Carbon dioxide stimulates lake primary production

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Gross primary production (GPP) is a fundamental ecosystem process that sequesters carbon dioxide (CO₂) and forms the resource base for higher trophic levels. Still, the relative contribution of different controls on GPP at the whole-ecosystem scale is far from resolved. Here we show, by manipulating CO₂ concentrations in large-scale experimental pond ecosystems, that CO₂ availability is a key driver of whole-ecosystem GPP. This result suggests we need to reformulate past conceptual models describing controls of lake ecosystem productivity and include our findings when developing models used to predict future lake ecosystem responses to environmental change.

The rate at which primary producers fix inorganic carbon controls the supply of organic matter to food webs and influences the biogeochemistry of aquatic ecosystems1,2. The role of inorganic nutrients and light in controlling GPP is indisputable and has been extensively studied1–3. Data also suggest that CO₂ availability can constrain GPP. Small scale experimental studies, and comparative studies of lakes, have shown that elevated CO₂ concentration promotes phytoplankton biomass and GPP6,7. Similarly, CO₂ concentration can limit phytoplankton growth in marine ecosystems8. However, aquatic ecosystems are comprised of both pelagic and benthic habitats, and recent studies emphasize the importance to include both habitats to understand whole-ecosystem productivity and function4,5. Presently, the role of CO₂ availability for whole-ecosystem GPP, especially in shallow lakes where both pelagic and benthic GPP may contribute significantly to whole-ecosystem GPP, has never been tested9. Improved knowledge of the control of GPP is fundamental for understanding ecosystem function and impacts of environmental change.

Here we carried out a large-scale experiment to test the role of CO₂ availability for whole-ecosystem GPP. We used a novel approach where part of the CO₂ that accumulated under ice cover over winter was released to the atmosphere by manipulation of the spring ice cover. The aim was to decrease CO₂ concentrations in the water column while having a limited effect on light, temperature, and other key environmental factors. In spring, the ice cover was decreased twice, by first removing 10% and then 50%, to impose a gradual but drastic difference in CO₂ concentrations in the treatment ponds. Whole-ecosystem GPP was estimated using dissolved oxygen time series data, and potential abiotic and biotic drivers of GPP were monitored over the course of the experiment.

Results and Discussion

The GPP and CO₂ concentration in control ponds were relatively high under ice compared to after ice break-up in spring (Figs 1, 2). The 10% ice-removal treatment did not change GPP or decrease CO₂ concentration compared to control ponds (Fig. 2 and Table 1). In contrast, the 50% ice-removal treatment decreased both GPP and CO₂ compared to control ponds (Fig. 2 and Table 1). Finally, after ice break-up, there were no differences in GPP and CO₂ between control and treatment ponds (Figs 1, 2 and Table 1). There was a positive correlation between GPP and CO₂ concentrations in both control and treatment ponds (Figs 1 and Table 1). These results show that the CO₂ concentration was a key controlling factor for GPP in the ponds.

No other abiotic or biotic factors explain the patterns in GPP over time and between treatments. There were no differences in nutrient concentrations (NO₃⁻, PO₄³⁻ and NH₄⁺, Supplementary Table S1), photosynthetically active radiation (PAR) or water temperature between control and treatment ponds (Fig. 2 and Table 1). Interestingly, GPP in control and treatment ponds was negatively correlated to both PAR and water temperature (Table 1). The consumer biomass (zooplankton and zoobenthos) did not differ between control and treatment ponds (Supplementary Table S1), suggesting that top-down effects on primary producers, if any, were similar and therefore should have no effect on the main patterns in GPP observed between treatment and control. As expected, pH of the water increased with decreasing CO₂ concentrations in the ponds (Supplementary Table S1).

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The change in pH and carbonate system is part of the change associated with changes in CO₂ saturation and cannot be easily separated from each other. Yet, previous studies on the effect of CO₂ on GPP have not found any effect of pH on GPP and attributed GPP responses to CO₂ availability and not to pH per se⁶,⁷,¹⁰.

This is the first experimental study on the role of CO₂ in controlling whole-ecosystem GPP. Although CO₂ has a key role as substrate for photosynthetic enzymes¹¹, the CO₂ supply is generally regarded as sufficient to meet the primary producer’s demands and that other factors are limiting photosynthetic rates. However, CO₂ may often be at suboptimal levels for maximum photosynthetic efficiency¹. Previous work has shown positive effects of CO₂ on pelagic GPP⁶,⁷, but here we show that this effect applies also at the whole-ecosystem scale. This is important since lake habitats are not isolated units and both benthic and pelagic habitats are important for understanding whole-lake ecosystem food web dynamics and productivity¹⁴.
CO₂ limitation effect on lake productivity is likely a general characteristic that needs to be taken into account in food web models for lakes. The CO₂ concentrations of lakes vary largely across systems but also within systems, both spatially and temporally12, implying that CO₂ availability can help to explain GPP at multiple scales. Given that lake CO₂ supersaturation is common, including in abundant northern lakes13, we suggest that CO₂ stimulation of GPP is a common but often overlooked phenomenon.

Although not explicitly studied in this experiment, these results also add important insight to the role of terrestrial organic matter for lake GPP. Export of dissolved organic matter from land is regarded to have two main effects on GPP in recipient lake ecosystems: (i) stimulating GPP by supplying nutrients14 and (ii) repressing GPP by supplying light absorbing substances4. Terrestrial organic matter is also mineralized in lakes, resulting in supersaturation of CO₂15,16, and our results suggest this will stimulate GPP. It is likely that CO₂ stimulates GPP in ecosystems with low to moderate terrestrial organic matter inputs, up to a point when light becomes suboptimal, after which further increase in terrestrial organic matter inputs will have an overall negative effect on GPP due to light limitation. Future research should assess the role of CO₂ for GPP across various type of lakes.

The results from this study provide fundamental knowledge of the effects of CO₂ dynamics on biomass production in lake food webs. More broadly, these results have major implications for a greater understanding of the effects of climate change on lake productivity, as CO₂ dynamics in lakes are controlled by climate-dependent factors such as terrestrial carbon export, hydrologic residence time, metabolic process rates, mixing regimes and ice conditions12,17,18. We stress the need for future research efforts, where the effects of CO₂ are incorporated in both experiments and models of lake ecosystem function to generate patterns at the whole-ecosystem scale.

### Methods

#### Study site.

The study was performed in the Umeå University Experimental Ecosystem Facility (EXEF) in northern Sweden (63°48’N, 20°14’E). The pond is divided into 20 enclosures (12.5 * 7.3 m, average depth 1.48 m) by thick non-permeable tarpsaulins and each pond has a 7.3 m long natural shoreline and a bottom substrate of soft sediment. Each pond has a natural food web including basal producers (algae and bacteria), consumers (insect larvae and zooplankton) and a reproducing top-consumer population (nine-spine Sticklebacks), where benthic GPP constitutes approximately 50% of whole-ecosystem GPP19. For further details see16,20. In this study we used eight enclosures: we manipulated the CO₂ concentration in four enclosures by ice cover removal (treatment), and four enclosures served as controls. Ice formed naturally on the ponds in mid-November 2015, and the experiment was carried out in March-April 2016. The ice cover removal was performed twice in 2016 on the treatment ponds; 10% removal from 28th March to 8th April and 50% from 9th to 19th April, i.e. ten days for each treatment. The four control ponds experienced a natural and gradual ice melting. The treatment and control

### Table 1. Statistical Analysis results. Control ponds, treatment ponds, number of replicates, degrees of freedom, 10% ice-cover removing treatment, 50% ice-cover removing treatment, Ice-free period, and linear correlation are abbreviated as C, T, n, df, 10%, 50%, Ice-free and r, respectively.

| Test | Ponds | Treatment | Variables | Statistics | p-values | n | df |
|------|-------|-----------|-----------|------------|----------|---|----|
| T-test |       |           |           |            |          |   |    |
| C vs. T | 10% | GPP | −0.77 | 0.49 | 4 | 3 |
| C vs. T | 10% | CO₂ | 0.49 | 0.56 | 4 | 3 |
| C vs. T | 10% | PAR | 0.23 | 0.82 | 4 | 3 |
| C vs. T | 10% | Temp. | 2.6 | 0.07 | 4 | 3 |
| C vs. T | 50% | GPP | 12.21 | <0.01 | 4 | 3 |
| C vs. T | 50% | CO₂ | 3.64 | <0.05 | 4 | 3 |
| C vs. T | 50% | PAR | −1.30 | 0.28 | 4 | 3 |
| C vs. T | 50% | Temp. | 0.97 | 0.40 | 4 | 3 |
| C vs. T | Ice-free | GPP | 1.18 | 0.32 | 4 | 3 |
| C vs. T | Ice-free | CO₂ | 0.13 | 0.90 | 4 | 3 |
| C vs. T | Ice-free | PAR | 0.92 | 0.42 | 4 | 3 |
| C vs. T | Ice-free | Temp. | 0.80 | 0.47 | 4 | 3 |

| Pearson’s r | | r-value |
|-------------|--|----------|
| C | GPP vs. CO₂ | 0.73 | <0.01 |
| T | GPP vs. CO₂ | 0.93 | <0.001 |
| C | GPP vs. PAR | −0.86 | <0.001 |
| T | GPP vs. PAR | −0.69 | <0.05 |
| C | GPP vs. Temp. | −0.80 | <0.01 |
| T | GPP vs. Temp. | −0.73 | <0.01 |

| ANOVA | | F-value |
|-------|--|----------|
| C vs. T | GPP * time | 6.92 | <0.05 | 2, 12 |
| C vs. T | CO₂ * time | 3.95 | <0.05 | 2, 12 |
| C vs. T | PAR * time | 2.16 | 0.15 | 2, 12 |
| C vs. T | Temp. * time | 0.3 | 0.74 | 2, 12 |
ponds became ice-free between 20th of April and 1st of May. The ice-removal treatment was executed by making vertical bores in the ice cover with an ice drill. Pieces of ice were cut out in-between the bores by a large-ice saw and then pushed in under the remaining ice.

**Data collection.** CO₂ concentrations were estimated every second day by using a headspace equilibration technique and analysis by gas chromatograph (Perkin Elmer Clarus 500). At the same time, nutrients (NO₃⁻, PO₄³⁻, and NH₄⁺) were sampled by filtering water through burnt (550 °C, 6 h) 0.45 μm GF/F filters to 50 ml Falcon tubes and stored in the freezer until analyzed with photometric flow injection analysis (FIA) methods.

Dissolved oxygen (DO) and water temperature were measured during the experiment period at ten-minute intervals by logging sensors (MiniDOT, PME, Vista, CA, USA) which were deployed at 0.5 m below the water surface in the center of each pond. Photosynthetic available radiation (PAR) was measured at ten-minute intervals by light sensors (SQ-110, Apogee USA) deployed at 0.8 m below the water surface in the center of each pond. Temperature and PAR data were converted to daily means for each enclosure based on 144 measures (6 per hour × 24 hours) and then a mean for all days within each treatment period during the experiment and finally we got a mean for each group (control and treatment) that contains four enclosures. Wind speed was recorded every ten minutes by a climate station next to the pond.

**Whole-ecosystem gross primary production (GPP) estimates.** From the oxygen sensor data whole-ecosystem GPP, integrating GPP in both benthic and pelagic habitats in these non-stratified ponds, was calculated with inverse modeling and Bayesian parameter estimation using a similar parameter estimation approach as for diel DO in streams, but modified for pond ecosystems “equation (1)”: 

\[
mO_2 = mO_{2-1} + \frac{GPP_{zmix} \cdot PAR_{zmix}}{PAR_{zmix}} + \frac{ER_{zmix}}{zmix} \cdot \Delta t + K_i (O_{at, i} - mO_{2-1}) \Delta t
\]

where mOi is modeled DO at time i (g O₂ m⁻³) given parameter estimates of GPP and ecosystem respiration (ER; g O₂ m⁻³ d⁻¹). Because changes in O₂ are a function of GPP, ER, and gas exchange, we calculated daily air-water O₂ fluxes based on O₂ saturation and the temperature-corrected gas exchange velocity for O₂ (K₁, d⁻¹). K₁ was estimated from K600 derived from wind speed. The emission flux of O₂ (K_i (O_{at, i} - mO_{2-1}) \Delta t) was corrected for changes in pond area open to the atmosphere (0–100% with changing ice cover) and zmix (mean of mixing depth; m), where zmix varied daily according to changes in water column depth depending on ice cover thickness. The metabolism model used a “random walk” metropolis algorithm and Markov Chain Monte Carlo (MCMC) sampling from the “metrop” function in the “mcmc” package of the statistical program R to find the best fit between measured and modelled O₂ data given model estimates of GPP and ER. Each parameter estimate was derived from 10000 model iterations after removing an initial 1000 iterations of “burn-in” from parameter starting values. We checked for convergence of parameter estimates and removed days with negative GPP and with poor fits between measured and modeled O₂ before assessing the response of GPP to changes in CO₂.

**Invertebrate sampling.** Zooplankton were sampled by a zooplankton net (diameter 20 cm, 100 μm mesh size) drawn vertically trough whole water column and preserved in Lugol’s solution. Zoobenthos were sampled with a net (30 cm wide, 1 mm mesh size), drawn at the bottom substrate for a distance of 1 m, and preserved in ethanol. Zooplankton and zoobenthos lengths were measured to obtain dry biomass using length-weight regressions.

**Statistical analyses.** Statistics (SPSS 20 and R v3.2.3) are based on individual pond means of measured response variables. Results were tested for time effects between treatments periods by using repeated measures ANOVA, for differences between the control and treatment within treatment periods by using standard t-tests, and for correlations between selected variables by using Pearson correlation coefficient.

**Data availability.** The datasets generated during and/or analyzed during the current study are available from the corresponding author on request.

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Author Contributions

M.H., J.K. and P.B. designed the study with contribution from J.A. M.H. and M.J.A. performed the field and lab work. E.H. wrote the metabolism codes. M.H. analyzed the data with contribution from E.H. and P.B. M.H. wrote the manuscript and all co-authors revised the manuscript.

Additional Information

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