flux estimates is the definition of a baseline value representing natural river emissions in order to calculate

downstream emissions of excess gas in the outflow attributable to damming. In Batang Ai, downstream emission was estimated assuming the GHG concentration 19 km downstream of the dam is a representative baseline for the outflow, however, measured values in the pre-impounded river would have substantially reduced the estimate uncertainty. Results from Batang Ai reinforce the importance of pre and post-impoundment sampling resolution and upscaling methods in annual reservoir-scale GHG flux estimates.

1.4.6 Reservoir C footprint and potential mitigation

Most of Batang Ai emissions occur downstream of the dam through degassing (64.2 %) and downstream emissions (25.0 %), while surface diffusion contributed only 10.6 %, and ebullition 0.14 % (Table 1.2). In all pathways, radiative potential of CH₄ fluxes were higher than CO₂ and N₂O (especially for degassing), accounting for 79.0 % of Batang Ai CO₂eq emissions. This distribution of the flux can be attributed mostly to the accumulation of large quantities of CH₄ in the hypolimnion, combined with the fact that the withdrawal depth is located within this layer, allowing the accumulated gas to escape to the atmosphere. Previous studies on reservoirs with similar characteristics to Batang Ai (tropical climate with a permanent thermal stratification and deep water withdrawal) have also found degassing and downstream emissions to be the major emission pathways, especially for CH₄ (Galy-Lacaux *et al.*, 1997 ; Kemenes *et al.*, 2007).

Overall, we estimated that the reservoir emits on average 2475 (\pm 327) gCO₂eq m⁻² yr⁻¹ which corresponds to 0.169 TgCO₂eq yr⁻¹ over the whole system. In comparison, the annual areal emission rate (diffusion and ebullition) of the inflows, based on a more limited sampling resolution, is estimated to range from 10.8 to 52.5 kgCO₂eq m⁻² yr⁻¹, mainly due to extremely high ebullition. When applied to the approximated surface area of the river before impoundment (1.52 km²), this rate translates to 0.016 – 0.080 TgCO₂eq (Table 1.2), assuming similar flux rates in the current inflows and pre-

impoundment river. While the emission rate of the river per unit of area is an order of magnitude higher than for the reservoir, its estimated total flux remains 2.1 to 10.6 times lower due to a much smaller surface. Higher riverine emissions rates are probably due to a shallower depth and higher inputs of terrestrial organic matter, both conducive to CO₂ and CH₄ production and ebullition. Changing the landscape hydrology to a reservoir drastically reduced areal flux rates, especially ebullition, however, it widely expanded the volume of anoxic environments (sediments and hypolimnion), creating a vast new space for CH₄ production. The new hydrological regime also created an opportunity for the large quantities of gas produced in deep layers to easily escape to the atmosphere through the outflow and downstream river.

One way to reduce reservoir GHG emissions is to ensure low CO₂ and CH₄ concentrations at the water withdrawal depth. In Batang Ai, maximum CO₂ and CH₄ concentrations are found in the reservoir deep layers, and rapidly decrease from 20 to 10 m for CO₂ and from 25 to 15 m for CH₄ (Figure S1.1). This pattern is commonly found in lakes and reservoirs and results from thermal stratification and biological processes (aerobic respiration and CH₄ oxidation). Knowing this concentration profile, degassing and downstream emissions could have been reduced in Batang Ai by elevating the water withdrawal depth to avoid hypolimnetic gas release. We calculated that elevating the water withdrawal depth by 1, 3, and 5 m would result in a reduction of degassing and downstream emissions by 1, 11, and 22 % for CO₂ and by 28, 92, and 100 % for CH₄, respectively (Figure S1.4). Consequently, a minor change in the dam design could have drastically reduced Batang Ai's C footprint. This should be taken in consideration in future reservoir construction, especially in tropical regions.

1.4.7 Measured versus modeled fluxes

Based on measurements, Batang Ai emits on average 113 (\pm 22) gCO₂eq m⁻² yr⁻¹ via surface CO₂ diffusion. This value is 41 times lower than predicted by Barros et al. model (4671 gCO₂eq m⁻² yr⁻¹, Table 1.2) based on reservoir age, DOC inputs (derived

from DOC water concentration), and latitude (Barros et al., 2011). The high predicted value for Batang Ai, being a relatively old reservoir with very low DOC concentration, is mainly driven by its low latitude. While reservoirs in low latitudes globally have higher average CO₂ fluxes due to higher temperature and often dense flooded biomass (Barros et al., 2011; St. Louis et al., 2000), our results provide a clear example that not all tropical reservoirs have high CO_2 emissions by simple virtue of their geographical location. Despite high temperature, Batang Ai's very low water organic matter content (Table S1.1) offers little substrate for net heterotrophy, and its strong permanent stratification creates a physical barrier potentially retaining CO₂ derived from flooded biomass in the hypolimnion. The only three other sampled reservoirs in Southeast Asia (Nam Leuk and Nam Ngum in Lao PDR, and Palasari in Indonesia) also exhibited low organic C concentration (for reservoirs in Lao PDR) and low to negative average surface CO₂ diffusion despite their low latitude (Chanudet et al., 2011; Macklin et al., 2018). This suggests that, while additional data are needed, low CO₂ diffusion may be common in Southeast Asian reservoirs, and likely linked to the low organic C content.

In comparison, the G-res model predicts a CO₂ surface diffusion of 577 (509-655) $gCO_2eq m^{-2} yr^{-1}$, which includes the flux naturally sustained by catchment C inputs (397 $gCO_2eq m^{-2} yr^{-1}$, predicted flux 100 years after flooding) and the flux derived from organic matter flooding (180 $gCO_2eq m^{-2} yr^{-1}$). While the predicted G-res value is much closer than that predicted from the Barros et al. model, it still overestimates measured flux, mostly the natural baseline (catchment derived) part of it. The G-res predicts baseline CO₂ effluxes as a function of soil C content, a proxy for C input to the reservoir. While Batang Ai soil is rich in organic C (~50 g kg⁻¹), it also has a high clay content (> 40 %) (ISRIC - World Soil Information, 2019 ; Wasli *et al.*, 2011) which is known to bind with organic matter and reduce its leaching to the aquatic environment (Oades, 1988). This may explain the unusually low DOC concentration in the reservoir and its inflows (0.3 to 1.8 mg L⁻¹, Table S1.1) that are among the lowest reported in

freshwaters globally (Sobek et al., 2007). Clay-rich soils are ubiquitous in tropical landscapes (especially in Southeast Asia, Central America, and Central and Eastern Africa) (ISRIC - World Soil Information, 2019), however, their impact on global-scale patterns of aquatic DOC remains unknown. This may be due to a lack of aquatic DOC data, with the most recently published global study on the subject featuring only one tropical system and a heavy bias towards North America and Europe (Sobek et al., 2007). Exploring the global-scale picture of aquatic DOC and its link to watershed soils characteristics would be a significant step forward in the modeling of reservoir CO_2 diffusion. Indeed, had the G-res model been able to capture the baseline emissions more correctly in Batang Ai (close to zero given the very low DOC inputs), predictions would have nearly matched observations. Finally, note that the G-res model is not suitable to predict CO_2 uptake, which was observed in 32 % of flux measurements in Batang Ai due to an occasionally net autotrophic surface metabolism favored under low C inputs (Bogard et del Giorgio, 2016). Improving this aspect of the model depends on the capacity to predict internal metabolism of aquatic systems at a global scale, which is currently lacking. Overall, reservoir CO₂ diffusion models may be less performant in certain regions, like Southeast Asia, due to an uneven spatial sampling distribution and a general lack of knowledge and data on C cycling in some parts of the world.

Our field-based estimate of Batang Ai CH₄ surface diffusion is 153 (\pm 22) gCO₂eq m⁻² yr⁻¹, which differs by only 5 % and 15 % from the G-res and Barros et al. modeled predictions of 161 (132-197) and 176 gCO₂eq m⁻² yr⁻¹ respectively (Table 1.2). Both models use as predictors age, a proxy for water temperature (air temperature or latitude), and an indicator of reservoir morphometry (% littoral area or mean depth), and Barros et al. also uses DOC input (Table S1.3). Similar predictors were identified in a recent global literature analysis, which also pointed out the role of trophic state in CH₄ diffusion, with Batang Ai falling well in the range of flux reported in other oligotrophic

reservoirs (Deemer *et al.*, 2016). Overall, our results show that global modeling frameworks for CH₄ surface diffusion capture reasonably well the reality of Batang Ai.

Measured estimate of reservoir-scale CH₄ ebullition averaged 3.4 (\pm 1.9) gCO₂eq m⁻² yr⁻¹ (Table 1.2), which is one of the lowest reported globally in reservoirs (Deemer *et al.*, 2016), and is an order of magnitude lower than the 52 (32 - 83) gCO₂eq m⁻² yr⁻¹ predicted by the G-res model (the only available model for reservoir ebullition). This contrasts with the perception that tropical reservoirs consistently have high ebullitive emissions, and support the idea that the supply of sediment organic matter, rather than temperature, is the primary driver of ebullition (Grinham *et al.*, 2018). Batang Ai sediment properties and focusing patterns mentioned earlier could explain the model overestimation of CH₄ ebullition. The G-res model considers the fraction of littoral area and horizontal radiance (a proxy for heat input) as predictors of ebullition rate, but does not integrate other catchment properties. Building a stronger mechanistic understanding of the effect of sediment composition and accumulation patterns on CH₄ bubbling may improve our ability to more accurately predict reservoir ebullition flux.

Our empirical estimate shows that 409 (\pm 23) and 1798 (\pm 255) gCO₂eq m⁻² yr⁻¹ are emitted as CO₂ and CH₄ respectively downstream of the dam (including degassing), accounting for 89 % of Batang Ai GHG emissions (Table 1.2). Currently there are no available model predicting downstream GHG emissions from reservoirs, except the Gres model which is able to predict only the CH₄ degassing part of this flux. Modeled CH₄ degassing in Batang Ai is 468 (266-832) gCO₂eq m⁻² yr⁻¹ compared to an estimated 1342 (\pm 190) gCO₂eq m⁻² yr⁻¹ based on our measurements. Predictive variables used to model CH₄ degassing are modeled CH₄ surface diffusion (based on % littoral area and temperature) and water retention time (Table S1.3). In Batang Ai main basin, the strong and permanent stratification favors oxygen depletion in the hypolimnion which promotes deep CH₄ accumulation combined with a decoupling between surface and deep water layers. The model relies strongly on surface CH₄ patterns to predict excess CH₄ in the deep layer, which could explain why it underestimates CH₄ degassing in Batang Ai. Similar strong stratification patterns are ubiquitous in the tropics, with a recent study suggesting a large majority of tropical reservoirs are monomictic or oligomictic (Lehmusluoto *et al.*, 1997 ; Scott Winton *et al.*, 2019), hence more often stratified than temperate and boreal ones. This suggests that CH₄ degassing is potentially more frequently underestimated in low-latitude reservoirs. The G-res effort to predict CH₄ degassing is much needed given the importance of this pathway, and the next step would be to refine this model and develop predictions for other currently missing fluxes like CO₂ degassing and downstream emissions in the outflow. Our results suggest that improving latter aspects requires a better capacity to predict GHG accumulation in deep reservoirs layers across a wide range of stratification regimes.

	Diffusion		Ebullition	Ebullition Degassing		Downstream river		Total	
	CO_2	CH_4	N_2O	CH_4	CO_2	CH_4	CO_2	CH_4	
	Flux rate $(gCO_2eq m^{-2} yr^{-1})$								
Reservoir									
Measured	113 (± 22)	153 (± 22)	-2.1 (± 4)	3.4 (± 1.9)	247 (± 14)	1342 (± 190)	163 (± 9)	456 (± 65)	2475 (± 327)
G-res model	577 (509 - 655)	161 (132 - 197)	NA	52 (32 - 83)	NA	468 (266 - 832)	NA	NA	1258 (1041 - 1636)
Barros et al. model	4671	176	NA	NA	NA	NA	NA	NA	4847
Inflows									
Measured	156 - 9538	248 - 22510	NA	10377 - 20498	0	0	0	0	10781 - 52546
Total flux ($TgCO_2eq yr^{-1}$)									
Reservoir (meas.)	0.008	0.010	-0.0001	0.0002	0.017	0.092	0.011	0.031	0.169
River*	0 - 0.014	0 - 0.034	NA	0.016 - 0.031	0.000	0.000	0.000	0.000	0.016 - 0.08

Table 1.2: Estimated reservoir and inflow areal and total GHG fluxes to the atmosphere (± standard error for measuredvalues, or 95 % confidence interval based on model standard error for G-res values) from different pathways based onmeasured and modeled approaches.

*Represents the estimated pre-impounded river fluxes assuming they were similar to current fluxes from the reservoir inflows.

1.5 Conclusions

The comprehensive GHG portrait of Batang Ai highlights the importance of spatial and temporal sampling resolution and the inclusion of all flux components in reservoir GHG assessments. Gas dynamics downstream of the dam (degassing, outflow emissions and CH₄ oxidation), commonly not assessed in reservoir GHG studies, are major elements in Batang Ai. We suggest that these emissions could have been greatly diminished with a minor change to the dam design (shallower water withdrawal). Mitigating GHG emissions from future reservoirs depends on the capacity to predict GHG fluxes from all pathways. In this regard, the comparison between Batang Ai measured and modeled GHG flux estimates allowed us to identify knowledge gaps based on which we propose the four following research avenues. 1) Refine the modeling of reservoir CO₂ diffusion by studying its link with metabolism and organic matter leaching from different soil types. 2) Examine the potential for CH₄ ebullition in littoral zones in relation to patterns of organic matter sedimentation linked to morphometry. 3) Improve the modeling of CH₄ degassing by better defining drivers of hypolimnetic CH₄ accumulation, namely thermal stratification. 4) Gather additional field data on GHG dynamics downstream of dams (degassing, river emissions, and river CH₄ oxidation) in order to incorporate all components of the flux to the modeling of reservoirs C footprint.

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Table S1.1: Summary of the mean, standard deviation (SD), range (min and max), and number of observation (n) of physical and chemical variables measured at the surface of the reservoir and its inflows.

		Reservoir						Ir	nflows		
Variables	Units	Mean	SD	Min	Max	n	Mean	SD	Min	Max	n
Water temperature	°C	31	0.8	24	32	134	27	2.4	24	30	10
DOC	mg.L ⁻¹	0.9	0.2	0.3	1.8	115	0.8	0.4	0.4	1.3	8
TP	μg.L ⁻¹	5.9	2.4	2.8	23.3	114	20.7	7.6	11.7	36	8
TN	mg.L ⁻¹	0.11	0.04	0.06	0.33	115	0.14	0.04	0.07	0.19	8
Chla	μg.L ⁻¹	1.3	0.7	0.3	4.0	112	2.1	1.7	0.1	4.4	9
CO ₂ diffusion	mmol.m ⁻² .d ⁻¹	7.7	18.2	-30.8	80.0	129	137.3	192.4	9.7	593.9	8
CH4 diffusion	mmol.m ⁻² .d ⁻¹	0.6	0.6	0.03	3.7	129	30.7	37.4	1.2	113.4	8
N ₂ O diffusion	nmol.m ⁻² .d ⁻¹	-0.2	2.1	-7.1	2.7	15					0
[CO ₂]	µmol.L ⁻¹	16.3	5.2	6.8	36.2	132	60.1	34.9	10.1	113.5	10
[CH ₄]	µmol.L ⁻¹	0.2	0.2	0.0	1.2	137	6.6	4.3	0.6	14.8	10
[N ₂ O]	nmol.L ⁻¹	5.6	1.2	3.9	7.8	15	6.1		5.2	7.0	2



Figure S1.1: Profile example of water temperature (black squares), dissolved oxygen concentration (gray diamonds), CO₂ and CH₄ concentrations (black circles) and isotopic δ^{13} C signature (gray triangles) in the main basin of the reservoir right upstream of the dam in Aug 2018. The horizontal dashed line represent the water withdrawal depth.



Figure S1.2: Maps of the spatially interpolated surface CO_2 diffusive flux in Batang Ai reservoir for each sampling campaign. Unit is in mmol.m⁻².d⁻¹. Graph axes are the spatial coordinates (latitude and longitude).



Figure S1.3: Maps of the spatially interpolated surface CH₄ diffusive flux in Batang Ai reservoir for each sampling campaign. Unit is in mmol.m⁻².d⁻¹. Graph axes are the spatial coordinates (latitude and longitude).



Figure S1.4: Simulated downstream emissions (including degassing) of CO_2 (gray squares) and CH_4 (dark circles) under different water withdrawal depth raise. Simulated emissions do not take into account CH_4 oxidation in the outflow river.

Table S1.2: Equations used to derive modeled CO₂ and CH₄ emissions based on the G-res and Barros et al. models. Input variables are described in Table S1.3.

	G-res model	Barros et al. model
log10 (CO2 diffusion)	$\begin{array}{l} 1.7892 - 0.3364 \ log_{10} \ (Age) + 0.0400 \ Effective \ temperature \ CO_2 + \\ 0.06918 \ log_{10} \ (Reservoir \ area) + 0.0216 \ Soil \ C \ content + 0.1472 \\ \ log_{10} \ (TP) \end{array}$	3.06 - 0.16 log ₁₀ (Age) - 0.01 Latitude + 0.41 log ₁₀ (DOC input) - log ₁₀ (400)
log10 (CH4 diffusion)	0.8804 - 0.0116 Age + 0.6068 log ₁₀ (% littoral area / 100) + 0.04828 Effective temperature CH ₄	$\begin{array}{c} 1.33 - 0.36 \log_{10}{(\text{Age})} - 0.32 \log_{10}{(\text{Mean depth})} + 0.39 \log_{10}{(\text{DOC input})} - 0.01 \text{ Latitude} \end{array}$
log ₁₀ (CH ₄ ebullition)	-0.98574 + 1.0075 log ₁₀ (% littoral area / 100) + 0.04928 (Cumulative global horizontal radiance)	-
CH4 degassing	$\frac{10 (-5.5029 + 2.2857 \log 10 (Modeled CH4 diffusion) + 0.9866 \log 10 (Water residence time))}{Discharge / Reservoir area}$	-

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Input variable	Value	Unit	Source
% Littoral area	5.15	%	(1 - (1 - (3 / Maximum depth)) ((Maximum depth / Mean depth) - 1)) x 100
Age	33	year	Mean reservoir age at sampling years (2016 - 2018)
Annual runoff	2219	mm.yr ⁻¹	UNH/GRDC Composite Runoff Fields V1.0 (Fekete et al., 2002)
Catchment area	1142	km ²	GranD database V1.3 (Lehner et al., 2011)
Cumulative global horizontal radiance	56.4	kWh.m ^{-2.} yr ⁻¹	Surface meteorology and Solar Energy (NASA, 2008)
Discharge	106	m ³ .s ⁻¹	Average during the four sampling times (provided by Sarawak Energy)
DOC	0.8	mg.L ⁻¹	Average of measured surface values in this study
DOC input	81.93	mg.m ⁻² .d ⁻¹	(DOC x Volume) / (Reservoir area x Water residence time x 1000 x 365)
Effective temperature CH ₄	26.37	C	Mean annual temperature corrected for the effect on CH ₄ from G-res tool (Prairie <i>et al.</i> , 2017)
Effective temperature CO ₂	26.37	C	Mean annual temperature corrected for the effect on CO ₂ from G-res tool (Prairie <i>et al.</i> , 2017)
Latitude	1.16	Decimal degrees	Google earth
Maximum depth	85	m	GranD database V1.3 (Lehner et al., 2011) – Height dam wall
Mean depth	34.4	m	GranD database V1.3 (Lehner et al., 2011)
Modeled CH ₄ diffusion	134	gCO2eq.m ⁻² .yr ⁻¹	Average over 100 years of CH ₄ diffusion modeled based on the G-res equation (Table S2)
Reservoir area	68.4	km ²	GranD database V1.3 (Lehner et al., 2011)
Soil C content	2.41	kgC.m ⁻²	Estimated based on Wasli et al. 2011
TP	6.77	μg.L ⁻¹	Average of measured surface values in this study
Volume	2360 x 10 ⁶	m ³	GranD database V1.3 (Lehner et al., 2011)
Water residence time	0.93	year	(Mean depth x Reservoir area) / (catchment area x Annual runoff) x 1000

Table S1.3: Input variables values, units, and source used in the modeling of Batang Ai CO₂ and CH₄ emissions.

1.7.1 Modeling downstream river oxidation

The following section describes the conceptual framework underlying the isotopic model used to estimate CH_4 oxidation in the outflow of the reservoir. For that, we assume that the only source of water and CH_4 to the outflow (starting right downstream of the power house) is the reservoir discharge. We also assume oxidation and evasion to the atmosphere are the only two loss processes for CH_4 in the outflow, and that both reactions have a constant specific rate (they are a linear function of CH_4 concentration). When following a parcel of water travelling along the river, the change in CH_4 concentration can thus be described as:

$$\frac{d[CH_4]}{dt} = -k_{oxi}[CH_4] - k_{eva}[CH_4]$$
(S1.1)

with k_{oxi} and k_{eva} the rate coefficients of CH₄ oxidation and evasion respectively, and [CH₄] the CH₄ concentration. Integration of Eq. (S1.1) yields the following generic time-depend solution:

$$[CH_4]_t = [CH_4]_0 e^{-(k_{oxi} + k_{eva})t}$$
(S1.2)

The instantaneous oxidation rate at a given time (t) or at a given point in the river is equal to:

$$oxidation \ rate = k_{oxi} \ [CH_4]_t \tag{S1.3}$$

Combining Eq. (S2) and Eq. (S1.3) yields the following equation:

$$oxidation \ rate = k_{oxi} \ [CH_4]_0 \ e^{-(k_{oxi}+k_{eva}) t}$$
(S1.4)

Hence, the total amount of CH_4 oxidized between time 0 and time t (or between km 0 and t in the river) is derived from the integration of Eq. (S1.4):

amount oxidized =
$$\int_0^t k_{oxi} \, [CH_4]_0 \, e^{-(k_{oxi}+k_{eva}) \, t} \, dt$$
 (S1.5)

Simplifying Eq. (S1.5) yields:

amount oxidized =
$$[CH_4]_0 \left(\frac{k_{oxi}}{(k_{oxi}+k_{eva})}\right) \left(1 - e^{-(k_{oxi}+k_{eva})t}\right)$$
(S1.6)

The amount oxidized can also be calculated as the product of the original CH_4 concentration and the fraction of CH_4 oxidized (F_{ox}):

$$amount \ oxidized = [CH_4]_0 \ F_{ox} \tag{S1.7}$$

According to Eq. (S1.6) and Eq. (S1.7), we can derive F_{ox} from the following equation:

$$F_{ox} = \left(\frac{k_{oxi}}{(k_{oxi}+k_{eva})}\right) \left(1 - e^{-(k_{oxi}+k_{eva})t}\right)$$
(S1.8)

In Eq. (S1.8), the second term of the product can be rearranged using Eq. (S1.2) to represent the remaining fraction of CH_4 concentration at time or km t:

$$F_{ox} = \left(\frac{k_{oxi}}{(k_{oxi} + k_{eva})}\right) \left(1 - \frac{[CH_4]_t}{[CH_4]_0}\right)$$
(S1.9)

Given a fractionation factor α for CH₄ oxidation, the behavior of the two CH₄ isotopes (¹²CH₄ and ¹³CH₄) can be described based on Eq. (S1.2) as following:

$$[12CH_4]_t = [12CH_4]_0 \ e^{-(k_{oxi}^{12} + k_{eva})t}$$
(S1.10)

$$[13CH_4]_t = [13CH_4]_0 e^{-\left(\frac{k_{oxi}^{12}}{\alpha} + k_{eva}\right)t}$$
(S1.11)

With [12CH₄] and [13CH₄] the concentrations of the two isotopes, and k_{oxi}^{12} the rate coefficient of oxidation for the ¹²CH₄ isotope. The isotopic ratio of CH₄ concentration at t (R_t) can be derived from Eq. (S1.10) and Eq. (S1.11) as following:

$$R_{t} = \frac{[13CH_{4}]_{t}}{[12CH_{4}]_{t}} = \frac{[13CH_{4}]_{0} e^{-(\frac{k_{oxi}}{\alpha} + k_{eva})t}}{[12CH_{4}]_{0} e^{-(k_{oxi} + k_{eva})t}}$$
(S1.12)

By simplifying Eq. (S1.12), R_t can be written as:

$$R_t = R_0 \, e^{(1 - \frac{1}{\alpha}) \cdot k_{oxi} t} \tag{S1.13}$$

The term t can be isolated from Eq. (S1.13):

$$t = \frac{\ln({^{R_t}/_{R_0}})}{(1 - \frac{1}{\alpha}) \cdot k_{oxi}}$$
(S1.14)

Similarly, the term t can be isolated by rearranging Eq. (S1.2):

$$t = \frac{\ln([CH_4]_t/[CH_4]_0)}{-(k_{oxi}+k_{eva})}$$
(S1.15)

Combining Eq. (S1.14) and Eq. (S1.15) results in:

$$\frac{\ln({}^{R_{t}}/_{R_{0}})}{(1-\frac{1}{\alpha})\cdot k_{oxi}} = \frac{\ln({}^{[CH_{4}]_{t}}/_{[CH_{4}]_{0}})}{-(k_{oxi}+k_{eva})}$$
(S1.16)

Rearranging E.S16 yields:

$$\frac{-\ln({^{R_t}/_{R_0}})}{(1-\frac{1}{\alpha})\cdot\ln({^{[CH_4]t}/_{[CH_4]_0}})} = \frac{k_{oxi}}{(k_{oxi}+k_{eva})}$$
(S1.17)

Using Eq. (S1.9) and Eq. (S1.17), we can derive F_{ox} as:

$$F_{ox} = \frac{k_{oxi}}{(k_{oxi} + k_{eva})} \left(1 - \frac{[CH_4]_t}{[CH_4]_0} \right) = \frac{-\ln({^Rt}/{_{R_0}}) \cdot (1 - {[CH_4]t}/{_{[CH_4]_0}})}{(1 - \frac{1}{\alpha}) \cdot \ln({[CH_4]t}/{_{[CH_4]_0}})}$$
(S1.18)
The isotopic ratios (R_t and R_0) can be converted to $\delta^{13}CH_4$ expressed in ‰ using the following standard isotopic equation calculation:

$$\delta^{13} CH_4 = \left(\frac{[13CH_4]:[12CH_4]}{[13CH_4]_{std}:[12CH_4]_{std}} - 1\right) \times 1000$$
(S1.19)

with $[13CH_4]_{std}$: $[12CH_4]_{std}$ the standard reference Pee Dee Belemite isotopic carbon ratio. The conversion of R_t and R_0 to $\delta^{13}CH_4$ in Eq. (S1.18) results in the following equation for F_{ox} calculation:

$$F_{ox} = \frac{-\left[\ln\left(\delta^{13}CH_{4_{t}} + 1000\right) - \ln\left(\delta^{13}CH_{4_{0}} + 1000\right)\right] \cdot \left(1 - \frac{[CH_{4}]t}{a}\right)}{\left(1 - \frac{1}{a}\right) \cdot \ln\left(\frac{[CH_{4}]t}{[CH_{4}]_{0}}\right)}$$
(S1.20)

Eq. (S1.20) was used to determine F_{ox} and CH₄ oxidation (product of F_{ox} and [CH₄]₀) in Batang Ai river outflow using km 0 and 19 as a start and end points of the river stretch. Sampling of CH₄ concentration and isotopic signature was conducted at two other points within this river stretch (km 0.6 and 2.7). This additional sampling resolution can help test the adequacy of the model and its conceptual assumptions by recalculating CH₄ oxidation in individual segments (between 0-0.6 km, 0.6-2.7 km, and 2.7-19 km). If the model assumptions are correct (absence of other CH₄ source in the river, constant specific oxidation rate, and constant river velocity), the amount of CH₄ oxidized in the entire stretch (Ox_{total}) should equal the sum of the amount oxidized in each individual segment (Ox_{sum}). In two of the sampling campaigns (Apr-May 2017 and Feb-Mar 2018), the difference between Ox_{total} and Ox_{sum} was very small (CV < 5 %). In the two other campaigns, this difference was larger (CV of 12 and 35 % for Nov-Dec 2016 and Aug 2018 respectively), due to an additional CH₄ source causing occasional increases of CH₄ concentration along the stretch. Overall, model assumptions appear reasonable, however, to avoid overestimating the amount oxidized in cases of additional CH₄ sources, we considered Ox_{total} as the more suitable estimate, since it is less influenced by CH₄ addition in individual segments.

CHAPITRE II

QUANTIFYING SOURCES AND PROCESSES SUSTAINING SURFACE CO₂ AND CH₄ FLUXES IN A TROPICAL RESERVOIR

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N.B. References cited in this chapter are presented at the end of the thesis.

2.1 Abstract

Freshwaters are important emitters of carbon dioxide (CO₂) and methane (CH₄) to the atmosphere, two potent greenhouse gases (GHG). While aquatic surface GHG fluxes have been extensively measured, there is much less information on their underlying sources. CO₂ and CH₄ air-water fluxes can originate from different sources in both lakes and reservoirs: horizontal riverine flow, vertical flux from the bottom layer, sediment inputs, and water column metabolism. These processes are generally studied separately, leading to a fragmented assessment of their relative role in sustaining CO_2 and CH₄ surface fluxes. In this study, we quantified sources / sinks of CO₂ and CH₄ in the epilimnetic layer of a permanently stratified tropical reservoir located on the Borneo Island. CO_2 and CH_4 concentrations, stable isotopes, and fluxes were measured at the sediment-water and air-water interfaces as well as within the water column in different locations and seasons. Results showed that horizontal inputs is an important source of both CO₂ and CH₄ (18 to 100 % of surface emissions) in the upstream reservoir branches. However, this contribution fades along the hydrological continuum, becoming negligible in the main basin of the reservoir, where CO₂ and CH₄ are uncoupled and driven by different processes. In the main basin, vertical CO₂ inputs and sediment CH₄ inputs contributed to on average 60 and 23 % respectively to the surface fluxes of the corresponding gas. Water column metabolism generally played a major role in the epilimnetic budget for CO₂ and CH₄. However, net metabolism was highly variable for both gases, shifting from positive to negative values, thus making it the most important but uncertain component of the epilimnetic gas budgets. Overall, our results highlight the importance of assessing all the components sustaining CO₂ and CH₄ surface fluxes and identify clear knowledge gaps hindering our predictive capacity of aquatic GHG fluxes.

2.2 Introduction

Surface inland waters are globally significant sources of greenhouse gases (GHG) to the atmosphere, namely carbon dioxide (CO₂) and methane (CH₄) (DelSontro et al., 2018a). Freshwaters act as both transport vessels for terrestrial carbon (C) and as active biogeochemical processors, making them key sites of GHG exchange with the atmosphere (Tranvik et al., 2018). The impoundment of rivers for hydropower generation, irrigation, flood control or other purposes, changes the landscape and its C cycling (Maavara et al., 2017), often resulting in increased aquatic CO₂ and CH₄ emissions due to the decay of flooded organic matter (Prairie et al., 2018; Venkiteswaran et al., 2013). Globally, reservoirs are estimated to emit between 0.5 and 2.3 PgCO₂eq.yr⁻¹ (Barros et al., 2011; Bastviken et al., 2011; Deemer et al., 2016; St. Louis et al., 2000), and this number is predicted to increase with a rapid growth of the hydroelectric sector in the upcoming decades (Zarfl et al., 2015). Several studies have focused on quantifying GHG surface diffusion from reservoirs around the world and have found extremely high variability temporally and spatially (Barros et al., 2011; Deemer et al., 2016), like it is the case in natural lakes (DelSontro et al., 2018a; Raymond et al., 2013). However, less research exists on the sources and processes sustaining latter surface diffusive fluxes and their variability, especially in reservoirs.

GHG sources to surface waters can be both internal and external. The magnitude of allochthonous inputs, namely terrestrial organic and inorganic C, is known to increase with the soil-water connectivity (Hotchkiss *et al.*, 2015), and with soil C content and leaching capacity (Kindler *et al.*, 2011 ; Li *et al.*, 2017 ; Monteith *et al.*, 2007). Soil-derived gas inputs are also temporally variable, generally increasing with discharge, like during storm events (Vachon et del Giorgio, 2014) or during rainy seasons (Kim *et al.*, 2000 ; Zhang *et al.*, 2019). Terrestrial inputs in the form of organic C can indirectly sustain surface GHG emissions by fueling lake / reservoir in situ organic matter aerobic and anaerobic respiration (Karlsson *et al.*, 2007 ; Pace et Prairie, 2005 ;

Rasilo et al., 2017). The metabolic balance between aerobic ecosystem respiration (ER) and gross primary production (GPP), and between methanogenesis and CH₄ oxidation, control the net internal production of CO₂ and CH₄ respectively, and in turn influence their surface fluxes. Respiration and primary production rates are highly variable in space and time, and generally a function of temperature, organic C content and nutrients (Hanson et al., 2003; Pace et Prairie, 2005; Prairie et al., 1989; Solomon et al., 2013). Net heterotrophy (ER > GPP) is mainly associated with systems receiving high external inputs of organic C (Bogard et al., 2020; Tank et al., 2010; Wilkinson et al., 2016), while net autotrophy (ER < GPP) has been associated to highly productive nutrient-rich systems (Hanson et al., 2003; Sand-Jensen et Staehr, 2009). However, a large part of the variability in measured metabolic rates remains unexplained (Bogard et al., 2020; Coloso et al., 2011; Solomon et al., 2013) and our ability to predict their net balance is still weak in part because the two processes are often studied separately. Additionally, anaerobic C degradation adds another level of complexity to the C metabolic balance by decoupling GPP and ER (Bogard et del Giorgio, 2016; Martinsen et al., 2020; Vachon et al., 2020). For instance, methanogenesis transforms organic C to CH_4 instead of CO_2 , a gas product with a radiative forcing 34 times higher than CO_2 over a 100 years horizon. CH₄ is known to be mostly produced in both profundal and littoral sediments, and reach the water surface by vertical or lateral diffusive processes (Bastviken et al., 2008; DelSontro et al., 2018b; Encinas Fernández et al., 2014; Guérin *et al.*, 2016). However, there is increasing evidence that CH_4 production in the oxic water column contributes significantly to lakes CH4 emissions (Bižić et al., 2019; Bogard et al., 2014; DelSontro et al., 2018b; Donis et al., 2017; Tang et al., 2014). Methanogenesis can be counter-balanced by the oxidation of CH_4 to CO_2 mainly in oxic and hypoxic environments (Conrad, 2009; Reis et al., 2020b; Thottathil et al., 2019). While several studies have measured rates of CH₄ production and oxidation in lakes and reservoirs, few have quantified the net balance of these two processes at an ecosystem-scale (Bastviken et al., 2008; Schmid et al., 2007; Vachon et al., 2019). According to Vachon et al. (2019), this balance is tightly controlled by physical

processes within the water column. Physical mixing in lakes and reservoirs indirectly impacts C metabolic processes by shaping the O_2 profile, and directly affects GHG surface diffusion by controlling the transport of CO_2 and CH_4 from bottom to surface water layers (Barrette et Laprise, 2005; Kreling *et al.*, 2014; Pu *et al.*, 2020). Despite its potential importance (Kankaala *et al.*, 2013a), very few studies have focused on quantifying vertical gas transport and the role of this process in fueling surface GHG diffusion.

Understanding what regulates surface CO₂ and CH₄ concentrations and fluxes to the atmosphere requires knowledge of the interplay between all physical and biogeochemical processes involved. While a number of studies have assessed some processes individually or by difference, very few have measured all relevant components of the epilimnetic mass-balance simultaneously. Here we report on a field study in a tropical hydropower reservoir quantifying external inputs, sediments inputs, net CO₂ and CH₄ metabolism, vertical diffusion from deeper layers and gas exchange at the air-water interface. This allowed us to estimate the relative contribution of each process in shaping surface GHG emissions from a tropical East Asian reservoir, and to test whether the epilimnetic mass balance can be closed. The two major rivers feeding the Batang Ai reservoir flow into two elongated reservoir branches, acting as a transition zones, before reaching the reservoir main basin. This configuration, common in reservoirs, allowed us to quantify and compare epilimnetic CO₂ and CH₄ regulation in two morphometrically different areas of the reservoir (branches and main basin). Overall, the aim of this study is to provide ecosystem-scale portrait of the processes sustaining surface CO₂ and CH₄ emissions and examine how they change when transitioning from a river delta to the open basin.

2.3 Materials and methods

2.3.1 Site and sampling description

The study was conducted in Batang Ai hydroelectric reservoir in Sarawak Malaysia (latitude 1.16° and longitude 111.9°). The reservoir is located on the Borneo Island in a tropical equatorial climate with a constantly high temperature averaging 23°C and 32°C during nighttime and daytime respectively (Sarawak Government, 2019). The region experiences two weak monsoon seasons (November to February and June to October) with a yearly average rainfall of 3300 to 4600 mm (Sarawak Government, 2019). The reservoir was impounded in 1985 with a dam wall of 85 m, a surface area of ~68.4 km² and a watershed area of 1149 km² of mostly undisturbed forested land (limited rural habitations and small scale croplands). The reservoir water residence time is 0.9 year.

We distinguish between three sections of the study site: inflows, reservoir branches, and reservoir main basin shown in Figure 2.1. The inflows are the two main reservoir inlets: Batang Ai and Engkari rivers (3 to 10 m deep where sampled). The two rivers flow into two arms that we refer to as the reservoir branches (10.8 km², mean and max depths of 18 and 52 m respectively). The reservoir branches merge into the main basin of the reservoir (58.9 km², mean and max depths of 30 and 73 m respectively). Surface sampling was performed in 36 sites across the three study sections, and water column profile sampling (from 0 up to 32 m, each 0.5 to 3 m) was done in 9 sites in the reservoir branches and main basin (Figure 2.1). Sampling was repeated (with a few exceptions) during four campaigns: 1) November 14th to December 5th 2016 (Nov-Dec 2016), 2) April 19th to May 3th 2017 (Apr-May 2017), 3) February 28th to March 13th 2018 (Feb-Mar 2018), and 4) August 12th to 29th 2018 (Aug 2018).



Figure 2.1: Map of Batang Ai reservoir with delimited sections (branches and main basin) and sampling points. * Represents sampling points at the branches extremities.

2.3.2 Physical and chemical analyses

Water temperature, dissolved oxygen, and pH were measured using a multi-parameter probe (YSI model 600XLM-M) equipped with a depth gauge and attached to a 12 Volt submersible pump (Proactive Environmental Products model Tornado) for water samples collection. Concentrations of dissolved organic carbon (DOC), total phosphorus (TP), total nitrogen (TN), and chlorophyll a (Chla) were measured during all campaigns in all surface sampling sites (Figure 2.1). Methods for latter analyses are described in detail in Soued et Prairie (2020). Briefly, TP and Chla (extracted with hot ethanol) were analyzed via spectrophotometry, and TN and DOC (filtered at 0.45 µm)

were measured on an Alpkem Flow Solution IV autoanalyser and on a Total Organic Carbon analyser 1010-OI respectively.

For each site, we defined the depths of the thermocline and the top and bottom of the metalimnion based on measured temperature profiles using the R package rLakeAnalyzer (Winslow *et al.*, 2018). The epilimnion was defined from the surface to the top of the metalimnion, and was assumed to be a mixed layer.

2.3.3 Gas concentration, isotopic signature, and water-air fluxes

CO₂ and CH₄ gas concentrations and isotopic signatures (δ^{13} C) were measured in duplicates at the surface in 36 sites and along vertical profiles in 9 sites (P1 to P9, Figure 2.1) using the headspace technique described in details in Soued et Prairie (2020). In brief, sampling was done by equilibrating the water sample with an air headspace inside a 60 mL syringe. The gas phase was then injected in a 12 mL prevacuumed air-tight vial, and analyzed on a gas chromatograph (Shimadzu GC-8A) for gas concentrations, and on a Cavity Ring Down Spectrometer (CRDS) equipped with a Small Sample Isotopic Module (SSIM, Picarro G2201 -i) for δ^{13} CO₂ and δ^{13} CH₄.

Surface diffusive fluxes of CO_2 and CH_4 were measured at all surface sampling sites during each campaigns. Flux rates were derived from linear changes in CO_2 and CH_4 concentrations in a static floating chamber connected in a closed loop to a gas analyzer (model UGGA, from Los Gatos Research). Measured gas concentrations, isotopic signature, and fluxes were spatially interpolated to the whole reservoir area using package gstat version 1.1-6 in the R version 3.4.1 software (Pebesma, 2004). Mean values were calculated for each campaign based on the interpolated maps (Soued et Prairie, 2020).

2.3.4 Horizontal GHG inputs

CO₂ and CH₄ inputs from the horizontal water flow (H) were estimated using Eq. (2.1):

$$H = C_{in} Q f_{epi} \tag{2.1}$$

With C_{in} the concentration of gas in the inflowing water, Q the reservoir water discharge, and f_{epi} the fraction of water flowing into the epilimnion. Q was taken as the mean of daily discharge during each campaign (considering minimal changes in inflow / outflow rates during campaigns). f_{epi} was estimated based on temperature profiles at the two ends of the right branch. To estimate gas inputs form the inflows to the branches, C_{in} was considered as the average of gas concentrations measured at the two upstream extremities of the branches (Figure 2.1). To estimate gas inputs form the branches to the main basin, C_{in} was considered as the gas concentrations measured at the confluence between the two branches (right upstream of the main basin).

2.3.5 Vertical GHG fluxes

We estimated CO₂ and CH₄ fluxes from the metalimnion to the epilimnion (F_v) based on the vertical gas diffusivity (K_z) and the gradient in gas concentration across the epilimnion-metalimnion interface using the Eq. (2.2) (Wüest et Lorke, 2009):

$$F_{\nu} = K_z \left(C_{meta} - C_{epi} \right) \tag{2.2}$$

where C_{meta} and C_{epi} are the gas concentrations at the top of the metalimnion and at the bottom of the epilimnion respectively, measured in profile sites (P1 to P9, Figure 2.1). K_z was derived from the following Eq. (2.3) (Osborn, 1980):

$$K_z = \Gamma \frac{\epsilon}{N^2} \tag{2.3}$$

where Γ is the mixing ratio set to 0.2 (Oakey, 1982), ϵ is the dissipation rate of turbulent kinetic energy, and N^2 is the buoyancy frequency. N^2 was calculated from measured temperature profiles (YSI probe) using function buoyancy.freq from the rLakeAnalyzer package (Winslow *et al.*, 2018) in the R software (R Core Team, 2017).

 ϵ was derived from measured vertical shear microstructure profiles performed in the Aug 2018 campaign in all profile sites shown in Figure 2.1 (except P1 due to floating logs). Shear profiles were measured with a high frequency (512 Hz) MicroCTD profiler (Rockland Scientific) equipped with two velocity shear probes, two thermistors, tilt and vibration sensors, and a pressure sensor. At each site, the profiler was cast 10 times, 5 with an uprising configuration (from bottom to top of the water column) and 5 with a downward configuration (top to bottom), with a 4 min waiting time between profiles to allow water column disturbance to subside. Data quality check and ϵ calculation for each profile cast were performed with ODAS v4.3.03 Matlab library (by Rockland Scientific) based on Nasmyth shear spectrum (Oakey, 1982), with ϵ values averaged among the two shear probes and binned over 1-2 m segments along the profile. For each site, continuous ϵ profiles were interpolated by fitting a smooth spline through all ϵ values from replicate casts as a function of depth.

At the epilimnion-metalimnion interface (top of the metalimnion ± 2 m), calculated ϵ averaged 7.7 x10⁻⁹ (range from 3.4 x 10⁻⁹ to 1.6 x 10⁻⁸) m².s⁻³ across all sites sampled with the microCTD, with no significant difference between the main basin and branches sites. In order to estimate vertical gas diffusion, we applied the latter ϵ average to Eq. (2.2) and (2.3) for all measured gas profiles (except P1). The resulting F_v values for each gas were averaged across sites in the main basin and branches separately to derive estimates of F_v for each of these two reservoir sections.

2.3.6 Sediment GHG inputs

In order to estimate CO₂ and CH₄ inputs from the sediments to epilimnetic waters, we calculated sediment to water gas fluxes based on gas profiles in sediment cores collected in Apr-May 2017 and Feb-Mar 2018 at 7 sites (P1 to P3 in the reservoir branches and P4, P5, P7, and P9 in the main basin, Figure 2.1). Sediment cores were collected using a Glew gravity corer attached to a 6 cm wide plastic liner. The liner was pre-drilled with 1 cm holes covered with electric tape at each centimeter up to 40

cm. Upon recovery of the sediment core, 3 mL tip-less syringes were inserted into each hole to extract sediments from each centimeter. The sediment content of each syringe was emptied into a 25 mL glass vial prefilled with 6 mL nano-pure water and immediately air-tight sealed by a butyl rubber stopper crimped with an aluminum cap. Glass vials were pressurized with 40 mL of ambient air using a plastic syringe equipped with a needle to pierce the rubber cap. Glass vial were shaken for 2 min for equilibration before extracting the gas with a syringe and injecting it into a pre-evacuated air-tight vial for analysis CO₂ and CH₄ concentration and isotopic signature as described above. Additionally, samples of the water overlaying the sediments (~1 cm above) were collected for similar analyses of CO₂ and CH₄.

Sediment CO₂ and CH₄ flux rates to the overlaying water column were derived from the vertical gradient of gas concentration measured in the sediment cores and overlaying water (Berg *et al.*, 1998). The slope of CO₂ or CH₄ concentration as a function of depth (*g*, in µmol.L⁻¹.m⁻¹) was calculated for measured values in the first 5 cm of sediments and overlaying water. Most cores exhibited clear linear slopes (p-value < 0.05 and R²_{adj} > 0.5). In the few cases where a linear slope was not evident, *g* was replaced by the gradient between the mean gas concentration in the first 3 cm and the overlaying water. The sediment gas flux rate (S_f in mmol.m⁻².d⁻¹) were calculated with Eq. (2.4):

$$S_f = \frac{g \times d}{p} \tag{2.4}$$

With *d* the diffusion coefficient set to $1.5 \times 10^{-5} \text{ cm}^2 \text{.s}^{-1}$ (Donis *et al.*, 2017), and *p* the sediment porosity assumed to be 2 % based on previous results in Batang Ai (Tan, 2015).

At an ecosystem scale, sediment CO_2 and CH_4 inputs to the water column (S) were estimated based on average and standard deviation values of sites located in each

section of the reservoir (branches and main basin). For each section, mean sediment CO_2 and CH_4 flux rates were multiplied by the areal ratio of epilimnetic sediments (A_{epi}) versus total water area (A₀). The latter ratio was calculated based on the hypsometric model (Ferland *et al.*, 2014 ; Imboden, 1973) as shown in Eq. (2.5) to (2.7):

$$q = \left(\frac{z_{max}}{z_{mean}}\right) - 1 \tag{2.5}$$

$$A_{epi} = A_0 \left(1 - \left(1 - \left(\frac{z_{epi}}{z_{max}} \right) \right)^q \right)$$
(2.6)

$$S = \frac{A_{epi}}{A_0} Sf \tag{2.7}$$

with q a parameter describing the general shape of the reservoir section, z_{max} and z_{mean} the maximum and mean depths respectively, and z_{epi} the mean depth of the epilimnion (8.0 and 10.5 m in the branches and main basin respectively).

2.3.7 Metabolic rates

Net metabolic rates of CO₂ and CH₄ production in the epilimnetic water column were estimated with *in situ* incubations. Incubations were performed in 5 sites (P2 and P3 in the branches and P4, P5 and P7 in the main basin, Figure 2.1). Water from 3 m deep was pumped into 5 L transparent glass jars with an air tight clamp lid. Before closing, jars were filled from the bottom and allowed to overflow, then sampled for initial CO₂ and CH₄ concentrations. Closed jars were fixed at 3 m to an anchored line in the reservoir sampling site, and incubated in *in situ* temperature and light conditions for 22.0 to 24.2 hours. Upon retrieval, samples of final CO₂ and CH₄ production were collected form the jars. Volumetric daily rates of net CO₂ and CH₄ production were

calculated based on the difference between final and initial gas concentrations rescaled to a 24 h period.

In addition to incubations, open water high frequency O_2 measurements were carried out to derive CO_2 metabolism on larger spatial and temporal scales. Rates of gross primary production (GPP), aerobic ecosystem respiration (ER), and net ecosystem production (NEP) were estimated in the reservoir surface layer by monitoring and inverse modeling diel O_2 changes in the epilimnion. O_2 was measured at a one minute interval using high frequency O_2 and temperature sensors (model miniDOT from Precision Measurement Engineering). Sensors were deployed in profile sites P1 to P3 in the branches and P4, P5, P7, and P9 in the main basin (Figure 2.1). Note that not all sites were sampled in all sampling campaigns. Sensors were attached to an anchored line at a depth of 0.7 to 3 m and deployment time varied between 4 days and two weeks. Upon retrieval of the sensors, a first data quality check and selection was made based on the sensor internal quality index and visual screening. Rates of ecosystem metabolism were then estimated based on an open system diel O_2 model (Odum, 1956), where change in O_2 concentration is a function of GPP, ER, and air-water gas exchange (K_{O2}) following Eq. (2.8) (Hall et Hotchkiss, 2017):

$$\frac{dO2}{dt} = \frac{GPP}{z_{epi}} + \frac{ER}{z_{epi}} + K_{O2} \left(O_{2_{sat}} - O_{2} \right)$$
(2.8)

with O_{2sat} the theoretical O_2 concentration at saturation considering the *in situ* temperature and atmospheric pressure, and O_2 is the actual measured O_2 concentration in the water. A detailed description of the model equations can be found in (Hall et Hotchkiss, 2017). Daily estimates of GPP, ER, and K_{O2} were derived by maximum likelihood fitting of the data to the model in Eq. (2.8) using the R package StreamMetabolizer (Appling *et al.*, 2018). In some cases, where the best K_{O2} was negative, the fitting process was rerun with a user defined positive K_{O2}, either equal to

a value estimated for the previous or subsequent day at the same site (range of $0.03 - 0.96 d^{-1}$) or fixed to $0.1 d^{-1}$ (if no other available estimate). A final selection of daily metabolic estimates was done based on the model goodness of fit assessed by calculating Pearson correlation coefficient between modeled and measured O₂ values and discarding days with a correlation lower than 0.9. Based on GPP and ER estimates, we calculated daily NEP as the balance between these two processes, and converted it to net CO₂ production rate by assuming a 1:1 metabolic quotient.

Average estimates of areal metabolic rates per campaign were obtained for the branches and main basin by first averaging data within each site and then across sites for each reservoir section. Note that one value derived from incubations was excluded from the calculation of the average net CH₄ production rate in the branches due to its high value of initial CH₄ concentration (an order of magnitude higher than in all other incubations and all epilimnetic data from this site). The high CH₄ concentration, unrepresentative of real conditions, was probably caused by CH₄ contamination during sampling, and triggered a high oxidation rate that would overestimate the real ecosystem average rate if included.

2.4 Results

2.4.1 Physical and chemical properties

Surface water temperature exhibited a marked increase from the inflows to the branches, averaging 27.1 and 30.7 °C respectively (Table 2.1). There was no difference in surface water temperature between the branches and the main basin. The depth of the epilimnion tended to increase and become more stable along the water flow, going from 1.3 (\pm 1.6) m in the Batang Ai river delta, to 8.0 (\pm 2.3) m in its branch, and 10.6 (\pm 1.7) m in the main basin (Table 2.1). Light penetration exhibited the same spatial pattern, with an increasing Secchi depth along the water flow, from a mean of 1.3 in

the inflows to 5.1 and 5.5 in the branches and main basin respectively (Table 2.1). All sections of the study system (reservoir inflows, branches, and main basin) exhibited oligotrophic water properties (Table 2.1).

Variables	Units	Inflows	Branches	Main basin
Z _{epi}	m	1.3 (± 1.6)	8 (± 2.3)	10.6 (± 1.7)
Secchi	m	1.2 (± 0.9)	5.1 (± 1.2)	5.5 (± 1.2)
Temp.	°C	27.1 (± 2.5)	30.7 (± 0.5)	30.6 (± 0.5)
рН		6.5 (± 0.3)	7.2 (± 0.2)	7.2 (± 0.2)
O ₂	%	94.9 (± 7.7)	102.7 (± 4.5)	99.3 (± 4.8)
DOC	mg C.L ⁻¹	0.8 (± 0.4)	0.9 (± 0.2)	0.9 (± 0.2)
ТР	µg P.L⁻¹	20.7 (± 7.6)	6.2 (± 1.7)	5.8 (± 2.6)
TN	mg N.L ⁻¹	0.14 (± 0.04)	0.12 (± 0.04)	0.1 (± 0.03)
Chla	µg.L⁻¹	2.1 (± 1.7)	1.7 (± 1)	1.2 (± 0.5)

Table 2.1: Mean (\pm SD) of physical and chemical variables measured at the surface
of the three reservoir sections.

2.4.2 Surface GHG concentrations, fluxes, and isotopic signatures

Surface CO_2 and CH_4 patterns are summarized in Figure 2.2, presenting campaign averages of spatially interpolated gas concentration, flux, and isotopic signature along the different reservoir sections. Despite the temporal variability, the gas patterns along the water flow are robust, remaining similar throughout time (Figure 2.2).

Average CO₂ air-water flux and surface concentration were systematically higher in the inflows (mean [range]: 135.3 [18.9 – 368.8] mmol.m⁻².d⁻¹ and 58.0 [24.5 – 113.0] μ mol.L⁻¹, respectively) compared to the branches (4.7 [-3.4 – 15.2] mmol.m⁻².d⁻¹ and 15.4 [12.2 – 19.3] μ mol.L⁻¹) or the main basin (7.5 [0.3 – 15.1] mmol.m⁻².d⁻¹ and 16.0 [14.2 – 17.7] μ mol.L⁻¹) (Figure 2.2). Surface CO₂ concentration in the reservoir (branches and main basin) was most strongly correlated inversely with water temperature (R²_{adj} = 0.22, p-value < 0.001, Figure S2.1 and Table S2.1). Except for the

Apr-Mar 2017 campaign, there is a modest increase (2.2 to 3.3 ‰) of surface $\delta^{13}CO_2$ towards more enriched values from the inflows to the branches (Figure 2.2).

Similarly, surface CH₄ flux and concentration continually decreased along the water channel, being an order of magnitude higher in the inflows compared to the branches, and about twice as high in the branches compared to the main basin (Figure 2.2). Of all measured water properties, TN was the most strongly linked to reservoir surface CH₄ concentration ($R^2 = 0.14$, p-value < 0.001, Figure S2.1 and Table S2.1). In the main basin surface CH₄ concentration significantly decreased with distance to shore in Nov-Dec 2016 ($R^2_{adj} = 0.54$, p-value < 0.001), but this correlation was weaker ($R^2_{adj} \le 0.13$, p-value ≥ 0.03) during other sampling campaigns (Figure 2.6). Surface δ^{13} CH₄ values varied widely, between -83.3 and -47.6 ‰, but did not show a consistent spatial pattern (Figure 2.2) apart from a positive correlation with distance to shore in the main basin in Nov-Dec 2016 ($R^2_{adj} = 0.29$, p-value = 0.01, Figure 2.6).

The degree of coupling between CO₂ and CH₄ also followed a clear spatial pattern. While CO₂ and CH₄ surface concentrations were strongly linked in the inflows ($R^2_{adj} = 0.54$, p-value = 0.006), they became only weakly correlated in the branches ($R^2_{adj} = 0.17$, p=0.005) and not correlated at all in the main basin ($R^2_{adj} = 0.01$, p-value = 0.11) (Figure S2.2).



Figure 2.2: Average of spatially interpolated surface CO_2 and CH_4 fluxes, concentrations, and isotopic signatures along the hydrological continuum from the reservoir inflows to the main basin in all sampling campaigns (blue: Nov-Dec 2016, orange: Apr-May 2017, green: Feb-Mar 2018, red: Aug 2018).

2.4.3 Horizontal GHG flow

Horizontal inputs from the inflows to the surface layer of the branches were estimated to vary between $0.34 - 0.71 \text{ mol.s}^{-1}$ for CO₂ and $0.02 - 0.25 \text{ mol.s}^{-1}$ for CH₄. When expressed as rates over the branches surface area (to facilitate comparison with other components), this results in 2.7 - 5.7 and $0.16 - 1.97 \text{ mmol.m}^{-2}$.d⁻¹ for CO₂ and CH₄ respectively (Table S2.2 and S2.3). These values are in the same order of magnitude as surface fluxes calculated in the branches (Figure 2.6 and Table S2.2 and S2.3). However, the effect of horizontal inputs faded spatially, with much lower inputs from the branches to the main reservoir basin, averaging 0.31 and 0.004 mmol.m⁻².d⁻¹ for CO₂ and CH₄, respectively (Figure 2.6 and Table S2.2 and S2.3). For CH₄, this fits spatial and temporal surface flux measurements, being systematically higher in the branches, and maximal during the two sampling campaigns with the highest recorded horizontal inputs from the inflows (Table S2.3). In contrast, CO₂ surface flux was typically lower (sometimes negative) in the branches compared to the main basin, despite substantial riverine inputs to the branches (Table S2.2).

2.4.4 Vertical GHG fluxes

Vertical fluxes depend on the gas diffusivity and concentration gradient. Gas diffusivity is a function of the strength of stratification (N²) and energy dissipation rate (ϵ). Measured values of N² and ϵ at the varied widely, from 5.9 x 10⁻⁵ to 2.3 x 10⁻³ s⁻² and from 3.4 x10⁻⁹ to 1.6 x 10⁻⁸ m² s⁻³ respectively, yielding a median K_z value of 0.13 m².d⁻¹ (range from 0.01 to 2.28 m².d⁻¹), but with no clear differences between the reservoir branches and main basin (Figure S2.3). Similarly, CO₂ and CH₄ concentration gradients varied substantially in both space and time (from -18.4 to 94.3 µmol L⁻¹ m⁻¹ for CO₂ and -0.19 to 0.4 µmol L⁻¹ m⁻¹ for CH₄). CO₂ concentration generally increased from the epilimnion to the metalimnion as a result of the respiratory CO₂ buildup in the deep layer. On rare occasions, an inverse gradient was observed, possibly due to autotrophic activity in the metalimnion. For CH₄, metalimnion to epilimnion

concentration gradients were generally modest averaging 0.04 μ mol L⁻¹ m⁻¹, and even negative in one third of the profiles leading to the diffusion of epilimnetic CH₄ toward deeper layers instead of the reverse. The low to negative CH₄ vertical flux results from a highly active methanotrophic layer reducing CH₄ concentration in the metalimnion, as evidenced by the strong enrichment effect observed in δ^{13} CH₄ profiles (Figure S2.4). The combination of vertical diffusivity and gas concentration gradients resulted in vertical fluxes averaging 3.4 (-1.8 to 20.5) mmol m⁻² d⁻¹ for CO₂, and 0.01 (-0.01 to 0.09) mmol m⁻² d⁻¹ for CH₄, with no significant differences between the reservoir branches and main basin (Figure S2.3).

2.4.5 GHG inputs from littoral sediments

Areal Gas fluxes from the sediments ranged from 1.2 to 4.0 and -0.29 to 1.10 mmol.m⁻ 2 .d⁻¹ for CO₂ and CH₄, respectively (Figure S2.5). These fluxes are in the range of previously reported values in lakes and reservoirs (Adams, 2005; Algesten et al., 2005; Gruca-Rokosz et Tomaszek, 2015 ; Huttunen et al., 2006). Sediment GHG fluxes were not different in the branches versus the main basin for both CO_2 (mean of 2.2 vs 2.4 mmol.m⁻².d⁻¹) and CH₄ (mean of 0.17 vs 0.48 mmol.m⁻².d⁻¹) (Figure S2.5). Applying the measured averages to the area of epilimnetic sediments in each section yields estimates of sediment inputs to the epilimnion of 0.6 (\pm 0.03) and 0.5 (\pm 0.11) mmol.m⁻ 2 .d⁻¹ for CO₂, and 0.04 (± 0.02) and 0.10 (± 0.06) mmol.m⁻².d⁻¹ for CH₄ in the branches and main basin respectively (Figure 2.6 and Table S2.2 and S2.3). These inputs from littoral sediments likely represent an upper limit since they are based on deep pelagic sediment cores (littoral area were too compact for coring), where a higher organic matter accumulation and degradation is expected in the system (Blais et Kalff, 1995; Soued et Prairie, 2020). Even as upper estimates, the calculated rates of sediment GHG inputs remain a relatively modest fraction of the average emissions to the atmosphere for the branches and main basin both for CO₂ (13 % and 7 %, respectively) and CH₄ (4 % and 23 %, respectively) (Tables S2 and S3).

2.4.6 Metabolism

*CO*₂ *metabolism*. Estimates of GPP and ER based on diel O₂ monitoring collectively ranged from 3.6 to 34.5 µmol.L⁻¹.d⁻¹ and from 5.8 to 29.5 µmol.L⁻¹.d⁻¹, respectively (Figure 2.4), which is well in range with reported rates for oligotrophic systems (Bogard et del Giorgio, 2016 ; Hanson *et al.*, 2003 ; Solomon *et al.*, 2013). As expected, rates of GPP and ER were correlated ($R^2_{adj} = 0.23$, p-value < 0.001, Figure 2.4), with photosynthetic production stimulating respiration of the produced organic matter. In most cases, GPP exceeded ER, and this was most pronounced in the branches and near aquaculture sites (Figure 2.4), where higher nutrients (TP and TN) and Chla concentrations were measured (Table 2.1).

In the reservoir branches, results from the diel O_2 monitoring method suggested systematic net CO_2 uptake ranging from -19.2 to -1.4 µmol.L⁻¹.d⁻¹, whereas results from two incubations were slightly above that range (-0.5 to 3.3 µmol.L⁻¹.d⁻¹) (Figure 2.4). In the main basin, incubation results ranged from -8.8 to 7.2 µmol L⁻¹ d⁻¹, while the diel O_2 technique captured a wider variability in net CO_2 metabolic rates from -19.2 to 6.1 µmol L⁻¹ d⁻¹, with an estimated CO_2 uptake in 39 out of 54 cases (Figure 4.2).

Areal net CO_2 metabolic rates, as the average of the two methods, yielded an ecosystem-scale estimate of -23.2 and -11.8 mmol.m⁻².d⁻¹ in the reservoir branches and main basin respectively (Table S2.2).

 CH_4 metabolism. Net metabolic CH₄ rates (obtained from incubations) ranged from -0.026 to 0.078 µmol.L⁻¹.d⁻¹, indicating that the CH₄ balance in the epilimnion of Batang Ai varied from net oxidation to net production (Table S2.3). CH₄ metabolic rates measured in Batang Ai are within the range of values observed in other systems for oxidation (Guérin et Abril, 2007 ; Thottathil *et al.*, 2019) and production (Bogard *et al.*, 2014 ; Donis *et al.*, 2017). No significant difference were detected between net metabolic CH₄ rates in the branches versus the main basin due to the high variability and limited data points. Temporal patterns could not be assessed since the majority of the incubations were performed during the same sampling campaign (Nov-Dec 2016).

2.4.7 Ecosystem scale GHG budgets

Collectively, the fluxes estimated above can be summed to explore whether the epilimnetic GHG budgets can be closed within the uncertainty limits of each individual component. Concurrently, it can be used to identify the main processes sustaining GHG emissions in the different sections of the reservoir surface. Figure 2.3 depicts such reconstruction of the epilimnetic GHG budget of Batang Ai. While each process varied in time, their relative importance in driving surface fluxes was generally similar from one sampling campaign to another (Table S2.2 and S2.3).

CO₂ budget. For CO₂, epilimnetic sediment inputs had the smallest contribution, being typically an order of magnitude lower than measured surface fluxes in both sections of the reservoir (Table S2.2). Vertical CO₂ inputs from lower depths on the other hand contributed substantially to surface fluxes in both the branches and the main basin (mean of 0.7 and 4.5 mmol.m⁻².d⁻¹ respectively, Table S2.2), indicating that hypolimnetic processes impact surface emissions despite the permanent stratification. Horizontal inputs of CO_2 were in the same range as vertical inputs (mean of 4.3) mmol.m⁻².d⁻¹) in the branches, however, they decreased by an order of magnitude when reaching the main basin (mean of 0.3 mmol.m^{-2} .d⁻¹). Thus, direct CO₂ inputs from the inflows notably increase surface flux rates in the reservoir branches but only minimally in the main basin. Net CO₂ metabolism was surprisingly variable (switching from negative to positive NEP on a daily time scale), thus making it difficult to derive a sufficiently precise ecosystem-scale estimate to close the epilimnetic budget (Figure 2.3), despite high sampling resolution (n = 66 daily metabolic rates). Including the metabolism substantially shifts the mean of the CO_2 epilimnetic budget (sum of sources and sinks) to a negative value and drastically increases its uncertainty (Figure 2.3 and Table S2.2), reflecting a potentially important but poorly resolved role of metabolism in the budget because of its variability. However, given that metabolism acts more likely as a CO_2 sink on average, our best assessment suggests that, vertical transport from deeper layers is the main source sustaining surface CO_2 out-flux in the main basin of Batang Ai.

 CH_4 budget. In contrast with CO₂, vertical transport was the smallest source of CH₄ to the epilimnion, contributing to less than 2 % to surface fluxes in both reservoir sections (Table S2.3). In the branches, sediment inputs and net CH₄ metabolic rates were both relatively low (mean of 0.04 ± 0.02 and 0.04 ± 0.05 mmol.m⁻².d⁻¹) and had little impact on the budget, corresponding each to 4 % of surface fluxes in that section (Table S2.3). On the other hand, horizontal inputs was the dominant and most variable source sustaining CH₄ emissions in the branches, where the epilimnetic mass balance closed almost perfectly (Figure 2.6 and Table S2.3). Despite being the main CH₄ source in the branches, horizontal transport was a negligible component in the main basin (< 1 % of the flux, Table S2.3). Instead, sediment inputs played a larger role in that section, with a mean of 0.10 (\pm 0.06) mmol.m⁻².d⁻¹, fueling 23 % of surface emissions in the main basin. As with CO_2 , the most variable CH_4 component of the mass balance in the main basin was the net metabolism within the epilimnion (mean of -0.16 ± 0.19 mmol.m⁻².d⁻ ¹). Considering all sources, the CH₄ budget indicates a deficit of 0.34 mmol.m⁻².d⁻¹ in order to explain measured surface emissions in the main basin (Figure 2.3 and Table S2.3).



Figure 2.3: Density distributions of the different components of CO_2 and CH_4 surface budget in the reservoir branches and main basin (H = horizontal flow inputs, S = sediment inputs, V = vertical inputs, M = net metabolism, T = sum of all estimated sources and processes in the surface layer, and F = measured surface fluxes). Density curves are based on simulated normal distributions using the mean and standard error of each component.

2.5 Discussion

Our results have highlighted both the importance and the challenges associated with quantifying simultaneously all the components of the epilimnetic CO_2 and CH_4 budgets, particularly in a hydrologically complex reservoir system. While mass fluxes (hydrological, sedimentary and air-water fluxes) are relatively easy to constrain, internal C processing, namely the net metabolic balances between production and consumption of CO_2 and CH_4 are highly dynamic in both time and space, leading to

significant uncertainties when extrapolated to the scale of whole ecosystem compartments (in our case, the epilimnion). In many studies, some components are only inferred by difference. While convenient from a mass-balance perspective, we argue that assessing all components together is necessary to clearly identify knowledge gaps as well as sources of uncertainty.

2.5.1 Spatial dynamics of CO₂ and CH₄

The decrease in gas concentration and air-water fluxes along the hydrological continuum observed across sampling campaigns and for both CH₄ and CO₂ reflects a robust spatial structure of the gases. Concurrently, estimates of the horizontal GHG inputs shows a clear and consistent spatial pattern, being high in the branches but negligible in the main basin. A temporal effect of riverine inputs was also observed as the two sampling campaigns with the highest horizontal CH₄ inputs coincided with the highest CH₄ emissions in the branches (Table S2.3). All these results concord with the progressively reduced influence of direct GHG catchment inputs and greater preponderance of internal processes along the hydrological flow (Hotchkiss *et al.*, 2015).

For CO₂, the sharpest change in surface metrics (concentration, flux, and isotopic signature) was observed between the inflows and the reservoir branches (Figure 2.2). Despite large riverine inputs (Table S2.2), the branches exhibited low CO₂ concentration and fluxes, as well as an increase in δ^{13} CO₂ matching with high GPP values (Figure 2.2 and 2.4). This may reflect increased light availability for phytoplankton when transitioning from the turbid inflows to the reservoir branches (higher Secchi depth, Table 2.1), a pattern previously reported in other reservoirs (Kimmel et Groeger, 1984; Pacheco *et al.*, 2015; Thornton *et al.*, 1990). While the branch areas are often associated with high CO₂ out-flux due to riverine inputs (Beaulieu *et al.*, 2016; Paranaíba *et al.*, 2018; Pasche *et al.*, 2019; Roland *et al.*, 2010; Rudorff *et al.*, 2011), they are occasionally observed to have low air-water flux due to

simultaneous nutrient inputs (Paranaíba *et al.*, 2018 ; Wilkinson *et al.*, 2016). In Batang Ai, inflows have a high nutrients (TP and TN) to DOC ratio compared to the reservoir branches (Table 2.1), providing higher inputs of nutrients relative to organic matter, and thus likely stimulating primary production more than respiration. This hypothesis is consistent with a higher GPP: ER ratio and mean Chla concentrations measured in the branches compared to the main basin (Figure 2.4 and Table 2.1). The fact that CO_2 concentration within the reservoir (branches and main basin) was negatively correlated to temperature (Figure S2.1), likely due to its effect on GPP (Bogard *et al.*, 2020), highlights the important role of primary production in modulating CO_2 dynamics throughout the reservoir, and particularly in the branches.

The correlation between surface CH₄ surface and TN in the reservoir suggests that primary production may also affect CH₄ dynamics. Nutrient content was shown in previous studies to enhance CH₄ production in the sediments (Beaulieu *et al.*, 2019 ; Gebert *et al.*, 2006 ; Isidorova *et al.*, 2019) and in the oxic water column (Bogard *et al.*, 2014), through its link with algal production and decomposition. However, CH₄ concentration and flux variability were strongly driven by a spatial / hydrological structure, gradually decreasing from the inflows to the main basin. This likely reflects the combined effect of terrestrial inputs and a decreasing contact of water with sediments along the water channel. Surface δ^{13} CH₄ signatures varied substantially but without a consistent spatial pattern (Figure 2.2), indicating that the surface CH₄ pool is shaped by multiple sources / processes varying through space and time.

The changing relative contribution of sources and processes shaping surface CO_2 and CH_4 concentrations varies with the system hydro-morphology, from the inflows to the main reservoir basin, and lead to a progressive decoupling between the two gases along the continuum (Figure S2.2). The observed CO_2 and CH_4 coupling in the inflows and branches is associated to a common catchment source, as previously reported in other systems including soil-water (Lupon *et al.*, 2019), streams (Rasilo *et al.*, 2017), and

lake and reservoir inflow areas (Loken *et al.*, 2019; Natchimuthu *et al.*, 2017; Paranaíba *et al.*, 2018). Indeed, horizontal inputs are the main source of both CO₂ and CH₄ in the upstream reaches of Batang Ai, accounting on average for 91 and 92 % of their respective surface out-flux in the branch section (Figure 2.3 and Tables S2.2 and S2.3). However, when reaching the main basin, driving sources diverge between the two gases, with vertical inputs from the bottom layer supporting on average 60 % of CO₂ compared to 2 % of CH₄ fluxes, while sediment inputs sustained 7 versus 23 % of CO₂ and CH₄ fluxes respectively in that section. This decoupling partly results from the two gases having distinct metabolic pathways: mainly aerobic for CO₂ and anaerobic for CH₄, leading to their sources and sinks being spatially disconnected in the main basin. Consequently, sediments being a mostly anaerobic environment are a more important source of CH₄ relative to CO₂, while the metalimnetic layer being oxichypoxic acts as a sink of CH₄ and source of CO₂ via aerobic CH₄ oxidation (Figure S2.4). Overall, the spatial patterns reported here highlight the hydrodynamic zonation common in reservoirs and its diverging effect on CO₂ versus CH₄ cycling.

2.5.2 CO_2 metabolism

Our observation that GPP often exceeded ER (Figure 2.4) was not unexpected given the low DOC concentration (< 1 mg.L⁻¹) throughout the study system. Previous work has reported that DOC > 4 mg.L⁻¹ is required to sustain persistent net heterotrophy and CO₂ evasion (Hanson *et al.*, 2003 ; Prairie *et al.*, 2002). Throughout the reservoir, we found high day-to-day variability in both ER and GPP, but with no apparent link to weather data, namely light and rain (unpublished). The absence of such a link at a daily time scale has been previously reported (Coloso *et al.*, 2011), while other studies associated daily variations in metabolism with changes in water inflows carrying nutrients (Pacheco *et al.*, 2015 ; Staehr et Sand-Jensen, 2007), or thermocline stability regulating hypolimnetic water incursions to the epilimnion. Such variations in thermocline depth are thought to be more common in warm tropical systems (Lewis, 2010), and were observed across sampling campaigns in Batang Ai, especially in the branches where the depth of the mixed layer varied considerably (SD = 2.3 m, Table 2.1). Hence, hydrological and physical factors may regulate spatial and daily patterns of GPP and ER rates in Batang Ai through their influence on nutrient dynamics.

The accuracy of rates derived from diel O_2 monitoring partly depends on the respiratory and photosynthetic quotients (RQ and PQ) assumed for the conversion of metabolic rates from O_2 to CO_2 . A quotient differing from the assumed 1:1 ratio can lead to an apparent under or over-estimation of net CO₂ production. The fact that net CO₂ metabolic rates were on average higher in incubations, based on direct CO_2 measurements, compared to diel O₂ monitoring (Figure 2.4 and Table S2.2), hints towards a deviation of the metabolic quotients form unity in Batang Ai. Additionally, surface O₂ versus CO₂ concentrations shows that the departure of these gases from saturation varies widely around the expected 1:-1 line, with many surface samples oversaturated in both O_2 and CO_2 , especially in the branches (Figure 2.5). This indicates an excess O₂ and / or CO₂ that can be due to a PQ and / or a RQ higher than 1, or to external CO₂ inputs to the epilimnion (Vachon et al., 2020), for instance from the inflows or the bottom layer. While CO₂ horizontal inputs in the branches and vertical inputs in the main basin (Table S2.2) certainly play a role in surface O_2 / CO_2 decoupling, this pattern is also likely shaped by metabolic quotients. Metabolic quotients have been shown to be widely variable, depending on the type and magnitude of photochemical and biological reactions at play (Berggren et al., 2012; Lefèvre et Merlivat, 2012; Vachon et al., 2020; Williams et Robertson, 1991). The lack of direct measure of metabolic quotients in Batang Ai adds uncertainty to the net CO₂ metabolism estimates based on O₂ data. The observed decoupling of O₂ and CO₂ metabolism in Batang Ai highlights the need for a deeper understanding of the biochemical reactions occurring in the epilimnion, and their effect on metabolic quotients.

Overall, our results from Batang Ai reservoir point to water column metabolism as both a key process in the CO₂ epilimnetic budget and a challenging one to estimate at an ecosystem scale (Figure 2.3). Improving this requires a better mechanistic knowledge of the physical and biochemical processes at play and how they interact to shape NEP.



Figure 2.4: Epilimnetic daily GPP versus ER rates (left panel) derived from free water O_2 sensors deployed in the reservoir branches (green points) and main basin (blue points, cross symbol indicates proximity to aquaculture sites), with the 1:1 line (dotted). Boxplots of the corresponding rates of CO_2 NEP (right panel) in the branches and main basin, with boxes bounds, whiskers, solid line, and open circles representing the 25th and 75th percentiles, the 10th and 90th percentiles, the median, and single data points. Grey squares are values from incubations.



Figure 2.5: Surface O_2 versus CO_2 departure from saturation for all sampled surface sites in the reservoir main basin (blue points) and branches (green points) across all sampling campaigns.

2.5.3 CH₄ metabolism

Incubation results exhibited a wide range of net CH₄ metabolism: from net oxidation to net production, with rates within the range of values observed in other systems for both oxidation (Guérin et Abril, 2007 ; Thottathil *et al.*, 2019) and production (Bogard *et al.*, 2014 ; Donis *et al.*, 2017). CH₄ oxidation is known to be highly dependent on CH₄ availability and is optimal in low oxygen and low light conditions (Borrel *et al.*, 2011 ; Thottathil *et al.*, 2018, 2019), whereas CH₄ production in the oxic water is still poorly understood but have been frequently linked to phytoplankton growth (Berg *et al.*, 2014 ; Bogard *et al.*, 2014 ; Lenhart *et al.*, 2015 ; Wang *et al.*, 2017). A large variability in results exists among the studies that have assessed the net balance of CH₄ metabolism in the water column, with some studies reporting pelagic CH₄ production

as a largely dominant process (Donis et al., 2017) while others find no trace of it (Bastviken et al., 2008). Based on spatial patterns of surface CH₄ concentration and isotopic signature with distance to shore, DelSontro et al. (2018b) showed that, in 30 % of their studied temperate lakes, CH₄ oxidation was dominant versus 70 % dominated by net pelagic production. In Batang Ai, surface δ^{13} CH₄ values were highly variable (-82.5 to -47.7 ‰) but mostly uncorrelated with distance to shore, except a positive correlation indicative of oxidation in the Nov-Dec 2016 ($R^{2}_{adj} = 0.29$, p-value = 0.01) coinciding with a strong inverse pattern for CH₄ concentration ($R^{2}_{adi} = 0.54$, p-value < 0.001, Figure 2.6). This suggests a temporal shift in processes driving surface CH₄ patterns. Also, some measured surface δ^{13} CH₄ values were lower than the mean δ^{13} CH₄ form the sediments (-66.0 ‰, unpublished data), suggesting another highly depleted source of pelagic CH₄ in the system. This is in line with water incubation results often showing positive net CH_4 production (Table S2.3). When reported as mean areal rates, CH₄ metabolism ranged from net consumption to net production of CH₄ (-0.29 to 0.94 mmol.m⁻².d⁻¹), which reflects its potential in having a high impact, either positive or negative, on the epilimnetic CH₄ budget at the reservoir scale (Figure 2.3 and Table S2.3). Results in Batang Ai show that the net balance of CH₄ metabolic processes varies widely even within a single system. However, the factors regulating this balance remain largely unknown. Investigating such factors constitute a key step in resolving CH₄ budgets in lakes and reservoirs.



Figure 2.6: Regression of CH₄ concentration (left panel) and isotopic signature (right panel) as a function of distance to shore in each sampling campaign in the main reservoir basin. For CH₄ concentration, regressions lines have the following statistics in order of sampling: p-values: < 0.001, 0.06, 0.03, 0.05, and R²_{adj}: 0.54, 0.13, 0.11. For δ^{13} CH₄, all regressions had p-values > 0.2 except for the Nov-Dec 2016 campaign with p-value = 0.01 and R²_{adj} = 0.29.

2.5.4 Epilimnetic GHG budgets

For CO₂, measured surface fluxes in both reservoir sections fall in the range of possible values estimated by the sum of epilimnetic processes and their uncertainties (Figure 2.6 and Table S2.2). However, the averages of those two terms differ substantially, due to negative values of metabolism shifting the mean of the mass balance towards net CO_2 consumption whereas, on average, surface out-flux was measured from the reservoir. This discrepancy indicates either a missing source of CO_2 in the budget or the underestimation of one of the processes. While lateral groundwater input is a

potential source not explicitly considered, it is probably modest given the small ratio of littoral area to epilimnion volume, and is unlikely to account for the large CO₂ deficit in the budget. On the other hand, underestimation of the CO₂ metabolic balance is much more likely, given its large variability and uncertainty around its mean value. Additionally, a systematic underestimation of the CO₂ metabolic rates derived from the diel O₂ method is very possible in Batang Ai given the likely deviation of metabolic quotients around the 1:1 line. As an example, when setting the photosynthetic quotient to 1.2 instead of 1, which remains well within the literature range (Lefèvre et Merlivat, 2012 ; Williams et Robertson, 1991), the average epilimnetic CO₂ mass balance would increase from -17.7 to 4.3 mmol.m⁻².d⁻¹ in the branches and from -6.5 to 6.2 mmol.m⁻².d⁻¹ in the respective sections. Thus, constraining the metabolic component, especially the O₂: CO₂ quotients, is key for closing the CO₂ epilimnetic budget.

In the case of CH₄, the epilimnetic mass balance in the branches is surprisingly close to the observed surface flux, largely fueled by horizontal inputs. Hence, CH₄ emissions from the branches reflect catchment CH₄ loads rather than internal processes. However, in the main basin, these inputs become negligible and the estimated budget does not match measured emissions, indicating a deficit of 0.49 mmolCH₄.m⁻².d⁻¹. This amount cannot be explained by a potential underestimation of horizontal or vertical inputs since they are two orders of magnitude lower. Similarly, sediment inputs would need to be six time higher than estimated to fulfill the budget deficit, which is unlikely given their much lower range of uncertainty. Thus, the most plausible source to close the mass balance in the main basin would be water column CH₄ production. Although the estimated CH₄ metabolism indicates an average net consumption rather than a net production (-0.16 mmol.m⁻².d⁻¹), this mean value is based on only 3 data points and has a high uncertainty associated to it (SE = 0.19 mmol.m⁻².d⁻¹, Table S2.3). Closing the mass balance would require a net volumetric CH₄ production of about 0.03 µmol.L⁻¹.d⁻¹ in the water column of the main basin. This value seems plausible since an equal

production rate was measured in one of the incubations, and it is at the low end of the range reported in other systems (Bogard *et al.*, 2014 ; DelSontro *et al.*, 2018b ; Donis *et al.*, 2017). The combination of our results point to water column metabolism as the main source of CH₄ in the main basin of Batang Ai, and could be sustaining up to 75 % of surface emissions in that section. However, this process seems highly dynamic and requires more intensive research into its controls at spatial and temporal scales, commensurate with CH₄ emissions.

2.6 Conclusion

The estimated epilimnetic CO_2 and CH_4 budgets in Batang Ai has helped define the role of different processes in shaping the reservoir surface GHG fluxes to the atmosphere. Results showed that horizontal riverine inputs are important sources of GHG in the reservoir branches (especially for CH_4). This creates a coupling between CO₂ and CH₄ close to the river deltas, which gradually fades along the water flow, until the surface concentrations of the two gases become completely uncoupled in the main basin being driven by different sources. For instance, vertical inputs from the bottom layer contributed significantly to surface CO₂ saturation, while being negligible in the case of CH₄ due to metalimnetic oxidation. Inversely, sediment inputs played a notably greater role in sustaining epilimnetic oversaturation of CH_4 compared to CO_2 in the main basin. Nonetheless, the epilimnetic budgets of both gases are highly sensitive to their respective water column metabolism. This result is likely representative of large systems with a high volume of water versus sediments, which is common for hydroelectric reservoirs. The metabolic balances of CO_2 and CH_4 were also extremely variable in space and time, switching from a net production to a net consumption of the gases, and leading to highly uncertain ecosystem-scale estimates, which emphasizes the key but unconstrained role of metabolism in the overall GHG budgets. Factors driving these metabolic changes are not well constrained based on current knowledge, highlighting the need for further research on the subject. Overall, this study gives an integrative portrait of the relative contribution of different sources to surface CO_2 and CH_4 fluxes in a permanently stratified reservoir including its transition zones (branches). Conclusions and insights derived from this work likely reflect C dynamics in other similar systems, and offer useful tools to guide future research.

2.7 Acknowledgments

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2.8 Supplementary information

	[CO ₂]		[CH4]		
Variables	cor	p-value	cor	p-value	n
Water temperature	-0.36	< 0.001	0.07	0.223	131
% O ₂	-0.01	0.933	0.15	0.013	130
Chla	0.12	0.056	0.18	0.005	110
DOC	0.01	0.939	0.17	0.008	113
TP	0.14	0.031	0.13	0.049	112
TN	0.1	0.112	0.27	< 0.001	113

Table S2.1: Non-parametric Kendall correlation (cor) and their p-value between reservoir surface CO₂ and CH₄ concentrations with physical and chemical variables.



Figure S2.1: Linear regressions of natural logarithm of surface CO₂ and CH₄ concentrations with water temperature (left panel, p-value < 0.001, $R^2_{adj} = 0.22$) and TN (right panel, p-value < 0.001, $R^2_{adj} = 0.14$) respectively in the reservoir (branches and main basin).



Figure S2.2: Linear regressions of natural logarithm of surface concentrations of CO₂ versus CH₄ in the reservoir inflows (brown, p-value = 0.006 and $R^2_{adj} = 0.54$), branches (green, p-value = 0.005 and $R^2_{adj} = .0.17$), and main basin (blue, p-value = 0.11 and $R^2_{adj} = 0.01$).



Figure S2.3: Boxplots of the buoyancy frequency (N^2 , bottom left panel), dissipation rate (ϵ , bottom right panel), and CH₄ and CO₂ concentration gradients at the epilimnion-metalimnion interface (top panels) at epilimnion-metalimnion interface in the reservoir branches and main basin. Boxes bounds, whiskers, solid lines, and open circles represent the 25th and 75th percentiles, the 10th and 90th percentiles, the median, and single data points respectively.



Figure S2.4: Example of a vertical profile (main reservoir, Aug 2018) of CO₂ (light gray points) and CH₄ (dark gray points) concentrations, as well as δ^{13} CH₄ (dark grey triangles) as a function of depth. The dotted line represents the top of the metalimnion.



Figure S2.5: Boxplot of sediment CO_2 and CH_4 flux rates (per sediment area) across sampled sites in the reservoir branches and main basin. Boxes bounds, whiskers, solid lines, and open circles represent the 25th and 75th percentiles, the 10th and 90th percentiles, the median, and single data points respectively.

area. *Represents missing value, assumed to be equal to the mean of other campaigns. н S V Т F M_inc M mod Nov-Dec 2016 1.4 (± 1.9) 5.7 (± 2.6) 0.6 (± 0)* -1.8 (± NA) -63.2 (± NA) -26.5 (± 36.1) 2.8 Apr-May 2017 15.2 2.7 (± 0.2) 0.6 (± 0)* 3.6 (± 2.3) 1.4 (± 1.9)* -32.4 (± 20) -8.6 (± 19.4) Branches Feb-Mar 2018 3.2 (± 1.1) 0.6 (± 0)* 0.9 (± 0.5) 1.4 (± 1.9)* -47.8 (± 20)* -18.6 (± 26.2) 4.1 Aug 2018 5.5 (± 3.3) 0.6 (± 0)* 0.1 (± 0.5) 1.4 (± 1.9)* -47.8 (± 20)* -17.1 (± 28.4) -3.4 Mean 4.3 (± 0.8) $0.6(\pm 0)$ 0.7 (± 1.1) 1.4 (± 1.9) -47.8 (± 15.4) -17.7 (± 26.5) 4.7 (± 3.9) Nov-Dec 2016 0.2 (± NA) 0.5 (± 0.1)* 6 (± 2.2) -2.6 (± 6.2) -21.4 (± 5.6) -5.3 (± 11.7) 11.3 Main basin Apr-May 2017 0.4 (± NA) 0.5 (± 0.1)* 3.1 (± 1.3) 7.2 (± NA) -49 (± 37.3) -16.9 (± 29.5) 15.1 Feb-Mar 2018 0.5 (± 0.1)* 3.3 0.5 (± NA) 4.7 (± 1) 2.3 (± 6.2)* -17 (± 4.5) -1.6 (± 10.8) Aug 2018 0.1 (± NA) 0.5 (± 0.1)* 4.3 (± 3.3) 2.3 (± 6.2)* -16.2 (± 20.9) -2 (± 12.6) 0.3 Mean $0.3(\pm 0.1)$ 0.5 (± 0.1) 4.5 (± 0.6) $2.3(\pm 4.9)$ -25.9 (± 7.8) -6.5 (± 14.9) 7.5 (± 3.4)

Table S2.2: Estimated averages (± SE) of rates of CO₂ inputs to the epilimnion from horizontal flow (H), sediments (S), vertical transport (V), and internal metabolism (M_inc and M_mod when derived from incubations and inverse O₂ modeling respectively), their sum (T, considering an average of M_inc and M_mdod) and surface flux to the atmosphere (F) in the branches and main basin for each sampling campaign and their mean. Units are mmol.m⁻².d⁻¹ of water surface area. *Represents missing value, assumed to be equal to the mean of other campaigns.

Table S2.3: Estimated averages (\pm SE) of rates of CH₄ inputs to the epilimnion from horizontal flow (H), sediments (S), vertical transport (V), and internal metabolism (M_inc), their sum (T) and surface flux to the atmosphere (F) in the branches and main basin for each sampling campaign and their mean. Units are mmol.m⁻².d⁻¹ of water surface area. *Represents missing value, assumed to be equal to the mean of other campaigns.

		Н	S	V	M_inc	Т	F
Branches	Nov-Dec 2016	0.94 (± 0.67)	0.04 (± 0.02)*	-0.001 (± NA)	0.04 (± 0.05)	1.03 (± 0.74)	1.28
	Apr-May 2017	0.58 (± 0.08)	0.04 (± 0.02)*	0.001 (± 0.008)	0.04 (± 0.05)*	0.67 (± 0.16)	0.71
	Feb-Mar 2018	0.16 (± 0.08)	0.04 (± 0.02)*	0.025 (± 0.016)	0.04 (± 0.05)*	0.27 (± 0.17)	0.84
	Aug 2018	1.97 (± 1.39)	0.04 (± 0.02)*	0.034 (± 0.029)	0.04 (± 0.05)*	2.09 (± 1.49)	1.13
	Mean	0.91 (± 0.39)	0.04 (± 0.02)	0.015 (± 0.009)	0.04 (± 0.05)	1.01 (± 0.46)	0.99 (± 0.13)
Main basin	Nov-Dec 2016	0.005 (± NA)	0.1 (± 0.06)*	0.001 (± 0.004)	-0.35 (± 0.32)	-0.24 (± 0.38)	0.36
	Apr-May 2017	0.003 (± NA)	0.1 (± 0.06)*	0.013 (± 0.008)	0.03 (± NA)	0.14 (± 0.39)	0.39
	Feb-Mar 2018	0.003 (± NA)	0.1 (± 0.06)*	0.001 (± 0.003)	-0.16 (± 0.32)*	-0.06 (± 0.38)	0.45
	Aug 2018	0.003 (± NA)	0.1 (± 0.06)*	0.018 (± 0.008)	-0.16 (± 0.32)*	-0.04 (± 0.39)	0.58
	Mean	0.004 (± 0.001)	0.1 (± 0.06)	0.008 (± 0.004)	-0.16 (± 0.19)	-0.05 (± 0.25)	0.44 (± 0.05)

CHAPITRE III

CO₂ AND CH₄ DYNAMICS AND ITS DETERMINANTS IN THE DEEP LAYER OF A TROPICAL RESERVOIR

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N.B. References cited in this chapter are presented at the end of the thesis.

3.1 Abstract

Deep layers of lakes and reservoirs are active sites of carbon dioxide (CO_2) and methane (CH_4) production and accumulation. These two potent greenhouse gases can be emitted to the atmosphere during water column mixing and, for hydropower reservoirs, from deep water discharge to the downstream river. Regulation of deep CO₂ and CH₄ concentrations depends on multiple factors including rates of sediment-water exchange, aerobic respiration, methanogenesis, CH₄ oxidation, and vertical diffusion. These drivers are interlinked but have mostly been studied separately, making their relative importance and interactions in a system difficult to disentangle. In this study, we use empirical observations combined with an integrative modeling approach to build a comprehensive picture of CO_2 and CH_4 dynamics in the deep layer of a permanently stratified reservoir located on the tropical Borneo Island (Malaysia). Results show that O2 acts as a key modulator of deep CO₂ and CH₄ metabolism, and largely shapes their spatial and temporal patterns at the ecosystem scale. The availability of O_2 in the deep layer depended on vertical diffusion, which also controlled CO_2 and CH_4 movement. Our model reproduced well the shape of vertical profiles of concentration and isotopic signature, and highlighted the central role of gas diffusivity in hypolimnetic GHG dynamics through its direct effect on gas transport and its interaction with water column metabolism. Furthermore, it suggests that the GHG stocks are now in a steady-state between production and removal processes. The predictive framework developed here could be applied to better understand and predict hypolimnetic GHG storage and emissions in lakes and reservoirs.

3.2 Introduction

Inland waters are key elements of the carbon (C) cycle, emitting large amounts of greenhouse gases (GHG) to the atmosphere in the form of CO_2 and CH_4 . These emissions are sustained by external inputs and internal processing of C in aquatic systems. While emissions occur at the airwater interface, a substantial portion of the C mineralization occurs below the thermocline, in the

water column and sediments of the hypolimnion, where CO₂ and CH₄ tend to accumulate. The importance of hypolimnetic GHG storage is evident during lakes mixing, like at fall overturn, known to induce large peaks in GHG out-flux (Bastviken et al., 2004; Ducharme-Riel et al., 2015; Encinas Fernández et al., 2014). Deep GHG storage is particularly important in permanently stratified systems (usually in the tropics), where rare mixing events can trigger alarming amounts of gas release (Schmid et al., 2005, 2006). In natural lakes, a large fraction of the CH₄ produced is consumed by methanotrophic bacteria before reaching the atmosphere. In reservoirs however, deep water withdrawal often leads to a direct release of hypolimnetic GHG to the atmosphere through degassing upon discharge to the outflow river and subsequent diffusive emissions (Demarty et Bastien, 2011; Guérin et al., 2006; Kemenes et al., 2016). The contribution of degassing and downstream emissions to the overall reservoir C footprint is highly variable, ranging from near negligible to dominant (Galy-Lacaux et al., 1997; Kemenes et al., 2007; Soued et Prairie, 2020; Teodoru *et al.*, 2012). What distinguishes such divergent behaviors resides in the interplay between the physical and biogeochemical processes regulating the dynamics of CO₂ and CH₄ and their interaction in the hypolimnion of these systems. Therefore, predicting reservoir degassing and downstream GHG emissions is important but challenging, being constrained by the current knowledge on hypolimnetic GHG dynamics.

Dissolved oxygen (O_2) has been shown to play a key role in both CO_2 and CH_4 dynamics. Indeed, the production of CO_2 derives mostly from aerobic respiration which depends on O_2 availability as an electron acceptor. Dissolved O_2 also inhibits CH_4 production and strongly modulates its oxidation (Conrad, 2009 ; Thottathil *et al.*, 2019). The vertical penetration of O_2 from the welloxygenated surface to the deep layers of aquatic systems is controlled by physical factors, namely gas diffusivity, which also controls both CO_2 and CH_4 transport throughout the water column (Barrette et Laprise, 2005 ; Kreling *et al.*, 2014 ; Pu *et al.*, 2020). The interplay between these physical processes and production and consumption rates of CO_2 and CH_4 in the water column and the sediments shapes the vertical profiles of the two gases and their temporal evolution.

The multiple processes regulating CO_2 and CH_4 dynamics in deep waters have often been examined separately, with studies focusing exclusively on aerobic respiration (Karlsson *et al.*,

2007 ; Solomon *et al.*, 2013), CH₄ oxidation (Bastviken *et al.*, 2002 ; Thottathil *et al.*, 2019), sediment C gas production (Cardoso *et al.*, 2013 ; Gruca-Rokosz et Tomaszek, 2015), or physical processes (Guérin *et al.*, 2016 ; Rimmer *et al.*, 2006). These studies have greatly improved our understanding of the magnitude and regulation of individual rates of processes involved in aquatic C cycling. However, integrative studies building a comprehensive picture of hypolimnetic CO₂ and CH₄ dynamics are rare and essentially limited to temperate lakes (Bastviken *et al.*, 2008 ; Vachon *et al.*, 2019). Such approaches yielded estimates of the relative proportion of CH₄ produced versus oxidized and emitted (Bastviken *et al.*, 2008), and also revealed the effect of internal gas diffusivity on this balance (Vachon *et al.*, 2019). However, the current understanding of C processes and how they interact at an ecosystem scale to shape deep C gas profiles is still limited, especially in tropical regions.

In this study, we investigate the hypolimnetic CO₂ and CH₄ dynamics in Batang Ai hydroelectric reservoir, located on the tropical Borneo Island (Malaysia). Deep GHG dynamics are particularly important in this system because its permanent stratification traps large quantities of CO_2 and CH_4 in the hypolimnion, and deep water withdrawal discharges this GHG rich waters directly downstream of the dam accounting for 89 % of the reservoir C footprint (Soued et Prairie, 2020). The research reported here has three main objectives. The first is to assess the spatial patterns of CO₂ and CH₄ concentrations in the deep reservoir layer along a hydrological gradient of distance from the inflowing river delta. This aims at understanding the effect of a changing flow and morphometry (common in reservoirs) on the spatial behaviour of deep CO_2 and CH_4 . The second is to re-construct, through modeling, the temporal evolution of CO₂ and CH₄ accumulation in the deep layer since the construction of the dam until 35 years later. This is useful to estimate past reservoir emissions and to determine whether deep reservoir gas storage is still increasing or has reached a steady state. The third objective is to combine field measurements of C processes into a mechanistic model of hypolimnetic O₂, CO₂ and CH₄ concentrations along a vertical profile. This framework was used as a tool to test the relative impact of different factors on deep GHG vertical profiles. Addressing these objectives jointly leads to a comprehensive understanding of deep GHG dynamics in the study system, and provides useful general knowledge transferable to other lakes and reservoirs.

3.3 Materials and methods

3.3.1 Site and sampling description

Batang Ai is a hydroelectric reservoir impounded in 1985 on the Borneo Island (Sarawak, Malaysia). The region experiences a tropical equatorial climate with annual rainfall between 3300 and 4600 mm, being more abundant from October to March (Sarawak Government, 2019; Weather Spark, s. d.). Average monthly temperature varies very little throughout the year (≤ 2 °C), changing mostly on a daily scale (day and night means of 32 vs 23 °C respectively) (Sarawak Government, 2019; Weather Spark, s. d.). The reservoir is located in the Batang Ai National Park (with limited human activities), drains a mostly forested catchment of 1149 km² and has a total area of 68.4 km², a water residence time of 0.9 year, and a maximum and mean depths of 28 and 73 m respectively. The two main reservoir inlets are the Batang Ai and Engkari rivers (max depth ~10 m) that flow into two wider reservoir branches which converge into a main basin (Figure 3.1). Batang Ai is a clear water oligotrophic reservoir with very low mean concentrations of dissolved organic C (< 1 mg.L⁻¹), phosphorus (< 6 µg.L⁻¹), nitrogen (< 0.2 mg.L⁻¹), and chlorophyll a (< 1 µg.L⁻¹) as described in (Soued et Prairie, 2020).

Four sampling campaigns were conducted in Nov-Dec 2016, Apr-May 2017, Feb-Mar 2018, and Aug 2018, with the corresponding weather and flow conditions shown in Figure S3.1. During each campaign, 4 to 9 sites (P1 - P9) were sampled along a hydrological gradient from the Batang Ai river delta to the main basin (Figure 3.1). In each, site vertical water column profiles were performed from 0 up to 32 m (each 0.5 - 3 m), and one water sample was collected at 60 m. At each sampled depth, water temperature and dissolved O₂ were measured using a multi-parameter probe (Yellow Spring Instruments, model 600XLM-M) equipped with depth gauge. The probe was attached to a submersible pump and plastic tubing from which water was collected for gas analyses. For each sampled profile, temperature data were used to calculate the buoyancy frequency (N²), the thermocline depth, and the top and bottom of the metalimnion with the package rLakeAnalyzer v1.11.4.1 (Winslow *et al.*, 2018) in the R software (R Core Team, 2017). In this

study we use the term "deep layer" to refer to the layer below the epilimnion (the metalimnion and hypolimnion).



Figure 3.1: Map of Batang Ai reservoir sections and sampling sites.

3.3.2 Gas concentrations and isotopic signatures

 CO_2 and CH_4 water concentrations and stable isotopic signatures ($\delta^{13}CO_2$ and $\delta^{13}CH_4$) were measured in all vertical profiles. Water was collected in 60 mL gas-tight syringes inside which a 30 mL ambient or C free air was created. The water and gas phases in the syringe were equilibrated by shaking, and then gas was injected in a 12 mL pre-evacuated air-tight glass vial. Sampled gas was analyzed for CO_2 and CH_4 concentrations on a Shimadzu GC-8A Gas chromatograph and for $\delta^{13}CO_2$ and $\delta^{13}CH_4$ on a Cavity Ring Down Spectrometer (CRDS) equipped with a Small Sample Isotopic Module (SSIM, Picarro G2201 -i). Original water gas concentrations and isotopic signatures were back-calculated accounting for temperature (gas solubility), atmospheric pressure, headspace ratio, and gas values (concentrations and stable isotopic signatures) before and after equilibration.

3.3.3 Sediment sampling

Sediment CO_2 and CH_4 concentrations, isotopic signatures, and fluxes to the overlaying water column were measured in 7 sites (P1-5, P7, and P9, Figure 3.1). Sediment cores were collected with a Glew gravity corer equipped with a plastic liner (6 cm diameter) pre-drilled with 1 cm holes at every centimeter up to 20 cm. During sampling, holes were covered with electric tape, and upon recovery the tape was removed and 3 mL tip-less syringes were inserted through the holes to extract sediments at different depths within the core. Extracted sediments were put in 25 mL glass vials, mixed with 6 mL nano-pure water, and sealed with an air-tight butyl rubber stopper crimped with an aluminum cap. Glass vials were pressurized by injecting 40 mL of ambient air using a syringe and needle. The water and sediment content was equilibrated with the gas phase in the vial by shaking, then the gas phase was collected and analyzed for CO_2 and CH_4 and water concentrations and isotopic signatures as described above (section 3.3.2).

Sediment to water fluxes of CO₂ and CH₄ were estimated based on the gradient in gas concentrations at the sediment-water interface (g, in µmol.L⁻¹.m⁻¹). In most cores, there was a clear positive linear slope between gas concentration and depth in the first 5 cm of sediments and overlaying water (p-value < 0.05 and R²_{adj} > 0.5). In these cases, the latter slope was considered as the term g in Eq. (3.1) below. Otherwise, g was calculated as the gas concentration gradient between the mean of the first 3 cm of sediments and the overlaying water. The sediment to water gas flux rate (S_f in mmol.m⁻².d⁻¹) was calculated with following Eq. (3.1):

$$S_f = \frac{g \times d}{p} \tag{3.1}$$

with *d* is the diffusion coefficient set to $1.5 \times 10^{-5} \text{ cm}^2 \text{.s}^{-1}$ (Donis *et al.*, 2017), and *p* the sediment porosity taken as 2 % based on previous results in Batang Ai reservoir (Tan, 2015).

3.3.4 Water column metabolic rates

Metabolic rates (aerobic respiration and net CH₄ oxidation) were measured at 5 sites (P2, P3, P4, P5, P7, Figure 3.1), during different sampling campaigns, and at different depths (3, 12, and 20 m)

in the water column of Batang Ai. Rates were derived from in situ water incubations in 5 L glass jars with an air-tight clamp lid. Jars were filled and overflowed with water from specific depths, then closed, wrapped inside a black light-proof bag, and fixed to a rope at the corresponding sampling depth for a period of 22 to 24h. To monitor O_2 changes inside the incubations, high frequency O_2 and temperature sensors (model miniDOT from Precision Measurement Engineering) were fixed inside the jars. CH₄ concentrations and stable isotopic signatures were measured at the start and end of the incubations with the same technique as described above (section 3.3.2). Net rates of aerobic respiration and CH₄ oxidation were calculated as the difference between initial and final gas concentrations divided by the incubation time.

CH₄ oxidation was also assessed based on δ^{13} CH₄ values measured in vertical profiles at all sampled sites. In this regard, the fraction of CH₄ oxidized (F_{ox}) was calculated with the following Eq. (3.2):

$$F_{ox} = -e^{(\ln(\delta_{13}C_{source} + 1000) - \ln(\delta_{13}C_{resid} + 1000))/(\alpha - 1)} + 1)$$
(3.2)

with $\delta^{13}C_{source}$ the $\delta^{13}CH_4$ value in the top sediment layer at each site, considered as the only CH₄ source to the deep layer, and $\delta^{13}C_{resid}$ is the residual $\delta^{13}CH_4$ measured at each point in the profile.

3.3.5 Modeling framework and parameters

General framework. The concentrations of O₂, CO₂, CH₄ were modeled in the deep layer of the main basin of Batang Ai as a function of vertical depth (z, from 10 to 60 m) and time (from 0 to 100 years after impoundment). δ^{13} CH₄ was also modeled based on the isotopic ratio of ¹²CH₄ and ¹³CH₄ predicted concentrations. The model framework is based on a system of coupled second degree partial differential equations described in Eq. (3.3) to (3.6):

$$\frac{d[O_2]z}{dt} = K_z \frac{\partial^2[O_2]}{\partial z^2} + R_w + R_s$$
(3.3)

$$\frac{d[CO_2]z}{dt} = K_z \frac{\partial^2[CO2]}{\partial z^2} + R_w + R_s + Sa_{CO2}$$
(3.4)

$$\frac{d[{}^{12}CH_4]z}{dt} = K_z \frac{\partial^2 [{}^{12}CH_4]}{\partial z^2} + P_{CH4} - MOX_{sp} [{}^{12}CH_4]$$
(3.5)

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$$\frac{d[{}^{13}CH_4]z}{dt} = K_z \frac{\partial^2[{}^{13}CH_4]}{\partial z^2} + S_{13:12} * P_{CH4} - \frac{MOX_{sp}[{}^{13}CH_4]}{\alpha}$$
(3.6)

where $[O_2]z$, $[CO_2]z$, $[^{12}CH_4]z$, and $[^{13}CH_4]z$ are the respective gas concentrations at depth z. K_z is the vertical water column gas diffusivity. R_w and R_s are the aerobic respiration rates in the water column and sediments, respectively. Sa_{CO2} is the CO₂ inputs to the water column derived from anoxic sediment processes. P_{CH4} is the CH₄ production rate, MOX_{sp} is the specific CH₄ oxidation rate with a fractionation factor of $\alpha = 1.02$ (Coleman *et al.*, 1981), and S_{13:12} = 0.0104 is the concentration ratio of ¹³CH₄ / ¹²CH₄ in the sediments (considered as the only CH₄ source, with a measured mean δ^{13} CH₄ = -66 ‰). The estimation of each parameter is described in the subsequent paragraphs. In brief, the change in concentration of each gas at a given depth depends on the sum of processes consuming or producing the gas and on the movement of the gas along the vertical profile. Note that the model, being one-dimensional, does not consider horizontal advective transport, and thus ignores the effect of riverine inputs and hypolimnetic water withdrawal at the dam, which introduces uncertainty in the modeling framework. The system of equation was solved in the software Matlab R2018a using the function 'pdepe' function to solve initial boundary value problems (Skeel et Berzins, 1990). Initial values were set to the concentrations of each gas corresponding to its atmospheric partial pressure and water solubility at 30.2 °C (the temperature at 10 m). Boundary conditions (Neumann) at the bottom (60 m) were set to 0, while at the top of the deep layer (10 m depth), Dirichlet boundary constant concentration values were set to 220, 21.2, and 0.07 µmol.L⁻¹ for O₂, CO₂, and CH₄ respectively (corresponding to measured means at the base of the mixed layer).

Gas diffusivity. K_z was estimated using two methods. The first one is based on the model of Hondzo and Stefan (1994) using N² and lake surface area (as a proxy for wind induced turbulence) following Eq. (3.7):

$$K_z = 8.17 * 10^{-4} * A_0 * N^{2^{-0.43}} * \frac{^{86400}}{^{1000}}$$
(3.7)

with $A_0 = 58.9 \text{ km}^2$ the surface area of the reservoir main basin. The second method is based on the general equation of Osborn (1980) outlined in Eq. (3.8):

$$K_z = \Gamma \frac{\epsilon}{N^2} \tag{3.8}$$

with $\Gamma = 0.2$ the mixing ratio (Lee *et al.*, 2017; Oakey, 1982), and ϵ is the turbulent dissipation rate (W kg⁻¹). N² and ϵ were both estimated based on empirical measurements of temperature and shear microstructure profiles using a MicroCTD probe (Rockland Scientific). The high frequency probe (512 Hz) was equipped with a temperature and two velocity shear sensors as well as tilt, vibration, and depth sensors. Measurements were done at 3 sites in the main basin (P8, P9, P10, Figure 3.1) during the Aug 2018 sampling campaign. At each site, 10 replicates cast from 2 to 23 m were performed with the profiler launched half the time from top to bottom (downward configuration) and the other half from bottom to top (uprising configuration). Casts were separated by a minimum waiting time of 4 min to avoid measuring shear created by the previous cast. After a quality check, ϵ was calculated for each profile based on Nasmyth shear spectrum (Oakey, 1982), using ODAS v4.3.03 Matlab library (by Rockland Scientific), and values were binned over 1-2 m and averaged among the two shear probes. Deep layer ϵ values did not yield any clear pattern as a function of depth. Thus, the parameter ϵ in the model (Eq. (3.7) and (3.8)) was considered to be a constant of 1.5 x10⁻⁸ m⁻².s⁻³ representing the mean of all field data along the deep layer vertical profile (10 to 23 m). N^2 as a function of depth (Eq. (3.7) and (3.8)) was calculated as described in section 3.3.1 based on a single temperature profile (site P10, Figure 3.1) on which a smooth spline curve was fitted (function 'smooth.spline' in R software). The resulting N² profile was best fitted to a sigmoid curve (Figure 3.S2) with the following Eq. (3.9):

$$N_z^2 = \left(\frac{1}{1 + e^{0.43 \, (z - 4.45)}}\right) \, 0.0025 + 3.8 * 10^{-5} \tag{3.9}$$

Combining Eq. (3.9) to Eq. (3.7) and (3.8) yields two predictive functions of K_z with depth in the deep reservoir layer. The mean of the two functions was used to calculate K_z as a function of depth in the general model framework (Eq. (3.3) to (3.6)).

Aerobic and anaerobic respiration. R_w represents water column O_2 consumption and CO_2 production (assuming a respiratory quotient of 1). R_w was modeled as a Michaelis-Menten function, with the advantage of considering O_2 consumption as a quasi zero-order reaction at high O_2 and approaching a first-order reaction at low O_2 , as described by Eq. (3.10):

$$R_w = r_{wmax} * \frac{[o_2]}{o_{x_{50}} + [o_2]} \tag{3.10}$$

with r_{wmax} the maximum respiration rate, and Ox_{50} the O_2 concentration at which the respiration rate reaches 50 % of r_{wmax} . This form of equation ties R_w to O_2 when available O_2 becomes limiting. In Eq. (3.10), Ox_{50} was set to 50 µmol.L⁻¹, and r_{wmax} to 21.7 µmol.L⁻¹.d⁻¹, which corresponds to the mean of measured respiration rates in *in situ* deep water incubations (described in section 3.3.4) for which O_2 concentration was higher than Ox_{50} . The model described in Eq. (3.10) and measured R_w data are shown in Figure S3.3. Modeling R_s , the aerobic sediment respiration, was done using the same equation structure and replacing r_{wmax} by r_{smax} which is the maximal sediment respiration rate, calculated with Eq. (3.11):

$$r_{smax} = S_{CO2} * \left(\frac{q}{z_{max} - z}\right)$$
 (3.11)

with S_{CO2} the mean sediment to water CO_2 flux measured in the main reservoir basin (as described in section 3.3.3) considered here as the upper limit for aerobic sediment respiration, and q the bathymetric shape index (Ferland *et al.*, 2014; Imboden, 1973) calculated based on the mean and max depths of the main basin ($z_{mean} = 30$ m, and $z_{max} = 73$ m respectively) following Eq. (3.12):

$$q = \left(\frac{z_{max}}{z_{mean}}\right) - 1 \tag{3.12}$$

This term allows the translation of areal sediment flux rates to volumetric sediment CO_2 inputs to a given water depth considering the changing ratio of water volume to sediment area along the vertical profile. Total O_2 consumption is the sum of R_w and R_s at each depth, however, CO_2 production can also occur under anaerobic conditions, mostly in the sediments. Some of the processes producing CO₂ without consuming O₂ include sulfate reduction and acetoclastic methanogenesis (Borrel *et al.*, 2011; Fahrner *et al.*, 2008; Gruca-Rokosz et Tomaszek, 2015; Holmer et Storkholm, 2001), the latter being observed in Batang Ai sediments (section 3.4.2). Anaerobic CO₂ production was included in the model as an additional term: Sa_{CO2} that was considered equal to 15 % of the maximum sediment respiration rate (r_{smax}), which occurs regardless of available O₂ (as opposed to the term R_s).

 CH_4 production and consumption. Deep CH₄ production was assumed to occur exclusively in the sediments with a volumetric rate (P_{CH4}) that varies as a function of depth (water volume to sediment ratio) following Eq. (3.13):

$$P_{CH4} = S_{CH4} * \left(\frac{q}{z_{max} - z}\right)$$
(3.13)

with S_{CH4} the mean of measured areal sediment to water CH₄ flux (described in section 3.3.3). CH₄ consumption by oxidation was calculated by multiplying CH₄ concentration by MOX_{sp}, the specific CH₄ oxidation rate (Eq. (3.5) and (3.6)). MOX_{sp} was modeled as a function O₂ concentration using the model of Thottathil et al. (2019) in Eq. (3.14) (Figure S3.3):

$$MOX_{sp} = e^{\lambda [O_2]} - e^{(\lambda + \beta) [O_2]}$$
(3.14)

with $\lambda = 0.01$ and $\beta = 0.18$ (same values as in Thottathil et al. (2019)). This model was chosen because it is the only model describing MOX inhibition by O₂ and because our data showed strong evidence of MOX dependence on O₂ concentration (Figure S3.7 and section 3.4.4). Measured MOX rates derived from incubations in Batang Ai could not be used to parametrize the model due to the low number and large variability of data (Figure S3.7).

3.3.6 Model evaluation and sensitivity analyses

Model results were compared to measured profiles at site P6 (Figure 3.1) during the four sampling campaigns (to cover temporal variability). For each modeled component (O₂, CO₂, CH₄ concentrations, and δ^{13} CH₄), we calculated the R² and the Nash-Sutcliffe efficiency (NSE) as a

goodness of fit indexes using the R package hydroGOF (Zambrano-Bigiarini, 2020). We performed sensitivity analyses to assess the relative impact of K_z , S_{CO2} , S_{CH4} , R_w , and MOX on model outputs by dividing and multiplying each of these parameters by a factor of 2. The effect of each parameter was reflected by the extent of change in model results it induced. In particular, changes in two components of the model output were examined: the change in the integrated mean CO_2 and CH_4 concentration in the deep layer, and the change in CO_2 and CH_4 concentrations at the withdrawal depth (21 m), which fuel degassing and downstream emissions.

3.4 Results and discussion

3.4.1 Physical water structure

Water temperature in the reservoir varied between 24 and 33 °C, being the coldest in bottom waters and close to the inflow. As the distance from the river delta increases, thermocline deepened from 1.3 (\pm 1.6) m at the river delta, to 8 (\pm 2.3) m in the reservoir branch, and to 10.6 (\pm 1.7) m in the main basin (Figure 3.2). The maximum buoyancy frequency (N²), observed in the metalimnion, was sufficiently large to create a strong density barrier between the epilimnion and the hypolimnion, albeit weaker in the reservoir branches compared to the main basin (Figure 3.2). This suggests a higher potential for exchanges between surface and deep waters in the branches.

Temporal variations in water temperature and stratification were mostly apparent in upstream reaches, especially at the river mouth. In Aug 2018, surface water temperatures were higher (Figure 3.2) and riverine flow rate was the lowest among sampling campaigns (49 vs 108 - 110 $m^3.s^{-1}$, Figure S3.1). This resulted in a deeper thermocline in the river delta in Aug 2018 with warm water flowing into the surface layer of the reservoir. In contrast, during the Apr-May 2017 and Feb-Mar 2018 campaigns, riverine inflows were colder and denser, thus likely plunged into the reservoir's deep layer introducing significant amounts of O₂ in the hypolimnion compared to the Aug 2018 campaign (Figure 3.2).



Figure 3.2: Cross-section plot of temperature (left), buoyancy frequency N^2 (middle), and dissolved O_2 concentration (right) as a function of the distance from the river delta in each sampling campaign. Interpolated data are based on measured vertical profiles (black dots). Black lines represent the thermocline (solid) and the top and bottom of the metalimnion (dashed)

3.4.2 Sediment GHG patterns

Measured CO₂ and CH₄ concentrations in the sediments ranged from 74 to 8750 μ mol.L⁻¹ and 0.63 to 6800 μ mol.L⁻¹ respectively, with generally increasing concentrations with sediment depth (Figure S3.4). Concentrations of CH₄ and CO₂ in the deep sediment layer (8-10 cm) were highest in the river delta, declined gradually along the reservoir branch and increased in the main basin, with a notably high value near the fish aquaculture area (Site P7, Figure 3.1 and S4). This trend could be due to higher deep sediment organic matter content in the river delta and aquaculture site promoting CO₂ and CH₄ production there. This hypothesis is in line with previous studies pointing to organic matter as a major driver of C mineralization in sediments, and identifying river deltas and aquacultures as GHG hot spots due to increased organic matter inputs (da Silva et al., 2018; DelSontro et al., 2011; Grinham et al., 2018; Troell et Berg, 1997). The δ^{13} CO₂ and δ^{13} CH₄ signature reflect the source and biochemical processing of C. Measured δ^{13} CO₂ in deep sediments was generally high (> -20, Figure S3.4 and S3.5), indicating a ¹³C enriched source, typical of terrestrial inputs (McCallister et del Giorgio, 2008), and / or a CO₂ consuming process, like methanogenesis by CO₂ reduction. In line with that, the large difference between δ^{13} CH₄ and δ^{13} CO₂ (> 55 ‰) in deep sediment samples (Figure S3.5) suggests a strong isotopic fractionation typical of CH₄ production by CO₂ reduction (Whiticar, 1999). This is also supported by a strong positive correlation between CH₄ concentration and δ^{13} CO₂ in the sediments (R²_{adi}= 0.84, p-value < 0.001, Figure S3.5).

In the top sediment layer (0 - 3 cm), CO₂ concentration was lower than in deep sediments and more homogeneous across sites, while CH₄ concentration was notably lower in the reservoir branch (29 to 100 μ mol.L⁻¹) compared to the main basin (254 to 857 μ mol.L⁻¹). The same spatial patterns were observed for CO₂ and CH₄ sediment to water fluxes averaging 2.3 (SD ± 0.8) mmol.m⁻².d⁻¹ for CO₂, and 0.16 (SD ± 0.12) compared to 0.47 (SD ± 0.58) mmol.m⁻².d⁻¹ for CH₄ in the branches and main basin

respectively. This suggests that CH₄ oxidation in the top sediment layer likely prevents the accumulation of CH₄ in the reservoir branches. That is supported by a steep increase of δ^{13} CH₄ from deep sediments to the overlaying water (Figure S3.4), indicating high net CH₄ oxidation at the sediment-water interface in the branch section. This pattern likely results from higher hypolimnetic O₂ concentration in the branch compared to the main basin (Figure 3.2). In the top sediment layer, δ^{13} CO₂ and δ^{13} CH₄ ranged from -23.2 to -12.3 ‰ and from -76.7 to -58.1 ‰, with a difference between δ^{13} CH₄ and δ^{13} CO₂ lower than 55 ‰ (Figure S3.5), which is more typical of CH₄ production via acetate fermentation (Whiticar, 1999).

The generally lower δ^{13} CH₄ : δ^{13} CO₂ ratio in the top versus deep sediment layers (Figure S3.5) likely reflects a combination of higher CH₄ oxidation in the top layer, and a change in the methanogenic pathway (acetate fermentation to CO₂ reduction). A similar shift in methanogenic pathway was reported in several other studies and related to the decreasing availability of acetate with sediment depth (Alperin *et al.*, 1992; Gruca-Rokosz et Tomaszek, 2015; Hornibrook *et al.*, 2000; Mandic-Mulec *et al.*, 2012). Overall, observations in Batang Ai suggest that CO₂ and CH₄ dynamics in sediments are influenced by organic matter composition and inputs (by inflows and aquaculture), as well as by benthic O₂ availability for CH₄ oxidation.

3.4.3 Hypolimnetic GHG patterns

Across all hypolimnetic samples, CO_2 and O_2 concentrations are strongly negatively correlated with a slope of 0.99 (p-value < 0.001 and $R^2_{adj} = 0.74$, Figure 3.3). This indicates that aerobic respiration, with a typical $CO_2 : O_2$ quotient close to one (Vachon *et al.*, 2020), is a major driver of CO_2 concentration in the reservoir deep layer. The average deep CO_2 concentration gradually increases from the river delta (mean of 82 µmol.L⁻¹) to the west of the main basin (mean of 368 µmol.L⁻¹), mirrored by a decrease in O_2 concentration (Figure 3.3). Based on the difference in mean deep CO_2 concentration between the river delta and the dam wall, we estimated that Batang Ai river inputs have the potential to sustain 31 % of the CO_2 concentration in deep waters discharged from the reservoir. The remaining 69 % is most likely produced by aerobic organic matter respiration along the passage of deep waters through the reservoir branch and main basin (Figure 3.3).

The δ^{13} CO₂ gradually decreases along the branch section (from a mean of -19.5 to -23.3 ‰, Figure 3.3), which may reflect isotopic fractionation of aerobic respiration producing increasingly ¹³C depleted CO₂ along this transect (Blair *et al.*, 1985). In the west section of the main basin (site P9) CO₂ concentration reaches its maximum, however, its ¹³C signature is notably enriched, averaging -19.5 (SD \pm 1.3) % compared to -23.3 (SD \pm 0.6) ‰ at the end of the branch section (site P4, Figure 3.1). An increase in δ^{13} CO₂ indicates either a more ¹³C enriched source of C or a process consuming CO₂. CO₂ consuming processes such as chemiolithotrophy or hydrogenotrophic methanogenesis in deep waters of the west reservoir basin could be at play. However, following this latter hypothesis, one would expect lower hypolimnetic CO₂ concentrations in the west basin whereas measured data show the opposite. A higher ¹³C in the organic matter of the west part of the main basin could be due to a higher retention time in this area which would result in a more degraded and ¹³C enriched organic matter over time. The west part of the main basin, being outside the preimpoundment river channel, may have a higher water residence time, allowing for the consumption of a larger fraction of the organic carbon which becomes an increasingly ¹³C enriched substrate for CO₂. This hypothesis is in line with results from sediment δ^{13} CO₂ values which exhibit a similar spatial pattern (Figure S3.4). Also, the higher residence time would explain the higher accumulation of CO₂ in the hypolimnion of the west basin.

In the branch section, average CH₄ concentration in the deep layer remains low (mostly $< 50 \ \mu mol.L^{-1}$) then increases close to the dam and reaches its maximum in the west part of the main basin (Figures 3). This pattern suggests that the CH₄ exiting the

reservoir was produced mostly in the main basin rather than upstream, with the hypolimnetic CH₄ concentration at the end of the branch (site P4) equal to only 5 % of the one at the dam wall (Site P6). A key element controlling CH₄ dynamics is O₂, which can inhibit its production and promote its consumption via CH₄ oxidation (Borrel et al., 2011). Conversely, Figure 3.3 shows that, above a O₂ concentration threshold of about ~ 30 μ mol.L⁻¹ there is very little CH₄ accumulation in the deep layer (except for two outlier points), whereas under this threshold CH₄ concentrations can increase to much higher levels. In the hypolimnion of the reservoir branch, mean O₂ concentrations varied between 238 and 29 μ mol.L⁻¹ (Figure 3.2), which is high enough to inhibit CH₄ production while allowing for CH₄ oxidation (Borrel et al., 2011; Thottathil et al., 2019). This can explain the lack of CH_4 accumulation in the hypolimnion of the branch section. In the main basin, especially in its west section, O₂ concentrations decrease below ~30 µmol.L-1 (Figure 3.3) allowing for net CH₄ production and accumulation. Highest hypolimnetic CH₄ concentrations were found in the west part of the main basin. This observation is also in line with the hypothesis of a longer residence time in that area, which would allow for more CH₄ accumulation and O₂ depletion there.

The measured hypolimnetic δ^{13} CH₄ signature was highly variable spatially (site and depth) and temporally (sampling campaigns) ranging from -77.8 to 12.7 ‰) from the river delta to the dam wall, but stabilized around an average of -58.4 ‰ in the west part of the main basin (Figure 3.3). This value of -58.4 ‰ is consistent with CH₄ originating from the sediments (pore water δ^{13} CH₄ of -66 ‰) and undergoing partial oxidation (increasing its δ^{13} CH₄). In the rest of the transect however, δ^{13} CH₄ variability spans a range of values from lower to much higher than the typical sediment δ^{13} CH₄ value. This indicates that, in those areas, both CH₄ oxidation (which increases δ^{13} CH₄) and production (which decreases δ^{13} CH₄) occur in the deep water column. However, the net result of these two processes is a generally low CH₄ concentration (< 50 µmol.L⁻¹, Figure 3.3) from the river delta to the dam wall.



Figure 3.3: Left panels show CH₄ and CO₂ average hypolimnetic concentrations (circles) and δ^{13} C values (triangles) as a function of distance from the river delta (0 km is site P1) for each sampling campaign (blue, orange, red, and green represent Nov-Dec 2016, Apr-May 2017, Feb-Mar 2018, Aug 2018 respectively, and grey boxes are boxplots of all campaigns). The right panels show measured deep CO₂ and CH₄ concentration as a function of O₂ concentration, with a color gradient representing an increased distance from the river delta (orange to green to blue). The regression on the top right panel has a slope of 0.99, a p-value < 0.001 and a R²_{adj} = 0.74.

3.4.4 Water column metabolic rates

Measured deep aerobic respiration rates based on O_2 changes in incubations averaged 15.6 (SD ± 10.5) µmol.L⁻¹.d⁻¹. These values are in the range of previously reported respiration rates in clear water oligotrophic lakes (Bogard *et al.*, 2020 ; Hanson *et al.*, 2003 ; Pace et Prairie, 2005). Due to the limited and non-normally distributed data points, it is difficult to clearly identify the factors controlling variability in measured respiration rates. Nevertheless, a non-parametric Kendall test showed a positive correlation between aerobic respiration rates and O_2 concentrations (p-value = 0.002, Figure S3.3), suggesting that O_2 can become a limiting factor for aerobic respiration in the deep layer of Batang Ai.

Net CH₄ oxidation rates (MOX) measured in *in situ* incubations ranged from 0.003 to 9.2 μ mol.L⁻¹.d⁻¹ (a few incubations in the epilimnetic layer showing a slight net CH₄ production were excluded from the analysis). The contribution of MOX to aerobic respiration (O₂ consumption in incubations) in the deep layer ranged from negligible (<1%) to more than 100%, and increased following a near-perfect linear fit with initial CH₄ concentration (R²_{adj} = 0.98, p-value < 0.001, Figure S3.6). Indeed, sites with high CH₄ concentration tend to coincide with low O₂ concentration and respiration, and high MOX rate (R²_{adj} = 0.78, p-value < 0.001, Figure S3.6). This makes MOX the main O₂ consuming process in the CH₄ rich deep hypoxic sites of Batang Ai. This is in line with results of Reis et al. (2020a) (submitted) finding unprecedented high levels of abundance and activity of CH₄ oxidizing bacteria in the hypoxic hypolimnion of Batang Ai (up to 19% of prokaryotic cell abundance and up to 60% of rRNA sequences).

MOX dependence on CH₄ and O₂ concentrations has been previously reported, with maximum rates at high CH₄ and low O₂ concentrations (Kankaala *et al.*, 2013b; Michaud *et al.*, 2017; Rudd *et al.*, 1974; Thottathil *et al.*, 2019). Thottathil et al. (2019) proposed a non-linear model (Eq. (3.12)) to account for the inhibitory effect of high O₂

on CH₄ specific oxidation rates (MOX_{sp}). To assess whether the same model structure was applicable to Batang Ai, we tested it on two sets of data: incubation derived MOX_{sp}, and F_{ox} (fraction of oxidized CH₄) calculated from δ^{13} CH₄ data in vertical profiles (section 3.3.4). In both cases, the best fit was obtained by a unimodal curve shape with inhibition at high O₂. While the MOX_{sp}-O₂ link was not as obvious in incubations results, field-based F_{ox} estimates strongly supported a unimodal relationship between the two. Our results are in good agreement with Thottathil et al. (2019) (Figure S3.7) suggesting that the MOX_{sp}-O₂ kinetic they describe is widely applicable. Nevertheless, unexplained variance was high in both empirical models in Batang Ai (incubations and F_{ox}), which suggests other local drivers of MOX_{sp} may be at play, possibly the microbial community structure (Reis *et al.*, 2020b).

In all vertical profiles, F_{ox} reaches high values (> 0.6), indicating all sites experience substantial CH₄ oxidation at a given depth. However, the depth at which highest CH₄ oxidation states are observed differ among sites, occurring higher in the sediments and water column when moving from upstream (river delta) to downstream (main basin). Indeed, in the river delta, CH₄ oxidation is apparent mostly within the sediments and at the sediment-water interface (Figure S3.4, increase in δ^{13} CH₄). Whereas in the branch, maximum F_{ox} are observed starting near the bottom of the hypolimnion, and gradually move higher in the water column, occurring at the top of the hypolimnion in the main basin (Figure S3.8). This spatial pattern of F_{ox} driven by O₂ concentrations (Figure S3.7) shapes deep CH₄ accumulation along the river to reservoir continuum (Figure 3.3). This effect is also reflected at a temporal scale, with mean hypolimnetic δ^{13} CH₄ values in the reservoir branch being higher in Aug 2018 (Figure 3.3), when O₂ concentrations are lower and closer to the MOX_{sp} optimal O₂ concentration compared to other seasons (Figure 3.1). Overall, experimental and field-based data show that metabolic processing of O₂, CO₂, and CH₄ are interconnected across space and time in the deep layer of Batang Ai, with CH₄ oxidation playing a major role in the hypoxic part of the hypolimnion.

3.4.5 Temporal evolution of deep GHG

The large stocks of both CO₂ and CH₄ in the hypolimnion raise the question of whether GHG are still accumulating in the deep layers of Batang Ai or instead have reached an equilibrium. To explore this question, we combined the rates of gas diffusivity, sediment production, and water column metabolism in the model framework described above (section 3.3.5) to recreate the evolution of O_2 , CO_2 , CH_4 concentrations and δ^{13} CH₄ through time since the impoundment of Batang Ai. Assuming that the physical and metabolic parameters have remained unchanged, results of the simulation show a rapid decrease of hypolimnetic O₂ concentration mirrored by an increase in CO₂ concentration immediately after impoundment (Figure 3.4). Within only two months, O₂ reaches a quasi-steady state being completely consumed in most of the hypolimnion due to aerobic respiration. At this stage, CO_2 production by aerobic respiration also stops in deeper layers, but CO₂ continues to accumulate due to anaerobic processes for ~ 10 years (Figure 3.4). Likewise, CH₄ hypolimnetic concentrations start increasing rapidly after the onset of anoxic conditions (during the first year), and continue to rise more slowly over ~ 10 years before reaching a quasi-steady state. Our results strongly argue that, despite the permanent stratification, the reservoir stops accumulating GHG in its hypolimnion a few years after its construction because the total amount of hypolimnetic CO₂ and CH₄ production are in a dynamic equilibrium with vertical gas transport and CH₄ oxidation in the upper portion of the deep layer.

Because of the deep water withdrawal in Batang Ai, hypolimnetic gas concentrations control GHG emissions downstream of the dam, which represent 89 % of the reservoir GHG footprint (Soued et Prairie, 2020). According to modeled deep CO_2 and CH_4 concentrations, downstream emissions increased mostly during the first year after impoundment, and then only slightly over ~10 years. Studies on Petit Saut and Eastmain-1 reservoirs both found CO_2 and CH_4 emissions below the dam to increase the first 1-2 years (Abril *et al.*, 2005 ; Teodoru *et al.*, 2012). This initial increase fits

the results in Batang Ai, and illustrates the initial peak in flooded organic matter decomposition. Surface GHG emissions are thought to decrease exponentially with reservoir age due to the degradation of flooded organic matter. However, our model results, in line with previous studies, suggest that downstream emissions follow a different temporal dynamic than surface ones, being potentially higher in older reservoirs when permanently stratified.



Figure 3.4: Modeled temporal evolution of O₂, CO₂, CH₄ concentrations and δ^{13} CH₄ along a vertical profile in the deep layer of Batang Ai. Lines represent different times after the reservoir impoundment: 5 days (dotted), 15 days (dash and dot), 100 days (short dash), 1 year (long dash), and 10 years (solid), with the light to dark gradient in line color following the same order.

3.4.6 Modeled vs measured deep GHG

Model accuracy can be evaluated by comparing measured versus modeled results at the time of sampling. In the case of O_2 , the model reproduces quite well the shape of the curve and concentrations along the depth profile ($R^2 = 0.91$, NSE = 0.75, Figure

3.5). Modeled CO₂ values also match well measured data ($R^2 = 0.82$, NSE = 0.50, Figure 3.5), indicating that the main processes are well formulated in the theoretical framework, with hypolimnetic CO₂ buildup originating from both aerobic and anaerobic processes. For comparison, the CO₂ profile modeling was tested without the inclusion of anaerobic production, and this scenario resulted in a large underestimation of hypolimnetic CO₂ buildup (Figure 3.5).

For CH₄, modeled and observed concentrations were well correlated but the model substantially underestimated the actual accumulation in the hypolimnion, with predicted bottom concentration reaching 56 µmol.L⁻¹ compared to a measured value of 290 µmol.L⁻¹ (R² = 0.83, NSE = 0.41 on log values, Figure 3.5). However, the model reproduced well the CH₄ isotopic signature (δ^{13} CH₄) profile and exhibited the same peak at ~13 m as the one observed (R² = 0.67, NSE = 0.52, Figure 3.5). This highly enriched δ^{13} CH₄ deep layer peak is the direct result of the inhibition of CH₄ oxidation at high O₂. It therefore suggests that the non-linear dependence of CH₄ oxidation on O₂ was well defined in the model.

The discrepancy in CH₄ concentrations potentially indicates an overlooked process in the theoretical model framework, which considered sediments as the sole source of CH₄. The formation and dissolution of bubbles may constitute a CH4 source not explicitly considerd in the model (McGinnis *et al.*, 2006). Another likely source of CH₄ is the production in the fully anoxic layer of the water column. While this process is much less documented than sediment production, it has been found in different systems (Biderre-Petit *et al.*, 2011 ; Iversen *et al.*, 1987 ; Winfrey et Zeikus, 1979 ; Youngblut *et al.*, 2014). The permanently anoxic hypolimnion of Batang Ai offers ideal conditions for water column CH₄ production, making it a likely process in this system. As a firstorder approximation, we evaluated its potential effect on model fitness by incorporating this putative source in as a negative exponential function of O₂ concentration as follows:

$$P_{WCH4} = P_{Wmax} \ e^{-[O_2]} \tag{3.15}$$

with P_{WCH4} the potential water column CH₄ production at a given point, and P_{Wmax} the maximal value it can attain. The maximum value (P_{Wmax}) is unknown for Batang Ai so we used two values derived from the literature (0.01 and 1 μ mol.L⁻¹.d⁻¹) as lower and upper bounds (Iversen et al., 1987; Winfrey et Zeikus, 1979). Results of the simulation show that, at its upper bound, this additional source yields ~7 fold increase in mean hypolimnetic CH₄ concentration, matching well with measured data. This increase is not observed in the top part of the deep layer, where oxidation consumes excess CH₄, so it does not affect surface emissions. However, it influences CH₄ concentration at the withdrawal depth and could thus substantially influence downstream emissions. While we have no direct measurements supporting the occurrence and amplitude of deep water column CH₄ production in Batang Ai, model simulations suggest that it has the potential to drastically increase hypolimnetic CH₄ concentrations. Yet, there is very little information about this process in the literature, and it is largely overlooked in aquatic CH₄ budgets, being considered low compared to sediment production (Bastviken et al., 2008; Eckert et Conrad, 2007; Liikanen et al., 2002). Our results suggest that water column CH₄ production may be an important process to investigate in large systems with a higher ratio of anoxic water volume to sediment area such as Batang Ai.



Figure 3.5: Modeled (solid line) versus measured (gray circles) O₂, CO₂, CH₄ concentrations and δ^{13} CH₄ along a vertical profile in the deep layer of Batang Ai. In the [CO₂] panel, the dotted line represent a model simulation without anoxic CO₂ production. In the [CH₄] and δ^{13} CH₄ panels, dashed lines represent model simulations with two levels of water column CH₄ production (note that the line for low production is hidden by the black solid lines).

3.4.7 Model sensitivity to parameters

While our model was able to capture the main features of the GHG profiles, we further investigated to which parameters it was most sensitive to and their impacts on deep CO_2 and CH_4 patterns in Batang Ai are shown in Figure 3.6. Gas diffusivity changes had the largest impact on CO_2 and CH_4 concentrations both at the withdrawal depth and as an integrated profile average. Gas diffusivity directly impacts CO_2 and CH_4 through vertical transport, which affects gas accumulation rates, and indirectly through its effect on O_2 concentration, which modulates both aerobic respiration and CH_4 oxidation. Lower gas diffusivity increases the gas residence time allowing for a higher CO_2 and CH_4 accumulation, as well as more O_2 depletion, which limits CH_4 oxidation.

The importance of gas diffusivity in Batang Ai is in line with the study of Vachon et al. (2019) which highlighted the importance of water column stability in a small lake CH₄ budget. The effect of physical water column structure on GHG emissions is also known to control variations in emissions downstream of dams (Bastien *et al.*, 2011) and to trigger emission peaks during lake turnovers (Encinas Fernández *et al.*, 2014). Given the importance of gas diffusivity, elements such as vertical advection via turbine withdrawal and thermal convection, not considered in the model, could play a large role in shaping hypolimnetic gas patterns at shorter time scales (daily to seasonal) in Batang Ai. Studies on freshwater CO_2 and CH_4 dynamics often ignore the physical dimension of their cycling, focusing mainly on their biological production and consumption. While biological factors are undoubtedly important, our results show that their interaction with gas diffusivity is a key element to consider at an ecosystem scale.

Sediment fluxes to the water column was the second most important parameter in terms of CO_2 and CH_4 deep layer concentrations (Figure 3.6). For both gases, this represents mostly the effect of anoxic processes. Indeed, for CO_2 the aerobic portion of sediment respiration is negligible in the hypoxic to anoxic water column due to its dependence on O_2 availability. While methanogenesis is extensively studied (Conrad, 2009), there is comparatively little information about the mechanisms, rates, and drivers of anaerobic CO_2 production in freshwater sediments. The latter component is a large source of uncertainty in the model due to the absence of direct measurements in Batang Ai and in other systems. Nevertheless, our modeling results in Batang Ai clearly show that anoxic sediment processes are determinants for both CO_2 and CH_4 accumulation in the deep layer and emissions to the atmosphere. Sediment processes are a particularly important component in reservoirs where the flooding of soils represent an additional substrate for anoxic GHG production (Prairie *et al.*, 2018).

Surprisingly, aerobic respiration rate had little effect on modeled CO_2 accumulation in the hypolimnion. Changing R_w resulted in small variations in O_2 and CO_2

concentrations only in the top 10 m of the deep layer (Figure 3.6) where O₂ penetration from the surface layer occurs. The rest of the water column, being anoxic, was not affected by R_w. However, R_w did have a notable effect on modeled CH₄ concentration. Indeed, by changing even slightly the O₂ profile in the top 10 m, variations in R_w induced changes in the depth of the CH₄ oxidation peak (visible on the δ^{13} CH₄ graph, Figure 3.6). Lower R_w resulted in slightly more O₂ penetration in the deep layer leading to a deeper CH₄ oxidation peak reducing CH₄ accumulation and downstream emissions.

Surprisingly, the magnitude of maximum specific CH₄ oxidation had only a small impact on the CH₄ concentration profile (Figure 3.6), notably smaller than any of the other tested parameters. While counter-intuitive at first glance, MOX is largely confined to the top section (< 20 m) of the deep layer due to its dependency on O₂ availability and therefore does not affect substantially deep CH₄ accumulation that is mainly concentrated in the anoxic deep section (> 20 m). This result on model sensitivity to MOX rate indicates that the spatial interaction between MOX, O₂, and CH₄ concentrations is more important than the magnitude of MOX rate alone. In other words, where MOX occurs constrains how much CH₄ it is able to consume. Thus, doubling the MOX rate at a depth where CH₄ concentration is low has little effect on the gas profile, whereas changing the O₂ profile and hence the depth of the MOX peak (case of R_w variations, Figure 3.6) substantially affects CH₄ accumulation. This, in line with our previous results, reinforces the importance of integrative approaches considering the interactions between multiple factors to better understand and predict freshwater GHG patterns.


Figure 3.6: Modeled (solid line) and measured (gray circles) O₂, CO₂, CH₄ concentrations and δ^{13} CH₄ along a vertical profile in the deep layer of Batang Ai. From top to bottom panels, graphs show results of sensitivity analyses of gas diffusivity (K_z), Sediment inputs (S_{CO2} and S_{CH4}), aerobic water column respiration (R_w), and CH₄ oxidation (MOX) respectively. Short and long dashed lines represent respectively halving and doubling of the parameter.

3.5 Conclusion

The general aim of this study was to improve our understanding of CO₂ and CH₄ dynamics in the deep layer of lakes and reservoirs; in particular, the spatial patterns of these gases, their temporal evolution, and their controlling factors. Our results from Batang Ai reservoir showed vertical variations in deep CO₂ and CH₄ concentrations reaching several orders of magnitude, tightly linked to O₂ concentrations. O₂ affected deep aerobic respiration, methanogenesis, and methanotrophy, and was thus a central player in the spatial structure of CO₂ and CH₄ vertically (with depth) and horizontally (along the river-reservoir continuum). Our simulations of the temporal evolution suggest that O₂ was rapidly depleted in the deep layer of the reservoir over a period of a few months after its construction. This constrained aerobic processes to the metalimnion, while anaerobic production fueled the accumulation of CO₂ and CH₄ in the hypolimnion for a decade before reaching a quasi-steady state. Comparing modeled to measured data suggested that anaerobic CO₂ production and water column methanogenesis are likely important but generally overlooked CO₂ and CH₄ sources. The importance of anaerobic CO₂ and CH₄ production in the sediments was corroborated by sensitivity analyses on the model. Latter analyses also showed that metabolic production of CO_2 (by aerobic respiration) and consumption of CH_4 (by oxidation) were highly dependent on the vertical O₂ profile. O₂, CO₂, and CH₄ profiles were all influenced by water column gas diffusivity, with higher mixing leading to less GHG accumulation and more oxygenation, which enhanced CH₄ oxidation and aerobic respiration. Thus, gas diffusivity played a paramount role in deep GHG dynamics, having a direct impact on their movement and an indirect one on their metabolism. This highlights the importance of comprehensive modeling approaches to understand the relative role and interactions between factors controlling CO₂ and CH₄ dynamics at an ecosystem level. In that sense, our integrative approach increases our ability to predict deep GHG behavior, and hence reservoirs downstream emissions.

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3.7 Supplementary information



Figure S3.1: Time-series of air temperature, wind speed, rainfall, and inflow rate in Batang Ai (data provided by Sarawak Energy hydroelectric company). Gray bars represent sampling campaigns.



Figure S3.2: Vertical profile of N^2 in the deep layer of Batang Ai. Orange points are the result of a smooth spline applied to a measured temperature profile and the red line is the modeled N^2 derived from fitting a sigmoid curve to the orange data points. The Nash-Sutcliffe index is 0.84 (on log values). The gray dashed line indicates the top of the metalimnion.



Figure S3.3: Measured (points) and modeled (line, derived from Eq. (3.10)) water column aerobic respiration rates (R_w) in Batang Ai deep layer (>10 m) as a function of available O₂ concentration.



Figure S3.4: Spatial interpolation of pore water sediment CO_2 and CH_4 concentrations and isotopic signatures as a function of sediments depth (+ overlaying water 1 cm above) and distance from the river delta. The solid line represents the sediment surface, and the black circles represent the sampled points.



Figure S3.5: Correlation between δ^{13} CH₄ and δ^{13} CO₂ in sediment and deep water samples (left panel) and between CH₄ concentration and δ^{13} CO₂ in the sediments (right panel, p-value < 0.001 and $R^2_{adj} = 0.84$). Light to dark blue represent shallow to deep water samples, and light to dark brown represent shallow to deep sediments samples. In the left panel, triangles and circles represents sites in the reservoir branches and main basin respectively, and dotted diagonals represent 8 different CO₂ – CH₄ fractionation lines (of 90, 80, 70, 55, 40, 30, 20 and 5).



Figure S3.6: CH₄ oxidation rate (MOX, point size and color gradient) and its contribution to total O₂ consumption by aerobic respiration (MOX : R_w ratio) as a function of initial CH₄ concentration in deep water incubations in Batang Ai. The ratio of O₂ : CH₄ consumed (ranging typically from 1 to 2) was set to 1 since a higher value would make the MOX : R_w largely exceed the maximal possible limit of 1. The line represents the linear regression of the variables on the two axes (p-value < 0.001, R^2_{adj} = 0.98). The linear regression of MOX vs initial [CH4] (line not shown) had a p-value < 0.001 and an R^2_{adj} = 0.78.



Figure S3.7: Specific CH₄ oxidation (MOX_{sp}, derived from incubations, purple squares) and fraction of CH₄ oxidized (F_{ox} , derived from δ^{13} CH₄ in vertical profiles, blue points) as a function of O₂ concentration. Lines represent Eq. (3.14) model fitted to the data in purple ($\lambda = 0.004$ and $\beta = 0.02$, p-values of 0.01 and 0.19 respectively), and using parameter values of Thottathil *et al.*, (2019) in gray ($\lambda = 0.01$ and $\beta = 0.18$).



Figure S3.8: Fraction of CH₄ oxidized (F_{ox}) as a function of depth and distance from the river delta for all sites and sampling campaigns.

CONCLUSION

La recherche présentée s'inscrit dans un contexte ou les réservoirs sont des sources de GES (surtout de CO₂ et de CH₄) vers l'atmosphère, mais comprendre et prédire leurs émissions est un défi nécessitant un approfondissement des connaissances sur le sujet. L'objectif général de la thèse est d'obtenir un portrait complet de la dynamique du CO₂ et du CH₄ à l'échelle écosystémique dans un réservoir tropical. En premier lieu, nous voulions comprendre où, quand, et comment les GES sont émis à l'échelle du réservoir, et tester notre capacité actuelle à les prédire (Chapitre I). Pour cela, nous avons fait une évaluation exhaustive des flux de GES du réservoir Batang Ai, prenant en compte la variabilité spatiale et temporelle, ainsi que toutes les voies d'émission en amont et en aval du barrage. Nous avons ensuite comparé les valeurs mesurées sur le terrain aux prédictions des modèles les plus récents pour tester leur performance et identifier leurs failles. Nous voulions ensuite définir l'origine et les facteurs qui déterminent l'ampleur des flux, en surface du réservoir (Chapitre II), et en aval (Chapitre III). Nous avons donc construit un bilan de masse des apports / exports de CO_2 et de CH_4 dans l'épilimnion du réservoir pour déterminer la variabilité et la contribution relative des différentes composantes au flux de surface (Chapitre II). Nous avons ensuite mesuré les taux de production, de consommation, et de transport du CO₂ et du CH₄ dans la couche profonde (qui alimente les émissions en aval), et modélisé l'interaction de ces processus dans l'espace et dans le temps (Chaputre III).

Les résultats du Chapitre I ont révélé l'importance, souvent négligée, de la résolution spatiotemporelle de l'échantillonnage et de l'inclusion de toutes les voies d'émission de GES dans l'estimation de l'empreinte C d'un réservoir. En amont du barrage, les flux diffusifs et ebullitifs mesurés sont relativement faibles et surestimés par les

modèles, dû au fait que ces derniers ne prennent pas en compte l'effet de la composition du sol argileux et de la forte pente qui empêche l'accumulation de matière organique dans la zone littorale. Cependant, la plus grande partie des émissions (89 %) ont lieu en aval du barrage, une voie d'émission rarement prise en compte dans la littérature. D'ailleurs, les modèles prédictifs testés sur Batang Ai sous-estiment largement ces émissions en aval, ne considérant que le dégazage du CH₄ dans cette section. En somme, le réservoir étudié émet une grande quantité de GES comparé à d'autres, surtout en aval du barrage et sous forme de CH₄. En effet, la construction du barrage a créé un vaste environnement anoxique (l'hypolimnion) propice à la production de CH₄, et offre aussi une échappatoire directe à ce gaz, à travers la sortie d'eau située en profondeur. Nous avons estimé qu'une sortie d'eau située 5 m plus haut réduirait drastiquement (de 77 %) l'empreinte C du réservoir.

Les résultats du Chapitre II ont mis en évidence les sources et processus régulant les flux diffusifs en surface du réservoir. Ces sources changent selon un gradient hydrologique allant des branches du réservoir au bassin principal, ce qui se reflète dans la structure spatiale des concentrations et des signatures isotopiques du CO₂ et du CH₄. Dans les branches du réservoir, l'apport de GES des flux horizontaux du milieu terrestre et riverain est élevé, et a le potentiel d'alimenter plus de 90 % des émissions de CO₂ et CH₄ dans cette section. Cependant, l'influence de ces apports provenant du bassin versant est localisée, et devient négligeable dans le bassin principal, où les sources de CO2 et CH4 divergent créant un découplage entre ces deux gaz. En effet, dans le bassin principal les apports verticaux des couches profondes vers l'épilimnion sont une importante source de CO₂ mais pas de CH₄, et inversement, les apports des sédiments contribuent substantiellement aux émissions de CH_4 (25 %) mais pas de CO_2 (7 %). Cependant, la dynamique des deux gaz dans l'épilimnion est fortement influencée par leur métabolisme dans la colonne d'eau, qui est en même temps un élément central, mais très variable, ajoutant une grande incertitude aux budgets de gaz dans les deux sections du réservoir. Ainsi, le métabolisme dans la colonne d'eau soutient jusqu'à 75 % des émissions de CH_4 dans le bassin principal, et module la direction et l'amplitude des flux de CO_2 .

Dans la couche profonde (Chapitre III), l'influence du métabolisme sur la dynamique des GES est aussi présente, et dépend fortement de la disponibilité de l'O₂ pour la respiration aérobique (production de CO₂) et pour l'oxydation du CH₄. Le lien entre les trois gaz (O₂, CO₂, et CH₄) crée un patron spatial le long du continuum hydrologique, où la concentration de CO₂ augmente et celle de O₂ diminue graduellement dans la couche profonde, alors que le CH₄ s'accumule uniquement dans le bassin principal devenu hypoxique / anoxique. La modélisation montre que l'accumulation de GES dans le bassin principal atteint un état d'équilibre environ 10 ans après la construction du barrage, et ce malgré la stratification thermique permanente. La quantité de GES accumulée dans la couche profonde module les émissions de GES en aval du barrage, et dépend de l'interaction entre plusieurs facteurs biogéochimiques. Notamment, la diffusivité verticale des gaz qui joue un rôle prédominant dû à son influence directe sur le mouvement des gaz, mais aussi indirecte via son effet sur la pénétration de l'O₂ dans la couche profonde qui affecte le métabolisme du CO₂ et du CH₄.

Dans l'ensemble, les résultats de l'étude montrent que les émissions en surface du réservoir sont relativement faibles et reflètent en grande partie les apports du bassin versant et le métabolisme interne de la colonne d'eau, tandis que les émissions en aval du barrage sont très élevées et surtout contrôlées par l'O₂ et la le transport vertical des des gaz dans la couche profonde (Figure 4.1). À l'échelle du paysage, l'impact de la construction du barrage sur le cycle du C est surtout apparent avec la création de la couche profonde anoxique du bassin principal, qui accumule de grandes quantités de GES notamment sous forme de CH₄. Dans une perspective de gestion, nos résultats indiquent que l'empreinte climatique du réservoir serait réduite par l'élévation de la prise d'eau, l'oxygénation de l'hypolimnion, et une augmentation du mélange des eaux profondes. D'un point de vue théorique, cette thèse a permis une compréhension

intégrale de la dynamique du CO₂ et du CH₄ à l'échelle de l'écosystème, et a identifié les facteurs de régulation les plus importants. L'étude a aussi mis en lumière plusieurs lacunes dans les connaissances actuelles sur le sujet: l'effet de la composition du sol et de la pente sur les flux diffusifs et ebullitifs (Chapitre I), les réactions et facteurs de contrôle du métabolisme du CO₂ et du CH₄ (Chapitre II), la production de CO₂ en anoxie et de CH₄ dans la colonne d'eau (Chapitre III). Explorer les éléments mentionnés dans des recherches futures améliorerait la modélisation des concentrations et de flux de GES dans les réservoirs.



Figure 4.1: Schéma des principales conclusions des trois chapitres de la thèse.

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