High contribution of methane in greenhouse gas emissions from a eutrophic lake: a mass balance synthesis

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Lakes are 'hotspots' for greenhouse gas (GHG) emissions, primarily carbon dioxide (CO₂) and methane (CH₄). Understanding the processes regulating GHG emissions from lakes, and their temporal variability, is essential for more accurately quantifying the role of lakes in global GHG cycles. In this study, we identified the processes that affect CO₂ and CH₄ concentrations in a small (0.3 km²) eutrophic monomictic lake (Okaro, New Zealand). A mass balance model was used to calculate changes in CO2 and CH4 storage in the lake as a result of internal cycling and atmospheric fluxes. To support model computations, CO₂ and CH₄ concentrations profiles were measured monthly over a one-year period, in addition to temperature, dissolved oxygen and chlorophyll a. Annually, Lake Okaro acted as a sink of CO₂ from the atmosphere (425.4 mmol CO₂ m⁻² y⁻¹) and a source of CH₄ (553.4 mmol CH₄ m⁻² y⁻¹) equating to a net GHG emission (diffusive fluxes of CO₂ and CH₄ combined) of 0.22 kg CO₂-eq m⁻² y⁻¹. Although it may be viewed as conservative and applies only to diffusive fluxes, our study indicates that eutrophic lakes with high rates of primary production may act as a net source of GHGs.

Keywords: lake, eutrophic, greenhouse gas, CO₂, CH₄, thermal stratification.

Introduction

Inland waters (including ponds, lakes, wetlands, reservoirs, stream and rivers) are an important component of the global carbon balance (Cole et al. 2007; Tranvik et al. 2018; Bastviken et al. 2011) and important in the transfer of terrestrial carbon to the ocean (Aufdenkampe et al., 2011; Regnier et al., 2013). They cover <3% of the Earth's surface (Downing et al., 2006) and emit about 0.8 – 2.1 Pg C y⁻¹ to the atmosphere as CO₂ (Cole et al., 2007; Raymond et al., 2013; Regnier et al., 2013). These emissions are similar in magnitude to global CO₂ atmospheric sinks of the open ocean (1.85 – 2.6 Pg C y⁻¹) and terrestrial systems (2.05 – 2.9 Pg C y⁻¹: Regnier et al., 2013; Le Quéré et al., 2015). Furthermore, Bastviken et al. (2011) estimated that 0.65 Pg C y⁻¹ is released to

the atmosphere as CH₄ from inland waters. Given that global warming potential of CH₄ over a 100-year horizon is 34 times larger than that of CO₂ (Myhre et al. 2013), inland waters are clearly highly important in global GHG emissions accounting.

Improvements in environmental sensor technology have enabled direct measurements of GHG emissions (e.g., Eugster et al. 2011; Gonzalez-Valencia et al. 2014). Together with more statistically robust upscaling techniques (Beaulieu et al. 2016; Wik et al. 2016), these advances have increased the accuracy of GHG emission estimates for inland waters. Better understanding of carbon processing by inland waters is essential, however, to improve GHG accounting (Hanson et al. 2015). Recent studies have shown that organic carbon processing in lakes, and the associated carbon emissions, are closely related to lake trophic status (Almeida et al. 2016; Santoso et al. 2017; Evans et al. 2017). Knowing that inland waters are dominated by small water bodies with surface area <1 km² (Messager et al. 2016), which are prone to eutrophication and high rates of organic carbon burial (Downing et al. 2008), small eutrophic lakes may make a significant contribution to global carbon budgets and climate change (Tranvik et al. 2018; Moss et al. 2011). Identification of the key parameters regulating eutrophication processes and GHG emissions from small lakes is therefore important to quantify these contributions.

Lakes with anoxic hypolimnia accumulate large quantities of CH₄ and CO₂ (Huttunen et al. 2003; Juutinen et al. 2009; López Bellido et al. 2009). In mono- and dimictic lakes, these stored gases can be released rapidly to the atmosphere during turnover periods (Fernández et al. 2014; Weyhenmeyer et al. 2015; Kankaala et al. 2007). Recent global estimates of 1.25 – 2.30 Pg of CO₂-eq y⁻¹ from lakes and impoundments are nearly 20% of those from fossil fuels (DelSontro et al. 2018). Both CO₂ and CH₄ are highly important in contributing to GHG emissions from lakes, as well

as N₂O, which is a more potent greenhouse gas also emitted from inland waters (Soued et al. 2016).

Eutrophication associated with greater productivity increases autotrophic fixation rates of CO₂ (Davidson et al. 2015) but is also associated with higher respiration rates (Yvon-Durocher et al., 2010). Organic carbon processing in the sediment of productive lakes may produce substantial quantities of dissolved inorganic carbon and CH₄ which, if subsequently exposed to the atmosphere, leads to high GHG emission (Bastviken et al. 2008; Walter et al. 2006). Previous estimates suggested that CO₂ and CH₄ contribute equally to the global warming potential of GHG emissions from lakes and impoundments (Tranvik et al. 2009; Bastviken et al. 2011). A recent synthesis, however, showed that eutrophic lakes release more CH₄ to the atmosphere (DelSontro et al. 2018). Under future scenarios of increasing nutrient loads to inland waters, CH₄ emissions from lakes and impoundments may substantially increase over the next century (Beaulieu et al. 2019).

Currently, estimates of CO₂ and CH₄ emissions from inland waters and their contributions to global warming potential are mainly based on statistical and upscaling methods (DelSontro et al. 2018; Beaulieu et al. 2019). Lack of detailed information on the interactions of biological and physical factors that affect the magnitude of the GHG fluxes limits the development of effective mitigation strategies to use in inland waters to combat global warming (Moss et al. 2011; Hamilton et al. 2013). Process based estimates could provide inexpensive alternatives in addressing ecosystem questions regarding the role of lakes in regulating carbon and GHG emissions (e.g. Hanson et al. 2004).

In this study, we applied a simple process-based mass balance model to a monomictic eutrophic lake to account for internal CO₂ and CH₄ cycling and the

atmospheric fluxes of these gases. The model was calibrated with measurements of dissolved CO₂ and CH₄ concentration profiles. Through this model we estimate the proportion CO₂ and CH₄ emissions in the total emitted GHGs. We compare our estimates of carbon emissions from our study lake with a recent synthesis of carbon emissions from lakes globally (DelSontro et al. 2018). We also discuss the role of lakes in regulating GHGs and climate change (Moss et al. 2011), specifically in relation to the magnitude of emissions from productive eutrophic lakes.

Methods

Study site

The study site was Lake Okaro, 27 km south of Rotorua township, North Island of New Zealand (Figure 1). The lake is small (0.32 km²), with an average depth of 12.5 m (max. 18 m) and is the most eutrophic of 12 major lakes in the Rotorua region. More than 90% of its catchment area (3.98 km²) is pasture used for dairy, deer, sheep and beef farming. Water residence time of Lake Okaro is estimated to be ca. 1.5 years, with surface inflows from two unnamed streams entering the lake from the north-west, and one outflow (Haumi Stream) in the south-east. The trophic state of this lake has increased since the 1960s (Forsyth et al. 1988) and is now considered eutrophic, with periodic blooms of cyanobacteria in spring and summer (Paul et al. 2008). The recorded chlorophyll a concentration ranged from 0.5 to 283 µg L⁻¹ (Santoso et al. 2017). The hypolimnion of the lake is anoxic throughout most of the 9-month stratified period, from October to June (Özkundakci et al. 2011).

The mass balance model

A conceptual model was developed to calculate the whole-lake CO2 and CH4 fluxes and

mass balance (Figure 2). The model was used to calculate changes in CO₂ and CH₄ stored in the lake due to 1) diffusion of CO₂ and CH₄ across the sediment-water interface, 2) atmospheric diffusive fluxes, 3) aerobic oxidation of CH₄, 4) net CO₂ uptake by phytoplankton, 5) phytoplankton decomposition, and 6) oxidation of dissolved organic carbon (DOC) (Table 1). The model assumed that lake sediment was the main source of CO₂ and CH₄. Phytoplankton respiration and decomposition, CH₄ oxidation and DOC oxidation supply CO₂ to the water column. Contributions of littoral aquatic plans to CO₂ and CH₄ storage in the lake were not included in the model as the depth limit for aquatic macrophytes (~4.5 m: Vant et al. 1996) is around 6% of the lakebed area and macrophytes were observed only at low densities. Bubble dissolution associated with ebullition of CH₄ from the lake sediment may contribute to the CH₄ pool in the water column (Schmid et al. 2017) but this process was not included in the model due to: 1) the stochastic process of bubble production and release (Varadharajan and Hemond 2012), and 2) gas bubbles that may escape CH₄ oxidation during transport to the water surface (DelSontro et al. 2010). Processes that may contribute to the wholelake CO₂ and CH₄ mass balance but were not conceptualized in the model are elaborated upon in the discussion.

The model was used to estimate changes in the whole-lake mass storage of CO_2 and CH_4 (ΔCO_2 and ΔCH_4 : Eqns 1 and 2, Table 1) based on the processes described above (see Eqns 3 to 8, Table 1). Assumptions used in the model are given in Table 2. The whole-lake mass storage was calculated as the total of CO_2 and CH_4 stored in 2 m depth intervals based on CO_2 and CH_4 concentrations in the layer and the hypsography of the lake.

The bottom sediment flux is the major source of GHG to the lake. Its dominance arises from high rates of remineralization of organic carbon accumulated in

the bottom sediments (Kuivila et al. 1988). The sediment of Lake Okaro has been shown to remineralize carbon at a rate of 2.32 mol C m⁻² y⁻¹, with <0.01 mol C m⁻² y⁻¹ in refractory form (Santoso et al. 2017). The carbon flux from the sediment to the water column was therefore estimated to be 6.35 mmol C m⁻² d⁻¹. Up to 60% of this remineralized carbon may be dissolved inorganic carbon, while 40% may be CH₄ (Klump et al. 2009). Data from Bastviken (2009) suggest that CH₄ is oxidized to CO₂ in the presence of DO, with an oxidation rate constant (*K*_{CH4-O2}; Table 2) in the range 0.001 – 1.0 d⁻¹, including in eutrophic lakes. We estimated the CH₄ water column oxidation rate as a first order reaction following Lopes et al. (2011), with Arrhenius equation adjustment for water temperature (Eq. 5; Table 1) based on a Q₁₀ temperature coefficient set to 2.2 (Bastviken 2009).

The model was used to calculate phytoplankton CO₂ uptake (*Up*; Eq. 6, Table 1) as the difference between the volumetric rate of production and respiration, with respiration rate defined as a proportion of production (*ra*; Table 2). Cole et al. (2002) suggest that 85-90% of gross primary production (GPP) may be directly respired back through autotrophic respiration, while a lower range of 54–88% has been recorded previously (Quay et al. 1986). In this model we assigned a wider range of *ra* value of 50 – 99%. We estimated phytoplankton production using an empirical relationship based on chlorophyll a (*Chl*) concentration (del Giorgio and Peters 1993).

The high productivity of Lake Okaro contributes to its DOC pool. A strong tryptophan-like component of the DOC (Hartland, pers. comm. 2016) coincides with the presence of a deep chlorophyll maximum during stratification (Simmonds et al. 2015), which suggests that contributions from dead phytoplankton cells may dominate labile DOC in this layer. The CO₂ production in this layer may therefore be significant. We assumed that 3% of phytoplankton production was exuded as refractory DOC (Hanson

et al. 2004). We used a sensitivity analysis to test a wide range of mineralisation rates of extracellular labile DOC (r_{Chl} in Table 2; 0.1 to 70%). The extracellular DOC (Chl_{lab} ; Eq. 7, Table 1) was assumed to decay rapidly in the surface layer to produce CO_2 (Connolly and Coffin 1995). The remaining phytoplankton carbon production was considered to enter the particulate organic carbon pool and be subject to sedimentation to the bottom sediment.

Contributions of allochthonous DOC to the CO₂ pool were assumed to be small. As residence time in Lake Okaro is 1.5 years, and considering that most DOC is from phytoplankton decay, we assumed that concentrations of DOC from the two streams were similar to those in the lake (\sim 5.2 mg L⁻¹: Bay of Plenty Regional Council, unpublished data), with a photo-oxidation rate (px) of 0.005 d⁻¹ (Hanson et al. 2004).

Gas exchange at the air-water interface was calculated using a boundary layer equation (Eq. 8; Table 1) with the gas exchange coefficient (k_{gas}) estimated as a function of wind speed (Cole and Caraco 1998). Average monthly wind speed from the National Climate Database was used as input to the model. Due to uncertainty in gas transfer coefficient estimates (Raymond et al. 2013; Dugan et al. 2016), a correction factor ($corr_k$; Table 2) was included in the gas exchange calculation for the purpose of sensitivity analysis. It was assumed that the atmospheric CO₂ and CH₄ concentrations were constant at 393.5 ppmv and 1.76 ppmv, respectively.

As a soft water lake with low alkalinity of \sim 0.6 meq L⁻¹ (data not shown), inputs of inorganic carbon from carbonate weathering in Lake Okaro would likely be minor (Marcé et al. 2015). McColl (1972) noted that changes in pH in the surface waters are mostly regulated by phytoplankton productivity, while the reduced pH in the hypolimnion is related directly to CO_2 production from decomposition in the tropholytic

zone. Therefore, we assumed that changes in the whole-lake bicarbonate (HCO₃) mass occurred mostly from CO₂ dissociation into the carbonate system.

An optimization routine was used to derive model parameter values which had ranges assigned based on those in the literature (Table 2). The optim() function with "L-BFGS-B" method (R Core Team 2015) was applied. The routine minimized the residual errors between the modelled whole-lake gas storage (Eqns 3 and 4, Table 1) and the calculated gas storage from field observations (Eq. 2, Table 1).

An interpolation using a trapezoidal rule (R package Hmisc: Harrell 2015) was performed to calculate the annual whole-lake CO₂ and CH₄ mass, both for the observed and modeled mass balance. The modeled annual whole-lake value was calculated by summing the processes rates involved in the whole-lake carbon storage dynamics, i.e., diffusion across the sediment-water interface, aerobic CH₄ oxidation, photosynthetic CO₂ uptake, phytoplankton decay, DOC oxidation, and atmospheric flux over the period of observation. Finally, the total GHG emissions estimates (kg CO₂-eq m⁻² y⁻¹) were calculated as the sum of the modelled CO₂ and CH₄ atmospheric diffusive fluxes (Eq. 5) by multiplying each gas flux by its global warming potential (GWP) over a 100-year period (i.e., CO₂ = 1, CH₄ = 34; Myhre et al. 2013).

Data collection

To support model computation, we collected water column samples on a monthly basis from September 2013 to September 2014 from the deepest station of the lake (Figure 1). In mid-winter (June, southern hemisphere) additional samples were collected on two occasions to target changes in the water column profile due to the onset of a period of lake mixing each year. Water samples were collected at the subsurface (0.5 m) and to a depth of 16 m with a vertical resolution of 2 m using a 10-L Schindler-Patalas water

sampler. Triplicate subsamples of 60 mL were taken from the sampler using a syringe, for dissolved CO₂ and CH₄ analysis. The aliquot in the syringe was transferred into a 45 mL glass bottle with excess water overflowing the bottle. The bottle was capped with a polypropylene screw cap with a silicon septum, leaving no bubbles in the bottle, and stored in the dark and cold (4 °C) until analysis. Water column profiles of temperature and dissolved oxygen were taken on each sampling occasion using a conductivity-temperature-depth (CTD) profiler (Sea-Bird Electronics 19 plus) with an additional mounted dissolved oxygen (DO, Sea-Bird Electronics) sensor. We observed thermal stratification and the depth of the thermocline from the CTD temperature profiles and used the rLakeAnalyzer package (Winslow et al. 2015) to analyse metrics of mixing and stratification (see theoretical basis given by Read et al. 2011). Chlorophyll *a* samples were collected as an integrated sample of the euphotic depth, from the immediate subsurface to a depth of approximately 2.4 times the average Secchi disk depth of ~2.5 m. Chlorophyll *a* samples were analyzed by spectrophotometer following standard methods involving acetone extraction and correction for phaeophytin (APHA 1998).

Dissolved CO₂ and CH₄ concentrations for gas samples from the field were determined in the laboratory within 24 h of sampling using the headspace equilibration method (Striegl and Michmerhuizen 1998). Briefly, approximately 20 mL of pure N₂ was injected into a 45 mL glass bottle through the silicon septum, replacing aliquots in the bottle to create headspace using a second needle to drain water. The remaining water volume in the bottle was kept at 28 mL. Additional nitrogen gas (5 mL) was added to the bottles to over-pressurize the headspace in the bottle. Equilibration was performed by shaking the bottles vigorously for 3 minutes. Bottles were kept upside down to avoid gas leakage from the septa. Subsamples of 5 mL of equilibrated gas were transferred from the headspace into 3.7 mL pre-vacuumed Labco Exetainer® vials. The CO₂

concentration in the headspace was measured by an infra-red gas analyser (IRGA, LI-COR® LI-6262). A gas chromatograph equipped with a flame ionization detector (GC-FID, Varian CP 3800) was used to determine CH₄ concentrations in the headspace. The concentrations of dissolved gas in the water samples were then calculated based on the solubility function for CO₂ and CH₄ following Weiss and Price (1980) and Wiesenburg and Guinasso (1979), respectively.

Uncertainty analysis

Spatial heterogeneity may affect diffusive greenhouse estimates from lakes (Schilder et al. 2013, Beaulieu et al. 2016). As our data collection was focused on a central lake site in small, circular Lake Okaro, we therefore performed a bootstrapping procedure to determine the uncertainty in field measurements and provide some indication of spatial variability. For each measured data point, we created 1000 psuedodata points generated from a random normal distribution of the 15% deviation in the original measurement. We then fitted the mass balance model to the 1000 pseudo-data points to provide uncertainty estimates of model parameters. Through this bootstrapping procedure we were able to calculate the average and the 95% confidence interval of both, field measurement and model estimates. The performance of the model was tested by calculating the mean absolute error and Pearson correlation coefficient (*R*) between simulation and observation.

Results

Thermal stratification, dissolved oxygen and chlorophyll a concentration

Between September 2013 and September 2014, Lake Okaro started to stratify around September 2013 (spring, southern hemisphere) and was completely mixed (i.e., isothermal) on 25 June 2014 (winter, Figure 3A). The thermocline was shallowest on 10

October 2013 (3.1 m) and gradually increased in depth thereafter, until turnover. Surface water temperature increased steadily from 11.6 °C (September) to a maximum of 23.4 °C (December, early summer) and continued to decrease gradually thereafter. Dissolved oxygen (DO) concentrations in the epilimnion were occasionally strongly supersaturated during stratification (126% of saturation), while the concentration in the hypolimnion decreased rapidly with the onset of stratification (Figure 3B). The hypolimnion became anoxic (< 1 mg L⁻¹) from early summer. As the lake underwent seasonal mixing (end of May), the DO concentration was relatively low (< 5 mg L⁻¹) throughout water column before recovering to be close to saturation (~8 mg L⁻¹) after a period of sustained mixing (late Jun-Jul). Chl concentrations peaked (>50 μ g L⁻¹) in August – October (spring) at the onset of thermal stratification but were consistently lower (mean 14.6 μ g L⁻¹) during the latter period of stratification from December to June.

CO₂ and CH₄ concentrations and storage dynamics

During the stratification period (September to May), concentrations of dissolved CO₂ and CH₄ in the epilimnion were mostly low, $<1-15.4~\mu mol~L^{-1}$ and $0.1-1.4~\mu mol~L^{-1}$, respectively (Figures 3C and 3D). In contrast, concentrations in the hypolimnion were relatively high, reaching >230 and >240 $\mu mol~L^{-1}$ for CO₂ and CH₄, respectively. While hypolimnetic CO₂ concentrations started to increase immediately following the depletion of DO in the hypolimnion (late September), CH₄ concentrations increased later (mid-January), as the bottom waters became anoxic. The highest bottom-water (16-m depth) concentrations of CO₂ and CH₄ occurred just before turnover, in May. On 27 May 2014, epilimnetic CO₂ and CH₄ concentrations were high (43.4 – 52.7 μ mol L⁻¹ and 6.5 – 19.9 μ mol L⁻¹, respectively). Just after the onset of turnover (end of May to early June), concentrations of CO₂ and CH₄ were relatively uniform through the water

column at $93.0 - 102.4 \mu \text{mol L}^{-1}$ and $15.6 - 20.0 \mu \text{mol L}^{-1}$, respectively.

The dynamics of CO₂ and CH₄ concentrations reflect the temporal variations of storage in the lake (Figures 4A and 4B). Whole-lake CO₂ mass was calculated to be low (63.0 mmol CO₂ m⁻²) when the lake started to stratify in September (Figure 4A). The mass initially increased gradually, at rate of 3.4 mmol CO₂ m⁻² d⁻¹ as stratification developed in September (Figure 4C), and had increased abruptly to 63.1 mmol CO₂ m⁻² d⁻¹ at turnover (late May to early June), yielding whole-lake CO₂ storage of 1412 mmol CO₂ m⁻² (Figure 4A). Soon after, in late June, CO₂ decreased rapidly from the lake, at a rate of 50.3 mmol m⁻² d⁻¹ (Figure 4C). This CO₂ loss continued until the end of August, but at a decreased rate of 1.6 mmol m⁻² d⁻¹ when the whole-lake CO₂ mass was 53.0 mmol CO₂ m⁻² (Figure 4A). Accumulation of CH₄ in bottom waters started later than CO₂, in late December, at a rate ca. 2.6 mmol m⁻² d⁻¹ (Figure 4D), after the lake has been stratified for some time (Figure 4B) and then increased abruptly, at a rate of 19.0 mmol CH₄ d⁻¹ (Figure 4D) through mid-April. By this time the whole-lake CH₄ storage was 723 mmol m⁻² (Figure 4B). As the thermocline deepened nearing the end of summer stratification (late May), the accumulated CH₄ in the lake started to decline. The rate of loss of CH₄ was calculated to be 7.4 mmol m⁻² d⁻¹ and was maximal at 18.8 mmol m⁻² d⁻¹ during prior to or during turnover (early June, Figure 4D). During the remaining period of mixing (June – August), however, CH₄ loss was reduced to <0.1 mmol m⁻² d⁻¹, leaving only 2.3 mmol CH₄ m⁻² in the lake by the end of August.

Mass balance model output and CO2 and CH4 internal cycling

An optimization procedure (see Methods) was used to calibrate parameters in the mass balance model. Performance statistics for the mass balance model comparisons against measured CO₂ and CH₄ storages were satisfactory. The mean absolute error was low for

comparisons of ΔCO_2 and ΔCH_4 in the model against field observations (3.10 mmol CO_2 m⁻² d⁻¹ and 2.66 mmol CH_4 m⁻² d⁻¹) and the correlation coefficient (R) was 0.933 and 0.831 for ΔCO_2 and ΔCH_4 , respectively.

Most of the time the lake acted as a sink for CO₂, except during the onset of seasonal mixing in May to early June when atmospheric emissions were highly variable (Figure 5A). In the early stages of thermal stratification (September), there was a large influx of atmospheric CO₂, calculated to be 15.12 mmol CO₂ m⁻² d⁻¹. In early June, however, emissions were calculated to be 60.85 mmol CO₂ m⁻² d⁻¹ to the atmosphere. Over the entire period of the study, we calculated that Lake Okaro took up CO₂ at a rate of 452.4 mmol m⁻² y⁻¹ from the atmosphere.

The annual water column CO₂ utilization by phytoplankton photosynthetic uptake was calculated to be up to 4950.4 mmol m⁻² y⁻¹ (Table 3), with the highest uptake during spring (August to October) of \sim 30 mmol CO₂ m⁻² d⁻¹ (Figure 5B). Decay of phytoplankton was estimated to release 2974.4 mmol CO₂ m⁻² y⁻¹ to the lake (Table 3). The calculated CO₂ sediment diffusion flux to the water column was 1430.2 mmol m⁻² y⁻¹ and the DOC oxidation was calculated to add 175.0 mmol m⁻² y⁻¹ of CO₂ to the water column CO₂ pool.

CH₄ was calculated to be continuously lost diffusively from the lake water surface to the atmosphere at low levels (~1 mmol CH₄ m⁻² d⁻¹), except during the onset of mixing when emissions peaked at 9.95 mmol CH₄ m⁻² d⁻¹ (Figure 5C). Annually, this atmospheric flux was calculated to be 553.4 mmol CH₄ m⁻² y⁻¹ (Table 3). Sediment diffusion flux of CH₄ to the lake water column was calculated to be 946.0 mmol m⁻² y⁻¹ (Table 3). The accumulated CH₄ in the whole water column was estimated to undergo oxidation at a rate of 508.5 mmol m⁻² y⁻¹. The highest CH₄ oxidation rate occurred during mixing in late May, at 10.8 mmol m⁻² d⁻¹ (Figure 5D). CH₄ is assumed to be

oxidized to CO_2 in the water column, equating to 508.5 mmol CO_2 m⁻² y⁻¹ annually (Table 3).

Greenhouse gas emissions

Based on our mass balance calculation, Lake Okaro was considered at annual time scales to be a sink of CO₂ from the atmosphere and a source of CH₄ (Table 3).

Translated into the global warming potential of these emissions, the lake is a sink for 5.6 x 10⁻³ kg CO₂-eq m⁻² y⁻¹ from the atmosphere and releases 226.0 x 10⁻³ kg CO₂-eq m⁻² y⁻¹. The net result of these GHG fluxes was estimated to be 222.6 x 10⁻³ kg CO₂-eq m⁻² y⁻¹.

Discussion

CO₂ and CH₄ dynamics and the mass balance model

The dynamics of dissolved CO₂ and CH₄ for the lake in this study were strongly seasonally dependent. The observed pattern is similar to observations in other monomictic and dimictic lakes, with at least four distinct periods (Striegl and Michmerhuizen 1998; Riera et al. 1999; López Bellido et al. 2011). These periods relate to (1) dissolved gas (CO₂ and CH₄) accumulation in the hypolimnion as stratification begins in spring, (2) high concentrations of dissolved gas in the hypolimnion and low concentrations in the surface layer during summer stratification, (3) high dissolved gas concentrations in the surface layer, as well as high atmospheric fluxes at the onset of breakdown of stratification in autumn, and (4) rapid loss of dissolved gases in the surface water during the period of turnover. Progressive deepening of the thermocline gradually entrained CO₂ and CH₄ from the hypolimnion into surface waters prior to complete water column mixing. When fully mixed, there was complete entrainment of

the highly enriched (with CO₂ and CH₄) remaining hypolimnetic waters with the surface waters, yielding major gas releases to the atmosphere (Fig. 3C, 3D, 5A and 5C).

While direct measurements of carbon emissions have been improved in recent times, processes involved in the genesis, accumulation and release of greenhouse gases are still difficult to monitor in natural aquatic ecosystems. The mass balance model used in this study was fed by monitoring data and utilized parameters from the literature to address gaps in monitoring fluxes, allowing a comparison of the relative magnitude of the major processes that contribute to the fluxes of CO₂ and CH₄. The model performed well in simulating changes in the whole-lake mass storage of CO₂ and CH₄ (Figure 4C and 4D). The model can be used to differentiate contributions of different processes that regulate loss and gain of CO₂ and CH₄ in the water column (see Figure 2), but uncertainties exist with model parameters (Figure 4C and 4D) that, based on sensitivity analysis, could be prioritised for further study.

Based on our model, we calculated that a large proportion of CO₂ was taken up by phytoplankton, consistent with the lake trophic state classification of eutrophic (Table 3). This high CO₂ demand was balanced by inputs from the sediment, atmospheric exchange, phytoplankton decomposition, DOC oxidation and CH₄ oxidation. Assuming that phytoplankton uptake results in loss of carbon to the sediments as the resulting organic material sediments out (and after subtracting water column losses of CO₂ to decomposition), the annual sedimentary carbon addition is estimated to be 2087.4 mmol C m⁻². This estimate is slightly lower our earlier prediction (2320 mmol C m⁻² y⁻¹) calculated using a first order diagenesis model for labile organic matter in lake sediment cores (Santoso et al. 2017). Some factors influencing lake CO₂ storages were not accounted for in the model, e.g., stream organic and inorganic carbon (Weyhenmeyer et al. 2015; Maberly et al. 2013; Wilkinson et al. 2016). These potential

sources of error are likely to have made only a small contribution to the lake CO₂ pool (see Figure 4C).

Through the model, we estimated that, on an annual basis, ~63% of CH₄ produced by the sediment (as diffusive flux) is oxidized to CO₂, resulting in reduced fluxes of CH₄ to the atmosphere (Table 3). There is considerable debate about the extent that oxidation reduces CH₄ emissions (Schubert et al. 2012; Fernández et al. 2014) and this uncertainty complicates global estimates of GHG emission potential from lakes, particularly from those that are eutrophic. In our model the fraction of CH₄ released to the atmosphere exceeded the residual component of sediment production and water column oxidation (Table 3). The CH₄ ebullition not accounted for may have influenced both CH₄ and CO₂ storages in the lake. As CH₄ bubbles rise following release from the sediment, a large proportion of the CH₄ dissolves into surrounding water medium (McGinnis et al. 2006; Ostrovsky et al. 2008). Some may be oxidized to CO₂, especially in the presence of dissolved oxygen that is usually present at levels close to saturation in the upper mixed stratum (Schmid et al. 2017), while a large proportion (>80%) may escape oxidation and be released to the atmosphere (DelSontro et al. 2010; Martinez and Anderson 2013). West et al. (2016) estimated that sediments in shallow areas (<6 m depth) of productive lakes may produce CH₄ at a rate of ca. 24 mmol m⁻² d⁻¹. This methane production coincided with high probability of emission from ebullition. Using these estimates, we can conservatively calculate CH₄ emission via ebullition. Assuming that the shallow zone of Lake Okaro is 6% of the total surface area (see Methods), methane production in the shallow zone is 1.44 mmol CH₄ m⁻² d⁻¹ which equates to an annual production of 525.6 mmol CH₄ m⁻² y⁻¹. Accordingly, 108.0 mmol m⁻² y⁻¹ may have dissolved in the water column, balancing the CH₄ pool in the lake, while 417.6 mmol m⁻² y⁻¹ may have escaped to the atmosphere through ebullition. However, as our

model did not incorporate CH₄ ebullition due to its stochastic nature (Varadharajan and Hemond 2012), estimates of carbon emissions from Lake Okaro in this study should be viewed as conservative and apply only to diffusive fluxes of CO₂ and CH₄. Thus, as fluxes may vary spatially and temporally (e.g. Schilder et al. 2013; Beaulieu et al. 2016), better quantification and resolution of both diffusive and ebullitive fluxes are an important prerequisite for high precission estimates of carbon emissions from lakes.

GHG emissions from eutrophic lakes

Moss et al. (2011) noted that eutrophication may promote climate change. Productive eutrophic lakes, indicated by elevated levels of chlorophyll *a*, may store more carbon in the sediment than less productive lakes (Santoso et al. 2017; Sobek 2009).

Remineralization of the deposited carbon produces dissolved organic and inorganic carbon as well as CO₂ and CH₄, which can re-enter the water column and may ultimately contribute to GHG emissions (Klump et al. 2009).

We compared our estimates of GHG emissions from Lake Okaro, computed from the mass balance model, with lakes across a range of trophic state gradients. Emissions from diffusive fluxes of CO₂ and CH₄ (converted to kg CO₂-eq m⁻² y⁻¹), together with chlorophyll a concentrations (n = 316), are available for lakes across the globe from Downing et al. (2018). Trophic states of these lakes were defined based on chlorophyll a concentration following Nürnberg (1996) where \leq 3.5 μ g L⁻¹ is classified as oligotrophic, > 3.5 μ g L⁻¹ and \leq 9 μ g L⁻¹ is mesotrophic, and > 9 μ g L⁻¹ is eutrophic. It was shown that chlorophyll a concentration was positively correlated with GHG emissions, although the statistical fit was modest (Figure 6). Results from Lake Okaro and global lakes imply that eutrophic lakes tend to have higher greenhouse gas emissions than oligo- and mesotrophic lakes. Our results reinforce recent syntheses that

conclude that productivity is an important factor in scaling the magnitude of GHG emissions from lakes (DelSontro et al. 2018).

The ratios of CO₂ and CH₄ in the total annual GHG emissions are varied across lake trophic gradient (Figure 7, also see density distribution of the data in Supplementary Information). Most eutrophic lakes release more CH₄ than CO₂, with ratios ca. 0.8. Oligo- and mesotrophic lakes tend to release more CO₂ than CH₄ with ratios of 2.4 and 1.2, respectively. Lake Okaro, on the other hand, released 100% CH₄, and received only a very small input of CO₂ (Figure 7). These results emphasize that CH₄ is the major contributor of emissions from eutrophic lakes. This study focused on diffusive fluxes only but emissions due to ebullition (mainly CH₄) from lakes and impoundments can be substantial and are often larger in small and shallow systems (Holgerson and Raymond 2016; Martinez and Anderson 2013) as well as in productive systems (DelSontro et al. 2018). High algal biomass in eutrophic systems may enrich the sediments with autochthonous carbon which is more labile and readily converted to CH₄ than allochthonous carbon, which tends to be more recalcitrant (West et al. 2012).

As eutrophication of inland waters is ongoing, and despite concerted efforts to thwart it (Smith et al. 2014), inland waters will probably be one of the main actors in enhancing global warming through CH₄ emissions (see Beaulieu et al. 2019). Knowing that inland waters are dominated by small lakes with surface areas of <1 km² (Downing et al. 2006) and are prone to eutrophication, as well as having high areal rates of organic carbon burial (Downing et al. 2008), lake restoration may be used as an important mitigation strategy to combat global warming.

Conclusions

This study demonstrated that CO₂ and CH₄ emissions in monomictic eutrophic lakes are

driven by an interplay between seasonal thermal stratification and biological processes (i.e., organic carbon mineralization, phytoplankton uptake and decay, and CH₄ oxidation). Through a mass balance model, we showed that productive eutrophic Lake Okaro has a net release of CH₄ that exceeds its uptake of CO₂ from the atmosphere, both in magnitude and more so in terms of GHG warming potential. This analysis may be viewed as conservative and applies only to diffusive fluxes, however it underscores that CH₄ emissions from eutrophic lakes are strongly influenced by eutrophication. As eutrophic lakes emit more greenhouse gases, we suggest that lake restoration should be considered as part of a mitigation strategy to combat global warming.

Acknowledgement

We thank the Bay of Plenty Regional Council (BOPRC) for technical support, and Paul Scholes of BOPRC for data provision. Special thanks to Joseph Butterworth (BOPRC) for field work assistance and Janine Ryburn (The University of Waikato) for helpful support in laboratory analysis.

Disclosure statement

The author(s) declare(s) that there is no conflict of interest regarding the publication of this article.

Author contributions

ABS: Conceptualization, Methodology, Software, Validation, Formal Analysis, Investigation, Writing – Original Draft, Writing – Reviewing and Editing,

DPH: Funding acquisition, Supervision, Resources, Writing – Reviewing and Editing.

LAS: Supervision, Resources, Reviewing and Editing.

ISO: Writing – Reviewing and Editing.

CHH: Supervision.

Funding

This study was part of a PhD project supported by the NZ-ASEAN Scholarship.

Research funding for this study was also supported through the Lake Biodiversity

Restoration programme funded by the Ministry of Business, Innovation & Employment

(UOWX0505) and the Bay of Plenty Chair in Lake Restoration at the University of

Waikato.

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Figure 1. Location map and the bathymetry of Lake Okaro, North Island, New Zealand. Sampling site is marked by a white circle.

Figure 2. Conceptual model of CO₂ and CH₄ annual budget. Ebullition (dashed box) is not accounted for in the model simulation (see text for details).

Figure 3. Measured spatiotemporal variations of temperature (A), concentrations of dissolved oxygen (B), CO₂ (C) and CH₄ (D) in Lake Okaro over a one-year period (September 2013 to September 2014). Solid black lines indicate calculated thermocline depth, dashed vertical lines indicate date of sampling.

Figure 4. Whole-lake storage of CO_2 (A) and CH_4 (B) in Lake Okaro and changes in the whole-lake mass storage (\triangle mass) of CO_2 (C) and CH_4 (D) over a one-year period (September 2013 to September 2014). Dark areas around solid mean lines indicate lower and upper 95% confidence interval of field observations (green) and model outputs (blue). Light grey colours indicate range of bootstrapped measurement data. Negative and positive values in \triangle mass indicate losses and gains of gases, respectively.

Figure 5. Rates of atmospheric flux of CO₂ (A), phytoplankton CO₂ uptake (B), and CH₄ atmospheric flux (C) and water column CH₄ oxidation (D) for Lake Okaro calculated from the mass balance model. Dark areas around solid mean lines indicate lower and upper 95% confidence interval. Light blue colours represent range of

bootstrapped model outputs. Positive and negative values represent loss and gain of gas, respectively.

Figure 6. Greenhouse gas emission from impoundments across trophic levels. Solid line is the log-relationship between trophic levels (indicated by chlorophyll a concentration) and total emission, $R^2 = 0.16$, p<0.001, with 0.95 confidence interval (light grey). Trophic levels (Trop) range: oligotrophic (oligo), mesotrophic (meso), eutrophic (eu).

Figure 7. Proportion of annual greenhouse emission rates from Lake Okaro and global lakes and impoundments across trophic state gradients. Impoundments emissions data (Downing et al. 2018) is depicted by their statistical mode. Light grey indicates CH₄, dark grey indicates CO₂ emission. Greenhouse emission in Lake Okaro is dominated by CH₄ (light grey) and the CO₂ sink is small, -5.1 x 10⁻³ kg C-CO₂ eq m⁻² y⁻¹.

Table 1. Equations for processes in the CO₂ and CH₄ mass balance estimation. See Table 2 for definitions of symbols used in equations.

Equation	Eq. No.	Units	Process definition	Source
$C_{strg} = \sum_{i=1}^{z} \frac{C_{(i)} dV_{(i)}}{A_{(i)}}$	(1)	mmol m ⁻²	Whole-lake water column gas storage	1
$\Delta C_{strg} = \frac{dC_{strg}}{dt}$	(2)	mmol m ⁻² d ⁻¹	Observed gas mass change over the sampling period	
$\frac{dCH_{4_{strg}}}{dt} = J_{CH4} - Rw_{oxi} - F_{CH4}$	(3)	mmol m ⁻² d ⁻¹	CH ₄ mass balance model over the sampling period	
$\frac{dCO_{2strg}}{dt} = J_{CO2} + Rw_{oxi} + Chl_{lab} + DOC_{px} - Up - F_{CO2}$	(4)	mmol m ⁻² d ⁻¹	CO ₂ mass balance model over the sampling period	
$Rw_{oxi} = \sum_{i=1}^{z} \frac{K_{CH4-O2}CH_{4(i)}O_{2}dV_{(i)}}{O_{2sol}A_{(i)}} e^{Q10(temp-20)/10}$	(5)	mmol m ⁻² d ⁻¹	Whole lake CH ₄ oxidation flux with Arrhenius temperature (temp) adjustment	2
$Up = \frac{10.3Chl^{1.9}V_{surf}}{12.01A}(1-ra)$	(6)	mmol m ⁻² d ⁻¹	Phytoplankton CO ₂ uptake	3
$Chl_{lab} = Chl r_{chl}$	(7)	mmol m ⁻² d ⁻¹	Phytoplankton decomposition	
$F_{gas} = corr_k k_{gas} (C_{surf} - C_{atm})$	(8)	mmol m ⁻² d ⁻¹	Atmospheric gas flux	

Sources: ¹Striegl and Michmerhuizen (1998), ²Lopes et al. (2011), ³delGiorgio and Peters (1993).

Table 2. Parameters used for CO_2 and CH_4 mass balance calculations.

Symbol	Description	Units	Value	Source
$C_{(i)}$	Concentration of the dissolved gas in layer (i)	μМ		
$V_{(i)}$	Lake volume in layer (i)	m^3		
V_{surf}	Lake volume of 0 to 2 m surface layer	m^3		
$A_{(i)}$	Lake surface area in layer (i)	m^2		
i	Water column layer			
n	Number of samples			
t	Date of sampling			
z	Lake depth	m		
Chl	Chlorophyll a concentration	$\mu g \; L^{\text{-}1}$		
k_{gas}	Gas exchange coefficient	m d ⁻¹		4
Csurf	Dissolved gas concentration at the airwater interface	μΜ		
C_{atm}	Gas solubility concentration calculated from atmospheric molar fraction concentration	μΜ		5, 6
DOC	Dissolved organic carbon concentration	μM		
px	Photo oxidation of DOC	d ⁻¹	0.005	7
O_2	Dissolved oxygen concentration	μM		
O_{2sol}	Oxygen solubility	μM		8
J_{CO2}	Diffusive flux of CO ₂ from the sediment to water column	mmol m ⁻² d ⁻¹	3.81	9
$J_{ extit{CH4}}$	Diffusive flux of CH ₄ from the sediment to water column	mmol m ⁻² d ⁻¹	2.54	9
Parameter	rs used for optimization routine			
$corr_k$	Correction factor for gas exchange coefficient		0.1 - 6.0	10
KCH_{4-O2}	Aerobic CH ₄ oxidation constant	d ⁻¹	0.001 - 1.0	11
ra	Proportion of phytoplankton production that is respired	d ⁻¹	0.5 - 0.99	7, 12, 13
r_{Chl}	Mineralization rate of dead phytoplankton	d ⁻¹	0.001 - 0.7	14

Sources: ⁴Cole and Caraco (1998), ⁵Weiss and Price (1980), ⁶Wiesenburg and Guinasso (1979),

⁷Hanson et al. (2004), ⁸Weiss (1970), ⁹Santoso et al. (2017), ¹¹Dugan et al. (2016), ¹¹Bastviken (2009)

and reference therein, ¹²Cole et al. (2002), ¹³Quay et al. (1986), ¹⁴Connolly and Coffin (1991)

Table 3. Estimated processes contributed to CO₂ and CH₄ whole-lake storage based on the mass balance model output over a one-year period in Lake Okaro. Values are given in mmol m⁻² y⁻¹. Numbers in parentheses indicate lower and upper 95% of confidence interval. Negative and positive values represent losses and gains of gases, respectively. NA is not available due to the nature of the process.

CO_2	CH ₄	
1430.2	946.0	
(1214.2 - 1646.2)	(801.8 - 1090.2)	
508.5	-508.5	
(385.6 - 631.4)	(385.6 - 631.4)	
-4950.4	NA	
(-4593.65307.2)		
2974.4	NA	
(2551.2 - 3397.5)		
175.0	NA	
(148.9 - 201.2)		
471.3	-545.2	
(391.2 - 551.4)	(-461.8673.6)	
	1430.2 (1214.2 - 1646.2) 508.5 (385.6 - 631.4) -4950.4 (-4593.65307.2) 2974.4 (2551.2 - 3397.5) 175.0 (148.9 - 201.2) 471.3	

Figure 1.

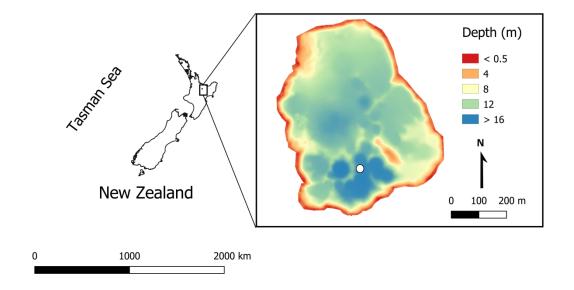


Figure 2.

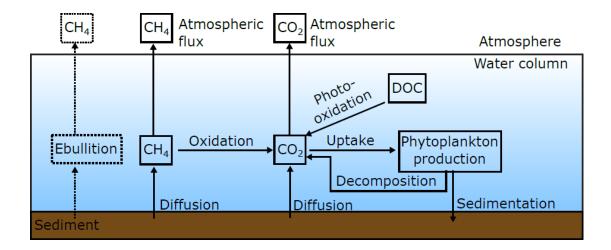


Figure 3.

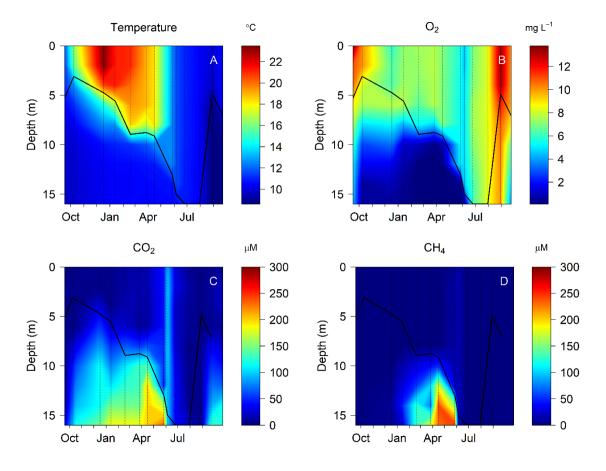


Figure 4.

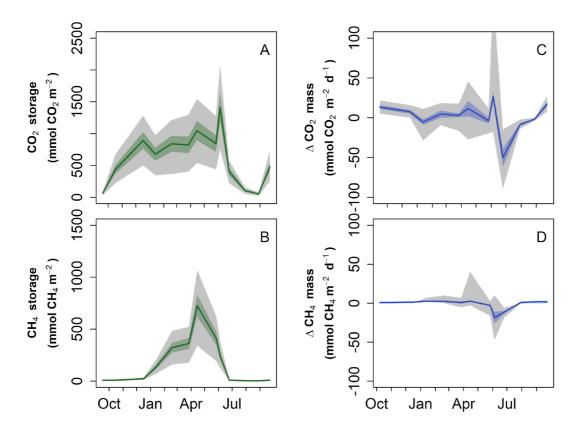


Figure 5.

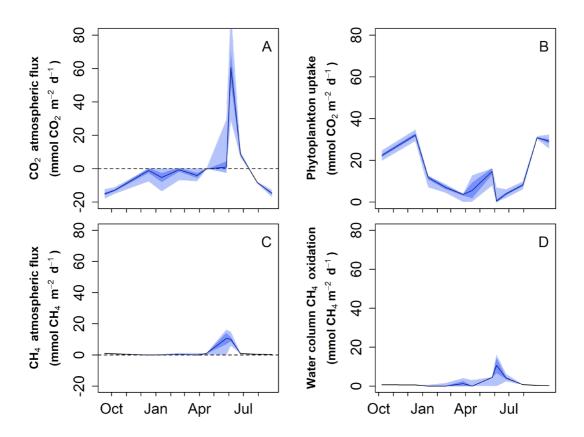


Figure 6.

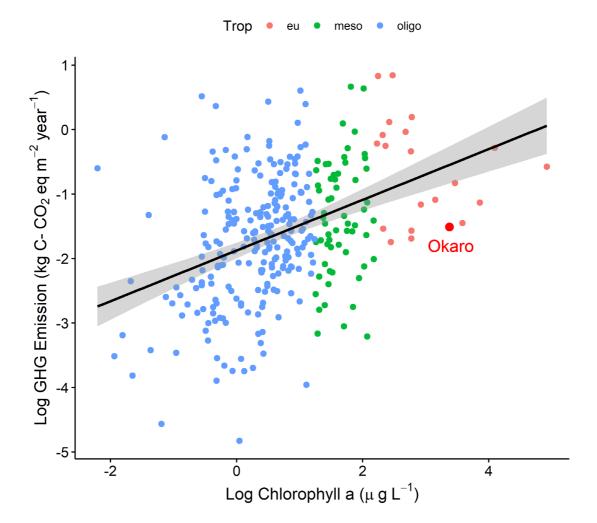


Figure 7.

