

1 **Macrophyte-hydrodynamic interactions mediate stratification and dissolved oxygen**
2 **dynamics in ponds**

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23 **ABSTRACT**

24 Small waterbodies are sensitive to stressors such as eutrophication and heatwaves;
25 however, interactions between macrophytes and hydrodynamics may mediate the effects of
26 compounding stressors. Leveraging an ecosystem experiment and hydrodynamic model, we
27 evaluated how macrophyte biomass, thermal structure, and dissolved oxygen (DO) responded to
28 the interaction of episodic nutrient loading and periods of high temperatures in two temperate
29 ponds. In one pond we experimentally added pulses of nutrients, simulating storm-driven loading
30 (the other pond served as an unmanipulated reference). Following the first nutrient pulse both
31 ponds experienced a 5-day period of high surface water temperatures. Macrophytes in the
32 nutrient addition pond began to senesce mid-summer due to phytoplankton shading from the
33 nutrient addition and heat stress while macrophytes in the reference pond followed expected
34 seasonal patterns, senescing in early autumn. Field observations and model results indicate that
35 macrophytes structured the thermal environment through vertical attenuation of turbulent kinetic
36 energy and light. Macrophytes reduced the vertical extent of water column warming during the
37 heat event by 0.25-0.5 m and maintained cooler bottom temperatures (up to 2.5 °C cooler)
38 throughout the summer, suggesting that macrophytes may buffer small waterbodies from
39 heatwaves. Seasonal patterns in DO saturation also followed trends in macrophyte biomass;
40 however, during the heat event, DO saturation fell sharply (-22.4 to 50.4 %) in both ponds and
41 remained depressed through the remainder of the summer. This experiment and modeling
42 exercise demonstrated that macrophyte influence on turbulent flows and light are pivotal in
43 mediating how small waterbodies respond to compounding stressors.

44

45 INTRODUCTION

46 Small lakes and ponds are the most abundant lentic ecosystems worldwide and sites of
47 critical importance for freshwater biodiversity and biogeochemical cycles (Scheffer et al. 2006;
48 Cole et al. 2007; Messenger et al. 2016). Small waterbodies are active components of regional-
49 and global-scale carbon cycling, both as sites of burial and storage (Wilkinson et al. 2018; Taylor
50 et al. 2019) and as sources of atmospheric carbon dioxide and methane (Holgerson and Raymond
51 2016; Peacock et al. 2021). Small lakes and ponds enhance local and regional biodiversity, due
52 in part to abundant macrophytes (Van Geest et al. 2003). Robust macrophyte communities
53 support greater diversity of other taxa by providing food resources and habitat structure
54 (Williams et al. 2004; Scheffer et al. 2006). In addition to these biotic interactions, macrophytes
55 play a pivotal role in the ecosystem structure and function of small waterbodies. Interactions
56 between hydrodynamics and the physical structure of macrophyte beds influence turbulent
57 mixing, water column thermal structure, and sediment stability as well as whole-lake
58 productivity and nutrient cycling (Carpenter and Lodge 1986; Scheffer et al. 1993; Licci et al.
59 2019). Despite the prevalence and importance of small, vegetated lakes and ponds, these
60 ecosystems remain understudied (Downing 2010; Biggs et al. 2017), particularly with regards to
61 the role of macrophyte-hydrodynamic interactions in mediating ecosystem response to stressors.

62 The characteristic morphometry of small waterbodies makes these ecosystems more
63 susceptible to stressors. For example, limited water volume and shallow mean depth reduce both
64 dilution potential and thermal inertia making small waterbodies sensitive to nutrient loading and
65 intense heatwaves (Biggs et al. 2017; Woolway et al. 2021a; Polazzo et al. 2022). As a result,
66 eutrophication and aquatic heatwaves may act as compounding stressors for small lakes and
67 ponds. Aquatic heatwaves are periods of sustained high surface water temperatures relative to

68 local and seasonal baseline conditions (Hobday et al. 2016; Tassone et al. 2021; Woolway et al.
69 2021a). The frequency, duration, and intensity of aquatic heatwaves have increased over the past
70 century as a result of anthropogenic climate change, a trend which is anticipated to continue,
71 even under low green-house-gas-emissions scenarios (Oliver et al. 2018; Woolway et al. 2021a,
72 Woolway et al. 2022). Aquatic heatwaves may lower dissolved oxygen (DO) concentrations
73 directly through reduced gas solubility in warmer waters and indirectly through increased
74 ecosystem respiration (Tassone et al. 2021). Low DO concentrations and thermal stress have
75 been linked to coral bleaching, declines in kelp forests, and mass seagrass mortality in marine
76 systems (Wernberg et al. 2016; Hughes et al. 2017; Strydom et al. 2020). Although heatwaves
77 are well-studied in coastal and marine systems, research on the effects of heatwaves in lakes and
78 ponds is relatively nascent (Woolway et al. 2021a; Woolway et al. 2022). As such, it remains
79 unclear how aquatic heatwaves affect the structure and function of lentic ecosystems, especially
80 small, vegetated waterbodies.

81 Though aquatic heatwaves are projected to be more severe in small, shallow waterbodies
82 (Woolway et al. 2021a), if present, macrophytes may buffer small lakes and ponds from extreme
83 temperatures by restricting vertical heat transfer. Canopy-forming macrophyte beds attenuate
84 incoming solar radiation and dissipate wind-driven turbulence, thus limiting the depth of heat
85 transport and creating a shallow mixed surface layer separated from cooler bottom waters (Herb
86 and Stefan 2004; Andersen et al. 2017a; Sand-Jensen et al. 2019). In short, macrophytes induce
87 thermal stratification in small waterbodies, which may isolate bottom waters from rising surface
88 temperatures during an aquatic heatwave. Stratification is also expected to restrict dissolved gas
89 transport and may produce a vertically heterogenous chemical environment within macrophyte
90 beds (Andersen et al. 2017b; Vilas et al. 2017). As a result, macrophyte structure could also

91 influence how DO responds to aquatic heatwaves. Rising surface water temperatures during a
92 heatwave are expected to lower DO concentrations; however, in maintaining cooler bottom water
93 temperatures, macrophyte-induced stratification may dampen the effects of a heatwave on gas
94 solubility and DO depletion. Altogether, macrophytes play a key role in the thermal structure and
95 DO environment of small waterbodies, potentially mediating the effect of temporary, external
96 stressors on these ecosystems' structure and function. However, an explicit evaluation of how
97 macrophyte-hydrodynamic interactions may alter ecosystem response to stressors is lacking.
98 This is especially true for lakes and ponds as research on macrophyte-flow interactions has
99 largely focused on coastal and lotic systems (e.g., Fonseca et al. 2019; Tseng and Tinoco 2022).

100 While macrophytes influence ecosystem structure and function in small waterbodies,
101 environmental conditions also affect macrophytes. The abundance and growth of macrophytes in
102 small waterbodies is driven by temperature and the availability of light and nutrients (Carpenter
103 and Lodge 1986; Phillips et al. 2016), making them sensitive to environmental stressors such as
104 nutrient loading and heatwaves (Wu et al. 2021). Warmer water temperatures can enhance
105 macrophyte growth (Olesen and Madsen 2001, Bertani et al. 2016; Hansson et al. 2020);
106 however, high temperatures may also induce heat stress, impeding reproduction (Li et al. 2017)
107 or prompting senescence (Hao et al. 2018). Similarly, nutrient loading may have either a positive
108 or negative effect on macrophyte growth. Although macrophytes can benefit from enhanced
109 nutrient availability through foliar uptake (DeMarte and Hartman 1974; Twilley et al. 1977),
110 nutrient loading can also increase algal biomass and subsequently reduce light availability
111 through shading (Scheffer et al. 1993; Short et al. 1995). The effect of shading from
112 phytoplankton-associated turbidity is expected to vary across macrophyte growth forms, as
113 submersed species are more vulnerable to light limitation than emergent or floating-leaf species

114 (Szabo et al. 2010). While the individual responses of macrophytes to nutrient enrichment and
115 heat stress have been established, it remains unclear how macrophytes respond to simultaneous,
116 compounding stressors and what the consequences are for the thermal and chemical structure of
117 small lentic ecosystems.

118 We experimentally evaluated the interacting effects of episodic external nutrient loading
119 and an extreme heat event on macrophyte biomass and the subsequent changes in the
120 physicochemical environment in two shallow, vegetated ponds. Specifically, we monitored
121 spatiotemporal variation in water column thermal structure, DO profiles, and macrophyte
122 biomass and canopy height in each pond and over the course of a growing season. We
123 experimentally pulsed one of the ponds with nutrients to simulate a typical storm-driven external
124 loading event, while the other pond served as a reference. Both ponds experienced a mid-summer
125 heat event following the first experimental nutrient pulse. In the absence of severe environmental
126 stressors, the phenology of the macrophyte communities in these temperate ponds follows a
127 pattern of spring emergence from over-wintering rhizomes, growth throughout the summer, and
128 then senescence of the leaves and stems in early autumn. However, we hypothesized that the
129 combined stress of the nutrient addition and heat event in the experimentally eutrophied pond
130 would induce early macrophyte senescence due to increased shading from phytoplankton and
131 heat stress, thus reducing macrophyte biomass and subsequently altering the temperature and DO
132 gradients in the water column. To further test the mechanistic relationship between macrophytes
133 and pond hydrodynamics, we applied a hydrodynamic model for one-dimensional heat transport
134 in vegetated ponds and simulated pond response to a heat event across a range of macrophyte
135 biomass and canopy height.

136 Leveraging our ecosystem experiment and hydrodynamic model, our first research
137 objective was to quantify macrophyte response to the compounding stressors of nutrient
138 enrichment and an extreme heat event as well as the subsequent response of pond thermal
139 structure and DO concentrations. We then explained the observed responses by focusing on the
140 reciprocal relationship between macrophytes and pond hydrodynamics. Specifically, we asked
141 (Q1) How do macrophyte biomass and canopy height influence pond thermal structure, and does
142 macrophyte presence control vertical heat transfer during a heat event? (Q2) How does
143 spatiotemporal variation in DO relate to macrophyte biomass and canopy height, and does
144 macrophyte presence alter the DO response to a heat event? We anticipated that greater
145 macrophyte biomass and canopy height would be associated with stronger thermal stratification
146 and greater differences in DO between surface and bottom waters. We also hypothesized that
147 macrophyte presence would maintain cool water habitat and stable DO concentrations during a
148 heat event. Through the combination of our experiment, observations, and numerical modeling,
149 we found that macrophyte structure can be a source of spatial and seasonal variation in the
150 thermal and chemical environments of shallow waterbodies.

151

152 **METHODS**

153 *Study Site and Field Measurements*

154 We studied how macrophyte structure influences spatiotemporal variation in both the
155 thermal and DO environments of small waterbodies using two ponds at the Iowa State University
156 Horticulture Research Station (42.110005, -93.580454) during the summer of 2020. The ponds
157 are rectangular, relatively small (surface area 400 m² at the water line; Figure 1), and shallow
158 (maximum depth 2 m and mean depth 0.8 m), with simple, bowl-shaped bathymetries. These

159 waterbodies fit morphometric and functional definitions of ponds (Richardson et al. 2022). Water
160 balance is driven solely from precipitation and evaporation as there are no surface inflows or
161 outflows and the ponds are lined with bentonite to prevent groundwater interactions. As there is
162 effectively no watershed for these ponds, we effectively controlled all external inputs of
163 nutrients. The ponds were initially filled with water from the research station's irrigation
164 reservoir to seed natural phytoplankton and zooplankton communities. Early in the spring the
165 ponds were stocked with yellow perch (*Perca flavescens*) and bluegill (*Lepomis macrochirus*) at
166 roughly 20 kg ha⁻¹ for both species, which is within the range of densities for temperate
167 waterbodies (Carlander 1977; Schneider 1999). The macrophyte community in both ponds was
168 composed of longleaf pond weed (*Potamogeton nodosus*; hereafter “floating-leaf”) and leafy
169 pondweed (*Potamogeton foliosus*; hereafter “submerged”).

170 External nutrient loading typical of large storm events was simulated by adding nitrogen
171 and phosphorus to one of the ponds (hereafter “nutrient addition pond”) twice over the course of
172 the summer. Nitrogen was added as ammonium nitrate (NH₄NO₃), and phosphorus was added as
173 sodium phosphate monobasic dihydrate ((NaH₂PO₄•(H₂O)₂) in a ratio of 24N:1P. The first
174 nutrient addition (day of year, DOY, 176) raised the mass of phosphorus in the pond by 3
175 percent, and the second addition (DOY 211) raised the mass by 5 percent. These increases are
176 consistent with the expected nutrient loading from a large storm event in an agricultural
177 landscape (Vanni et al. 2001; Lüring et al. 2018). Nutrients were not added to the other pond
178 (hereafter “reference pond”). To assess and compare the effects of the nutrient additions on algal
179 biomass, we measured chlorophyll-*a* concentrations daily at the deep site of each pond (Figure 1)
180 using the Total Algae sensor as a part of the YSI ProDSS multiparameter sonde (Yellow Springs
181 Instruments, Yellow Springs, Ohio USA). The sensor logged continuously and was slowly

182 lowered through the water column. Surface concentrations of chlorophyll-*a* were averaged over
183 0-0.3 m depth in the water column. To minimize drift, the sensor was calibrated regularly against
184 laboratory standards.

185 To monitor thermal structure and dynamics across the ponds, we deployed vertical strings
186 of high frequency temperature sensors (HOBO 8K Pendant Temperature Data Logger) at three
187 sites, aligned from the deep (2 m) to shallow (1.5 m) ends of each pond (Figure 1). Temperature
188 loggers were distributed throughout the water column every 0.25 m until 1.5 m deep and then
189 every 0.5 to the bottom. Water temperature was logged every 30 minutes from May to August
190 (DOY 143-233). We also took high frequency measurements of surface water DO concentration
191 and saturation at the deep site of both ponds (miniDOT Clear Logger; logging every 30 minutes).
192 In order to quantify the three-dimensional DO environment, manual profiles of DO
193 concentration, DO saturation, and water temperature were taken weekly (13 sampling events
194 total) on a grid of 18 sites across each study pond using a kayak (Figure 1). Measurements were
195 taken in the late morning, between 10:00AM and 12:00PM. A continuous profile of DO was
196 logged at each site as the YSI ProDSS multiparameter sonde was slowly lowered through the
197 water column.

198 We quantified variation in macrophyte structure over space and time by monitoring
199 community composition, canopy height, and biomass from May to August (DOY 143-233)
200 across the study ponds. Species presence-absence and canopy height were measured weekly at
201 the three sites established from the deep to shallow ends of the ponds (Figure 1). Canopy height
202 was measured as the distance from the sediment-water interface to the top of the canopy. Water
203 depth was also measured at these points so that canopy height could be expressed as a proportion
204 of the water column. Macrophyte dry biomass was sampled every other week (6 sampling events

205 total) on a grid of 18 sites across each pond where manual profiles were also taken (Figure 1).
206 Profiles were always taken before biomass was sampled. To collect a biomass sample, a two-side
207 rake was lowered to the bottom and used to remove submersed and floating-leaf macrophytes
208 from an approximately 0.4 m² area (Mikulyuk et al., 2011). All the above-sediment tissue was
209 collected from the rake to determine dry biomass. Samples were dried at 60°C to a constant mass
210 before weighing.

211

212 *Hydrodynamic Modeling*

213 In order to explore the mechanisms by which changes in macrophyte biomass and canopy
214 height influence pond thermal structure, we applied a one-dimensional hydrodynamic model for
215 vegetated waterbodies. Our model builds upon the one-dimensional, integral energy model by
216 Herb and Stefan (2004) for heat transport in lakes with submersed macrophytes. Macrophytes are
217 incorporated into the model through two key parameters: a light extinction coefficient due to
218 macrophyte biomass shading and a term for turbulent kinetic energy dissipation by macrophytes
219 based on their surface area and drag. Our contributions to the original model include adding
220 convective overturn to address density instabilities over the diurnal cycles, a dynamic
221 macrophyte growth and senescence boundary condition over time, dynamic light attenuation
222 coefficient for water based on daily Secchi depth field data, and coding the model for use
223 through open-source software (R; see Appendix S1 for more details on model algorithms and
224 additional citations).

225 For boundary conditions, the model requires an initial water temperature profile,
226 timeseries of meteorological drivers (air temperature, relative humidity, air pressure, wind speed,
227 and short-wave radiation), and waterbody hypsography (depth-area relationship). We determined

228 the hypsography of the study ponds through manual calculations based off depth measurements
229 from the profiles taken at 18 sites across each pond. We obtained the necessary meteorological
230 data at an hourly timestep from a weather station located approximately 550 meters from the
231 study ponds and maintained by the Iowa State University Soil Moisture Network (Iowa
232 Environmental Mesonet, ISU Soil Moisture Network 2021). The initial temperature profile was
233 supplied from a single reading from the temperature chain located at the deep site of the
234 reference pond (Figure 1). Macrophyte data were incorporated into the model through weekly
235 measurements of canopy height at the deep site of the reference pond as well as biomass density
236 measurements taken every two weeks. Biomass density standardizes macrophyte biomass to the
237 volume of water sampled (g m^{-3}). We used the average biomass density from the two sampling
238 sites closest to the deep site temperature sensors, in the southern region of the pond.

239 The model was run on a 1-hour timestep from late May-August (DOY 143-241; 99 days
240 total). The observed, high frequency water temperature profiles from the deep site of the
241 reference pond were used to calibrate 9 model parameters (Table S1) and assess model fit
242 (quantified as root mean square error, RMSE). Once the model was sufficiently calibrated to the
243 full duration of the observed field data, we manipulated the input macrophyte data to simulate
244 the effect of different macrophyte biomass and canopy dynamics on thermal structure (Table S2).
245 Specifically, we tested a scenario with no macrophytes present in the pond compared to
246 macrophyte biomass and canopy heights that were measured in the reference pond. All modeling
247 analyses were completed in R version 4.1.2 (R Core Team 2021) using the gotmtools
248 (<https://github.com/aemon-j/gotmtools>), rLakeAnalyzer (Winslow et al. 2019), lubridate (Garrett
249 and Wickham 2011), and tidyverse (Wickham et al. 2019) packages.

250

251 *Statistical Methods*

252 Our delineation of extreme heat events in the experimental ponds was informed by the
253 definition developed for marine heatwaves (Hobday et al. 2016) and applied more recently to
254 estuaries (Tassone et al. 2021) and lakes (Woolway et al. 2021a). Following this definition, an
255 aquatic heatwave occurs when daily mean surface water temperatures exceed the seasonal 90th
256 percentile threshold for at least 5 days, without dropping below that threshold for 2 or more days.
257 Since we did not have long-term data on summer water temperatures in the experimental ponds,
258 we could not use this formal definition. Therefore, we elected to use temperature monitoring data
259 for the study period only, but a more stringent definition of a heat event threshold (i.e., 95th
260 percentile) for the same extended period (5 days). To determine the 95th percentile surface water
261 temperature for each pond, we used temperature loggers placed at 0 and 0.25 m at the middle at
262 deep sites. For each site, we averaged the readings from these surface loggers, calculated a daily
263 mean, and then averaged values from the middle and deep sites. These steps produced a daily
264 mean surface water temperature for each pond, which we used to calculate the 95th percentile
265 threshold. We then identified any days in which the daily mean surface water temperature
266 exceeded this threshold. A period with at least five days above this threshold was classified as a
267 sustained aquatic heat event (hereafter, “heat event”).

268 To visualize pond thermal structure from our high frequency water temperature data, we
269 used a linear interpolation to estimate temperature for depths at which we either did not place
270 sensors (1.75m deep for both ponds) or for depths at which sensors ceased to function during the
271 summer (1m at the middle and deep sites of the nutrient addition pond, 0.25-0.75m at the
272 shallow site of the nutrient addition pond, and 0.25m at the middle site of the reference pond).
273 The interpolated values were used solely for visualization. Only empirical measurements of

274 water temperature were used in the model and to calculate indices of pond thermal structure (i.e.,
275 thermocline depth and Lake Number). Lake Number is a ratio that compares the internal stability
276 of the water column to external wind stress across the waterbody surface, so it can be used to
277 determine the strength of stratification and likely response to observed wind stress (Robertson et
278 al. 1990). Thermocline depth and Lake Number were calculated using the rLakeAnalyzer
279 package. (Winslow et al. 2019). We defined surface water temperature as the average from the 0
280 and 0.25m sensors. We used the deepest sensors (2m at the middle and deep sites, 1.5m at the
281 shallow sites) for measurements of bottom water temperature.

282 In order to compare macrophyte biomass between the study ponds, we calculated the
283 mean macrophyte biomass across 18 sampling sites for each sampling event. However, we
284 weighted this average so that sampling sites around the perimeter of the pond (n=14) had the
285 same weight as sites in the interior of the pond (n=4; Figure 1) so that the mean value would
286 not be biased by water depth. Weighted, pond-averaged macrophyte biomass values were used to
287 plot seasonal changes and explore the relationship between biomass and water column thermal
288 stability. We also standardized macrophyte canopy height as a percent of the water column depth
289 for each sampling site and event to aid comparisons between shallow and deeper sites.

290 To analyze the manual DO profiles, we first determined that there were no instances of
291 mid-depth DO maximum for any sites or sampling events. The vertical distribution of DO
292 saturation was either uniform, highest in the surface waters, or highest in the bottom waters
293 depending on the site and time of year (Figure S1). Therefore, we focused on DO dynamics in
294 the surface and bottom layers of ponds by averaging values from 0-0.25 m and 0.25 m from the
295 sediment-water interface respectively. Surface and bottom water DO values were used to explore
296 temporal variation, calculate the difference between surface and bottom DO, and quantify

297 variation over space and time. We used coefficients of variation to quantify temporal and spatial
298 variation in surface and bottom water DO saturation. Specifically, spatial variation was defined
299 as the coefficient of variation for DO for each sampling day within each pond, so variation was
300 due to differences among the 18 sampling sites (Figure 1), standardized by the pond-mean for
301 that day. Temporal variation was calculated based on the coefficient of variation in DO for each
302 sampling site, so variation was due to differences in DO over time, standardized by the summer
303 mean DO saturation for that site.

304 A generalized additive model (GAM) was used to smooth the daily time series of
305 chlorophyll concentrations and weekly time series of DO saturation to visually highlight
306 differences in trends over time between the nutrient addition and reference pond. All statistical
307 analyses were completed in R version 4.1.2 (R Core Team 2021) using the mgcv (Wood 2017),
308 rLakeAnalyzer (Winslow et al. 2019), lubridate (Garrett and Wickham 2011), and tidyverse
309 (Wickham et al. 2019) packages.

310

311 **RESULTS**

312 *Environmental Stressors and Macrophyte Response*

313 Chlorophyll-*a* concentrations were variable in both ponds throughout the experiment, but
314 there was a clear increase in chlorophyll-*a* following the sequential additions of nutrients in the
315 experimental pond (Figure 2A). Early in the season, chlorophyll-*a* concentrations were low in
316 the reference pond, likely due to high *Daphnia sp.* biomass (Butts et al., *in prep*). Following this
317 early clear-water phase and prior to the experimental nutrient additions, chlorophyll-*a*
318 concentrations were similarly low in both ponds (DOY 160). Nutrient additions on DOY 176 and
319 211 resulted in chlorophyll-*a* concentrations 2-3 times higher in the nutrient amended pond

320 compared to the reference pond. Heightened chlorophyll-*a* concentrations in the nutrient addition
321 pond are indicative of higher algal biomass and thus phytoplankton-associated turbidity, likely
322 resulting in reduced light availability in the nutrient addition pond.

323 Both ponds experienced a heat event from DOY 185-190 (Figure 2B). During this period,
324 mean daily surface water temperatures exceeded the seasonal 95th percentile in each pond
325 (greater than 29.8 °C in the nutrient addition pond and 30.6 °C in the reference pond). Maximum
326 daily surface water temperatures ranged from 34.5-37.1 °C in the nutrient addition pond and
327 33.7-38.6 °C in the reference pond during the heat event. Although both ponds had brief periods
328 of high surface water temperatures at other points throughout the summer, DOY 185-190 was the
329 only sustained period of high surface water temperatures (≥ 5 days). The heat event began 9 days
330 after the first nutrient pulse, creating multiple stressors (i.e., reduced light availability and
331 prolonged heat stress) for macrophytes in the nutrient addition pond.

332 Surface DO saturation was variable over the study period, at times differing between the
333 two study ponds (Figure 2C). Early in the summer, both ponds were supersaturated with DO, and
334 mean surface saturation was similar between the ponds before the initial nutrient addition. In the
335 reference pond, DO remained fairly stable until the heat event (DOY 185-190), after which DO
336 saturation declined and generally remained below saturation for the remainder of the study
337 period. In contrast, mean surface DO saturation in the nutrient addition pond declined
338 immediately following the first nutrient addition (DOY 176) and remained below saturation for
339 the rest of the summer.

340 Spatiotemporal patterns in macrophyte biomass differed between the ponds and
341 macrophyte growth forms (Figure 2D, 3). The floating-leaf species was largely limited to a ring
342 around the shallower edges of the ponds where the floating leaves could easily reach the surface.

343 The submerged species was present across a range of depths but reached the highest biomass in
344 the deeper, central region of each pond. In the reference pond, macrophytes followed expected
345 phenology with increasing biomass through late summer and then senescence of both growth
346 forms beginning in early autumn (Figure 2D, 3). The heat event did not affect macrophyte
347 biomass or assembly in the reference pond. However, in the nutrient addition pond, submerged
348 macrophytes growing in 1.75 to 2 m of water began to senesce following the initial nutrient
349 addition and heat event such that no plants were present in the deepest areas of the nutrient
350 addition pond 9 days (DOY 199) after the heat event and the middle site of the pond 23 days
351 (DOY 213) after the heat event ended (Figure 3). Early senescence of submersed macrophytes in
352 the nutrient addition pond resulted in rapid declines in both biomass (Figure 2D) and canopy
353 height (Figure 4). Both submersed and floating-leaf plants persisted in shallow water around the
354 edge of the pond through the end of the summer.

355

356 *Spatiotemporal Variation in Pond Thermal Structure*

357 Both ponds experienced intermittent thermal stratification during the study period and
358 similar spatial patterns of stratification from shallow to deep ends of the pond (Figure 4). Early
359 in the summer, both ponds were cool and isothermal. As surface waters began to warm, the
360 middle and deep sites of both ponds began to stratify while the shallow sites remained fairly
361 well-mixed. At the center of the ponds, a stable thermocline between 1-1.5 m depth was
362 observed in the reference pond between DOY 155-161, and almost constantly between DOY
363 168-210 (Figure S2A), while the thermocline in the nutrient addition pond was more intermittent
364 (DOY 157-159, 170-174, and 178-192; Figure S2B). When present, pond thermoclines cooled
365 and deepened slightly (0.05-0.1 m) during the night, indicating nighttime convective cooling and

366 mixing in the surface layer. During periods of thermal stratification, differences between surface
367 and bottom water temperatures ranged from 4.3 to 11.5 °C (mean 8.1°C) in the reference pond
368 and 4.5 to 10.2 °C (mean 7.3 °C) in the nutrient addition pond.

369 Similarities in pond thermal structure occurred during times of similar macrophyte
370 coverage, either early or late in the season (Figure 4). For example, both ponds experienced a
371 strong mixing event in late summer (DOY 216-220), driven by cooler air temperatures and
372 higher wind speeds (Figure 5) and facilitated by declining macrophyte structure (Figure 4).
373 Similar meteorological conditions occurred earlier in the summer and resulted in some water
374 cooling (e.g., DOY 164, 175); however, the ponds did not mix as fully during these events, likely
375 due to the macrophyte structure present in both ponds in early summer (Figure 4). Similarities in
376 pond thermal structure in relation to extrinsic factors (i.e., wind speed) are further demonstrated
377 by similar lake number values for both ponds over the study period (Figure S3).

378 Although overall seasonal patterns in thermal structure were similar between the ponds,
379 there were brief dissimilarities, related to differences in the timing of macrophyte senescence
380 (Figure 4). There was a prolonged period of stable thermal stratification from DOY 180-210 in
381 the reference pond, with only brief instability between DOY 198-199. A stable thermocline set-
382 up in the nutrient addition pond around the same time (DOY 178); however, stratification began
383 to break down approximately 18 days earlier (DOY 192) than in the reference pond (Figure S2).
384 The timing of destratification in the nutrient addition pond coincides with declines in macrophyte
385 biomass (Figure 3) and canopy height (Figure 4) in the center of the pond. Destratification in the
386 reference pond also followed patterns of biomass senescence and loss of canopy height, although
387 this did not occur until later in the summer in this pond. In both ponds, there was strong temporal
388 coherence between the timing of destratification and macrophyte senescence, driven by the

389 compounding stress of eutrophication and a heat event in the nutrient addition pond and typical
390 seasonal senescence in the reference pond. The declines in canopy cover and biomass were
391 associated with increased thermal mixing and warmer bottom water temperatures, especially in
392 the nutrient addition pond.

393 During the aquatic heat event (DOY 185-190), both ponds experienced a similar increase
394 in surface water temperatures (Figure 2B, Figure 4). However, there was a stark difference in the
395 response of bottom water temperatures at the deep and middle sites between the ponds (Figure
396 S4). During the heat event, bottom water temperature remained consistent at both the deep and
397 middle sites of the reference pond. The daily mean bottom water temperature increased by only
398 0.2 °C at both sites. In contrast, bottom water temperatures in the nutrient addition pond did
399 respond to the heat event, increasing by 1.3 °C and 1.4 °C at the deep and middle sites,
400 respectively. At the shallow site in both ponds bottom water temperatures increased 1-1.3 °C
401 during the heat event, suggesting an interaction with water depth. The different response of
402 bottom water temperature to the heat event related to differences in macrophyte biomass between
403 the two ponds. At the time of the heat event, macrophyte biomass (Figure 2D) and canopy height
404 (Figure 4) were greater in the reference pond because submersed macrophytes in the center of
405 the nutrient addition pond began to senesce following the first nutrient pulse (DOY 176; Figure
406 3). Overall, bottom waters in the nutrient addition pond warmed over the course of the heatwave
407 while cooler bottom water temperatures were maintained in the reference pond, despite
408 significant warming in the surface layers.

409

410 *Spatiotemporal Variation in Dissolved Oxygen*

411 In both ponds, DO varied across sites, vertically in the water column, and over time.
412 Weekly DO profiles captured broad seasonal trends in surface DO saturation as similar temporal
413 patterns were observed between weekly and high-frequency values (Figure 2C, S5A). Surface
414 waters were supersaturated with DO early in the summer in both ponds (Figure S5A). In the
415 reference pond, mean pond surface DO declined through DOY 170, remained fairly stable, and
416 then declined again during the heat event (DOY 185-190). In the nutrient addition pond, surface
417 DO saturation was similar to the reference pond before the nutrient addition. Surface DO then
418 decreased after the nutrient addition and through the heat event. Over a one-week period during
419 and after the heat event (DOY 186-192), DO saturation fell 22.4% and 50.4% in the nutrient
420 addition and reference ponds, respectively. After the heat event, surface DO in both ponds was
421 generally below saturation, with increased instances of DO saturation below 25%. Bottom water
422 DO saturation mirrored the seasonal pattern in surface DO (Figure S5B). For both ponds the
423 difference in DO saturation between surface and bottom waters was most pronounced early in
424 the season (reference pond range -87.1 to 23.3%; nutrient addition pond range -56.8 to 58.8%).
425 Surface and bottom water DO saturation became more similar throughout the summer, especially
426 following the heat event (reference pond range -2.1 to 17.0%; nutrient addition pond range -0.3
427 to 12.2%; Figure S5C).

428 For both surface and bottom water DO saturation, temporal variation on a seasonal scale
429 was greater than spatial variation across the ponds (Table 1). The coefficient of variation at a
430 given site in the pond over time was 3.6-5.2 times greater than the variation among sites on a
431 given sampling day. The magnitude of temporal versus spatial variation was similar between
432 ponds. Seasonal patterns in DO saturation followed some trends with declining macrophyte

433 biomass following the nutrient addition associated with lower DO; however, the heat event had
434 the most pronounced effect on seasonal DO patterns (Figure 6).

435

436 *Mechanisms of Macrophyte Control of Pond Hydrodynamics*

437 We used a one-dimensional hydrodynamic model to illuminate the mechanistic
438 relationship between macrophytes and pond thermal structure and test how macrophyte presence
439 mediates the degree to which bottom water temperatures warm in response to a heat event. We
440 based our model on the deep site of the reference pond. The model projections were consistent
441 with the observed temperature profiles (mean RMSE 1.67 °C across 8 discrete depths; Figure
442 S6). Error was greatest in the surface waters (0 m; RMSE 2.47 °C), where the model
443 underpredicted water temperature and did not capture the full range of observed diel oscillations,
444 especially early in the summer. Error was lower in the bottom waters of the pond (1.25-2 m). The
445 model underpredicted water temperature early in the summer but then showed good agreement
446 with the observed temperatures (RMSE range 1.05-1.52 °C). Overall, the model captured the
447 magnitude and seasonal trends of temperature throughout the water column.

448 We applied the hydrodynamic model to quantify differences in pond thermal structure
449 under contrasting scenarios of macrophyte presence (Figure 7). Specifically, we tested a scenario
450 in which there were no macrophytes present (i.e., canopy height and biomass density were set to
451 zero over the study period) and compared the results to the model output for the reference pond
452 when the observed macrophyte data were used (Table S2). Under the scenario where
453 macrophytes were present, the model predicted strong stratification during the period of peak
454 macrophyte biomass and canopy height in the middle of the summer and cool bottom water
455 temperatures throughout the study period (Figure 7A). In contrast, under the scenario in which

456 macrophytes were entirely absent from the pond, the model predicted complete water column
457 mixing and much warmer temperatures throughout the water column, especially following the
458 aquatic heat event (Figure 7B). When macrophytes were present, the increase in bottom water
459 temperatures over the course of the aquatic heat event was negligible (0.01 °C), and bottom
460 temperatures then remained below 25 °C for the remainder of the summer (maximum
461 temperature 23.3 °C; Figure 7A, C). However, the no-macrophyte scenario resulted in a 1.92 °C
462 increase in bottom water temperatures during the heat event and much higher maximum water
463 temperatures later in the summer (33.4 °C; Figure 7B, C). The model simulations demonstrate
464 that macrophyte structure increases thermal stratification and promotes cooler bottom water
465 temperatures, even during an extreme heat event.

466

467 **DISCUSSION**

468 We leveraged an ecosystem experiment and complimentary hydrodynamic model to test
469 macrophyte response to compounding stressors (i.e., stochastic external nutrient loading and
470 extreme heat events) and evaluate the consequences of stress-induced macrophyte senescence for
471 the physicochemical environment in two shallow, vegetated ponds. The type and severity of
472 environmental stressors differed between the study ponds, producing distinct temporal patterns in
473 macrophyte assembly and coverage. In the reference pond, both submersed and floating-leaf
474 macrophytes followed the expected phenology of spring emergence, summer growth, and then
475 senescence beginning in late summer. This natural phenology may be attributed to the lack of
476 severe physiological stressors in the reference pond. Although this pond did experience a heat
477 event, dense macrophyte beds limited the depth of water column warming and maintained stable
478 bottom water temperatures. In contrast, submersed macrophytes in the nutrient addition pond

479 began to senesce in mid-summer after the first experimental nutrient pulse and continuing after
480 the heat event. This early senescence was dominated by submersed macrophytes in the deeper,
481 central region of pond, suggesting that light limitation from phytoplankton shading was a
482 contributing stressor (Scheffer et al. 1993; Short et al. 1995; Phillips et al. 2016). Light limitation
483 may have been exacerbated by heat stress, which was likely more acute in the nutrient addition
484 pond as declining macrophyte biomass following the nutrient pulse allowed temperatures to rise
485 throughout the water column rather than just the surface layers. Early senescence of submersed
486 macrophytes in the nutrient addition pond as a result of compounding stressors altered the
487 thermal environment and DO saturation in relation to the reference pond.

488

489 *Thermal Structure*

490 On a seasonal scale, we observed strong temporal coherence between macrophyte growth
491 and senescence and pond thermal structure. Both ponds began to experience intermittent thermal
492 stratification in early summer, once macrophytes had grown to fill the majority of the water
493 column. This threshold is consistent with the expectation that canopy-forming macrophyte beds
494 may induce diurnal stratification once they occupy at least 50 percent of the water column (Vilas
495 et al. 2017). By mid-summer we observed strong thermal stratification, characterized by large
496 differences (around 10°C) between surface and bottom water temperatures, comparable to
497 stratification in other small, vegetated ponds (Andersen et al. 2017a; Vilas et al. 2017). However,
498 in other studies, macrophyte-induced stratification followed diel cycles, with daytime
499 stratification and nighttime overturn due to convective mixing (Martinsen et al. 2019; Sand-
500 Jensen et al. 2019), which can move through macrophyte structure more readily than external
501 turbulent mixing (Herb and Stefan 2005; Andersen et al. 2017b). Although we observed

502 nighttime convective cooling and mixing in the surface of the ponds during periods of
503 macrophyte-induced stratification, convective cooling did not mix the entire water column.
504 Stable thermal stratification began to break down in both study ponds during macrophyte
505 senescence. Declining macrophyte canopy height and biomass removed structural barriers that
506 had previously attenuated incoming solar radiation and dissipated wind-driven turbulent mixing
507 (Herb and Stefan 2004), allowing the warm, mixed surface layer to deepen and eventually mix
508 the entire water column. However, the timing of both destratification and macrophyte senescence
509 differed between the study ponds, driven by compounding stressors in the nutrient addition pond
510 and typical seasonal senescence in the reference pond. In the nutrient addition pond consistent
511 water column mixing began when macrophyte biomass and canopy height in the center of the
512 pond began to decline in response to phytoplankton-shading following the experimental nutrient
513 pulse and the heat event.

514 Macrophyte presence mediated how pond thermal structure responded to an extreme heat
515 event by restricting vertical heat transfer and maintaining cooler bottom water temperatures.
516 Both study ponds saw high surface water temperatures during the heat event. However, as
517 anticipated, the depth of water column warming and overall change in bottom water temperature
518 differed between the ponds as a result of differences in macrophyte coverage. At the time of the
519 heat event, macrophyte canopy height and biomass were rapidly declining in center of the
520 nutrient addition pond due to phytoplankton-shading. In the reference pond, macrophyte biomass
521 was still increasing and the canopy reached to fill over 75 percent of the water column. During
522 the heat event, a layer of cooler water between 0.25 and 0.5 m thick remained in the reference
523 pond, with only subtle increases in daily mean temperature (+0.2 °C). In contrast, the nutrient
524 addition pond saw increasing water temperatures throughout the water column, including a 1.3°C

525 increase in daily mean temperatures in the deepest waters of the pond over the 5-day heat event.
526 Given the observed differences between the study ponds, it is very likely that macrophyte
527 structure played a pivotal role in mediating the depth and degree of water column warming
528 during the aquatic heat event. Nevertheless, thermal structure in small waterbodies arises from
529 complex interactions between waterbody features and external forces (Herb and Stefan 2004;
530 Branco and Torgersen 2009). In order to explicitly test the mechanisms underpinning the
531 differences we saw between the study ponds and evaluate the role of macrophytes, we used a
532 hydrodynamic model for heat transport in vegetated waterbodies and simulated water column
533 thermal structure during a heat event under contrasting scenarios of macrophyte density.

534 The model results support our field observations and demonstrate that macrophytes
535 structure the thermal environment in shallow ponds through vertical attenuation of both light and
536 turbulent kinetic energy. Simulations of seasonal patterns in pond thermal structure under
537 variable macrophyte density revealed that macrophyte presence reduces the depth and degree of
538 water column warming both during and after an aquatic heat event and maintains cooler bottom
539 water temperatures throughout the course of the summer. Our experimental observations and
540 model simulations provide evidence that macrophytes may buffer small waterbodies from brief
541 aquatic heatwaves, at least with the vertical extent of warming in the water column. As aquatic
542 heatwaves become more widespread and frequent, heatwaves in small waterbodies are expected
543 to be shorter but more intense compared to larger waters (i.e., due to lower thermal inertia;
544 Woolway et al. 2021a; Polazzo et al. 2021; Woolway et al. 2022). Our findings suggest that
545 macrophytes can mediate the intensity of brief aquatic heatwaves in small lakes and ponds. As
546 such, managing for robust macrophyte communities is a potential tool for mitigating heatwave

547 effects in small waterbodies and promoting thermal heterogeneity to support other aquatic
548 organisms (Carpenter and Lodge 1986; Till et al. 2019).

549 Our hydrodynamic model enhances a conceptual and analytical framework for
550 incorporating biological structure into one-dimensional heat transport models (Herb and Stefan
551 2004). Overall, there was good agreement between the model predications and the observed
552 temperature profiles. The largest discrepancies between the model output and field data occurred
553 in early summer when the model underpredicted temperature. The early summer period
554 coincides with rapid growth in macrophyte biomass and canopy height. It is possible that the
555 temporal resolution of our field data during this time was insufficient to accurately capture
556 biological dynamics in the pond for the model input. Canopy height was measured weekly while
557 biomass was collected every two weeks. More frequent observations could have fit the
558 macrophyte input data closer to conditions captured by the high frequency temperature profiles.
559 Additionally, other one-dimensional heat transport models (i.e., Simstrat, GOTM, GLM) have
560 been shown to have greater error during extreme meteorological events, including atmospheric
561 heatwaves (Mesman et al. 2020). This tendency could explain some of the disparities between
562 model output and field data during a brief period of high air temperatures around DOY 155,
563 reflected in rapid increases in water temperature in the first 1.25 m of the profile. Our model
564 would benefit from additional testing across a variety of vegetated waterbodies to inform best
565 practices for applying this approach in other waters. However, our findings indicate that
566 macrophytes should be incorporated into hydrodynamic models for vegetated systems, and our
567 modeling approach provides an example of how to do so.

568

569 *Dissolved Oxygen*

570 Within both study ponds, DO saturation varied on a seasonal scale and responded to the
571 experimental nutrient pulse as well as the heat event. We had anticipated greater spatial variation
572 in DO saturation within each pond as a result of macrophyte-induced thermal stratification. This
573 expectation was based on evidence of steep vertical gradients of DO and dissolved inorganic
574 carbon in other shallow, vegetated waterbodies (Andersen et al. 2017a; Vilas et al. 2017).
575 However, most of the DO profiles were uniform with depth, suggesting that the balance of
576 processes releasing and consuming oxygen (e.g., gross primary production versus respiration)
577 were similar throughout the ponds or that DO was able to diffuse readily throughout the water
578 column. Temporal patterns in surface and bottom water DO saturation were connected to the
579 compounding environmental stressors, beginning with the initial experimental nutrient addition.
580 Following the first nutrient pulse, DO remained stable in the reference pond but declined below
581 100 percent saturation in the experimental pond. We might have expected to see increased DO
582 after the nutrient addition as a result of enhanced phytoplankton production, but the observed
583 decline in DO suggests that ecosystem respiration was stimulated by the added nutrients,
584 outweighing any increases in gross primary production. Stress-induced macrophyte senescence
585 likely contributed to this pattern through reduced macrophyte photosynthesis and dramatic
586 increases in macrophyte necromass, the decomposition of which would consume oxygen.
587 Although we observed differences in DO saturation between the ponds after the nutrient
588 addition, the most pronounced temporal shift in DO occurring following the heat event.

589 During the heat event, DO saturation fell sharply in both ponds and remained depressed
590 through the remainder of the study period. We had expected macrophyte structure to maintain
591 stable DO saturation in the bottom waters of the reference pond; however, both ponds responded
592 similarly, regardless of the differences in macrophyte coverage and thermal structure. Our

593 observations support findings from freshwater and marine systems that aquatic heatwaves are
594 associated with declines in DO due to reduced gas solubility and increased respiration rates (Ali
595 et al. 2016; Brauko et al. 2020; Tassone et al. 2021). Reduced DO availability has consequences
596 for nutrient and carbon cycling as well as greenhouse gas production. For example, anoxia within
597 the sediment profile can mobilize phosphorus from redox-sensitive minerals (i.e., anaerobic
598 internal phosphorus loading; Orihel et al. 2017). Anoxic conditions in the sediments also
599 promote denitrification and methanogenesis, two microbial processes that can produce potent
600 greenhouse gases (i.e., nitrous oxide and methane respectively; Fonseca et al. 2017; Ming et al.
601 2022). Both denitrification and methanogenesis require organic carbon substrate, which was
602 augmented by macrophyte necromass in the nutrient addition pond (Boon et al. 1995; Bastviken
603 et al. 2005; Deng et al. 2020). Although both ponds likely experienced shifts in other elemental
604 cycles as a result of reduced DO saturation during the aquatic heatwave, the consequences for
605 greenhouse gas production were likely more pronounced in the nutrient addition pond due to
606 early macrophyte senescence increasing organic matter inputs to the sediments. This difference
607 between the study ponds highlights the complex interactions that may arise from compounding
608 stressors in lentic ecosystems.

609

610 *Conclusions*

611 Aquatic heatwaves are becoming more frequent and are likely to co-occur with
612 chlorophyll-*a* maxima in lakes and ponds (i.e., algal blooms; Woolway et al. 2021b). As such,
613 there is a growing need to understand and forecast how lentic ecosystems will respond to
614 compound, extreme events. Our findings illuminate the role that macrophyte-hydrodynamic
615 interactions play in mediating ecosystem response to multiple stressors. Overall, dense

616 macrophyte structure can buffer small waterbodies from brief aquatic heat events by restricting
617 vertical heat transport. However, declines in DO saturation are expected during heat events,
618 regardless of macrophyte coverage. Our findings also reinforce that small lakes and ponds are
619 vulnerable to nutrient loading and temperature extremes. Given the prevalence and importance of
620 small waterbodies for biodiversity and biogeochemical processing, it is essential that we
621 continue to build our mechanistic understanding of the effects of compound, extreme events in
622 these ecosystems and the role that macrophytes can play in ameliorating stressors.

623

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636

637 *Data Availability Statement*

638 The data supporting the conclusions and the analysis code are available in the Github repository
639 https://github.com/AlbrightE/pond_physics_2022. Following manuscript acceptance, the Github

640 repository will be archived on Zenodo, and the data will be published on the Environmental Data
641 Initiative's online repository under a Creative Commons Attribution license (CC-BY).

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825 **TABLES**

826 **Table 1.** Temporal versus spatial variation in DO Saturation

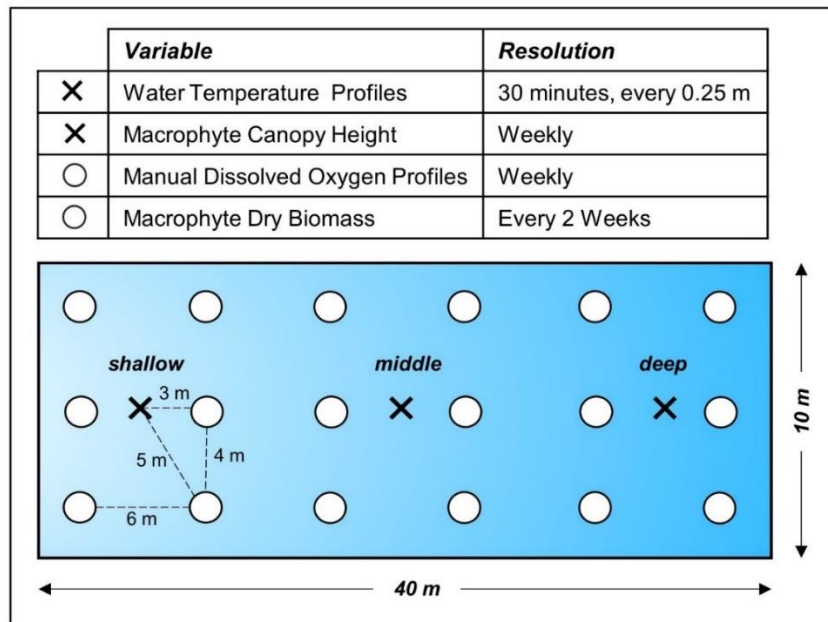
Layer	Pond	Temporal Variation in DO Saturation		Spatial Variation in DO Saturation	
		CV (%) Mean	CV (%) Range	CV (%) Mean	CV (%) Range
Surface	Reference	28.9	20.2 – 34.4	5.9	1.8 – 11.3
	Nutrient Addition	22.8	20.6 – 28.8	4.4	2.1 – 9.5
Bottom	Reference	34.4	24.1 – 45.6	7.5	2.0 – 16.3
	Nutrient Addition	26.8	22.1 – 40.3	7.5	2.0 – 16.9

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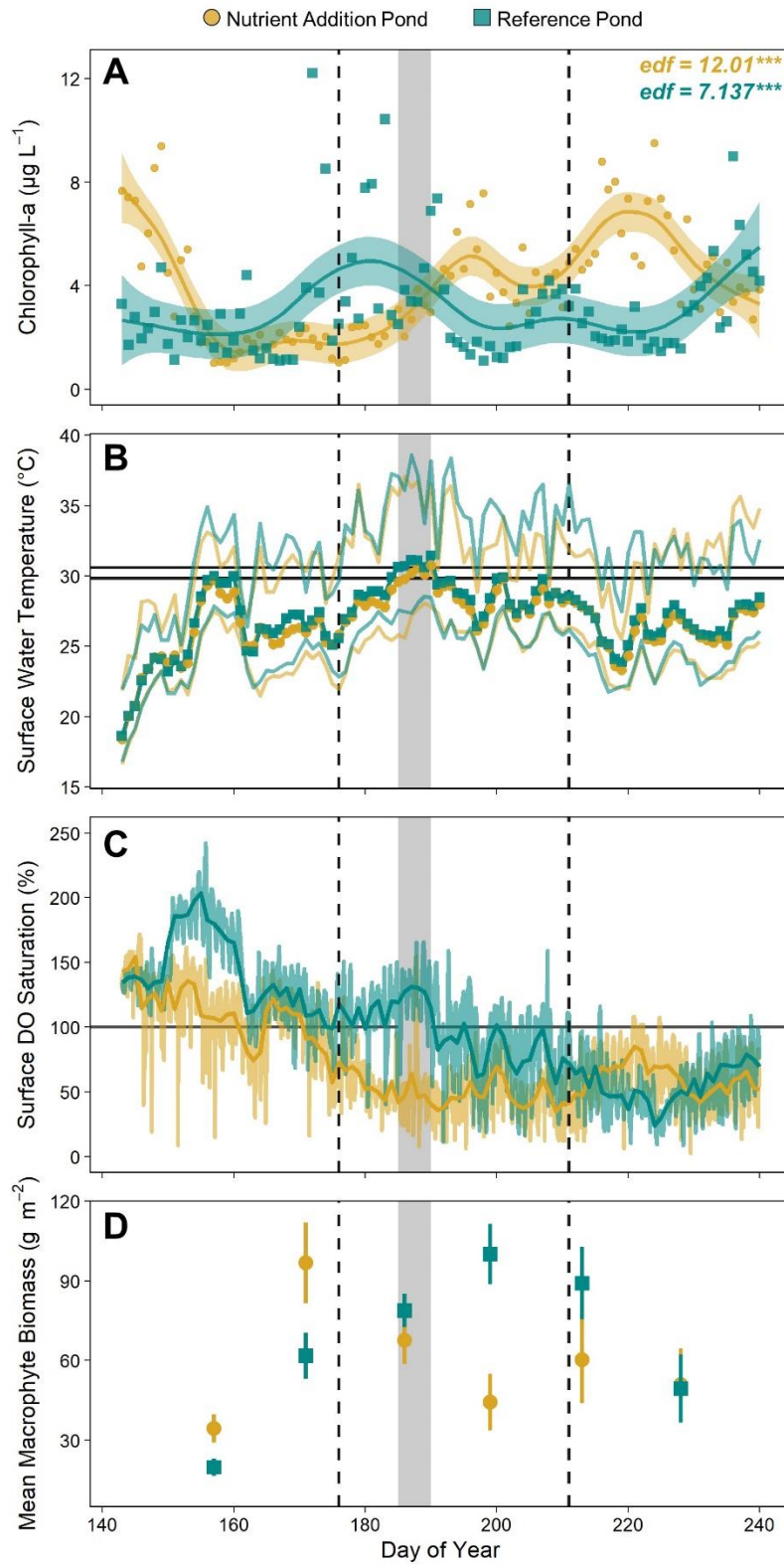
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830 **FIGURES**



831

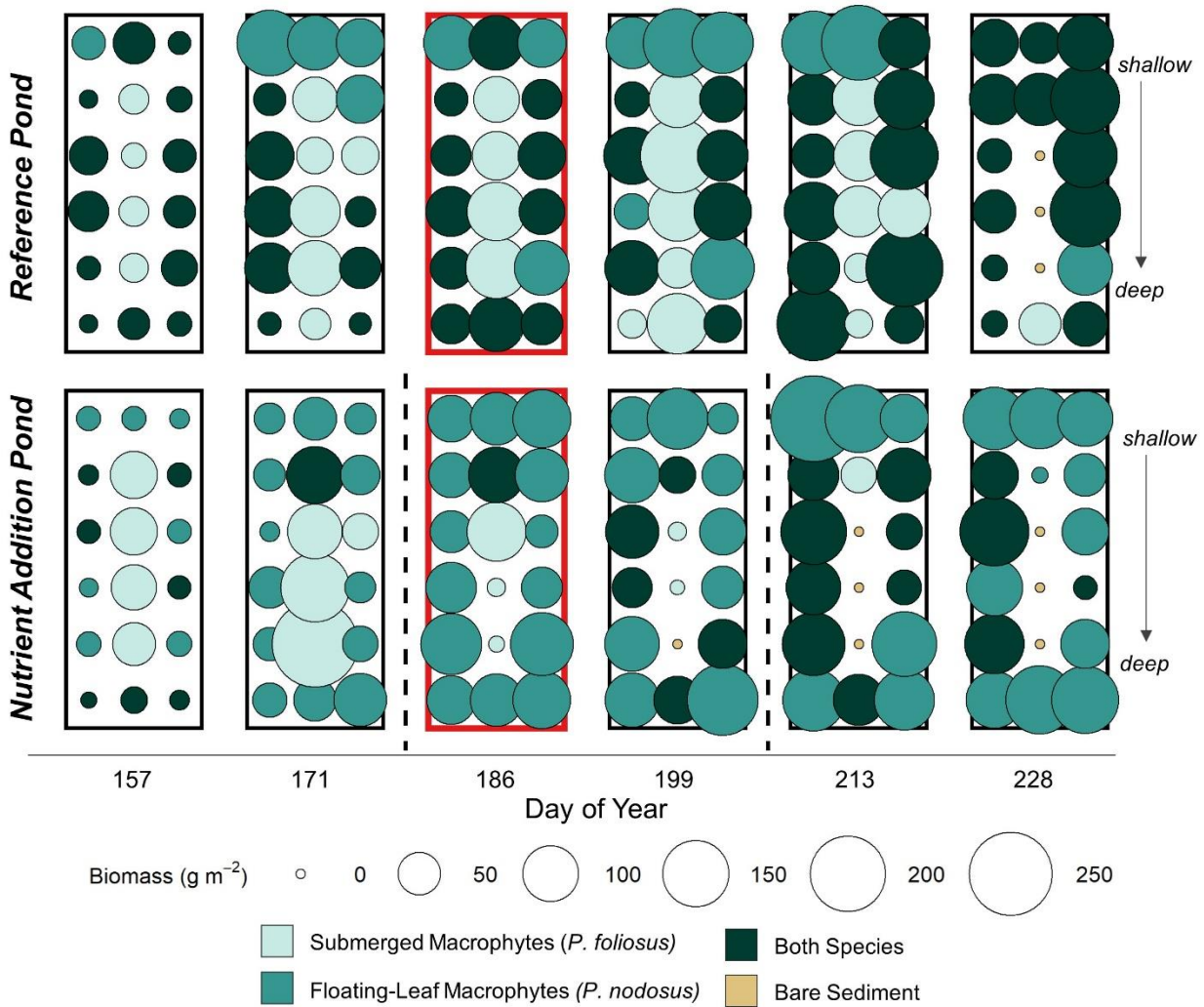
832 **Figure 1.** Spatial and temporal resolution of measured variables and sample site locations.



833

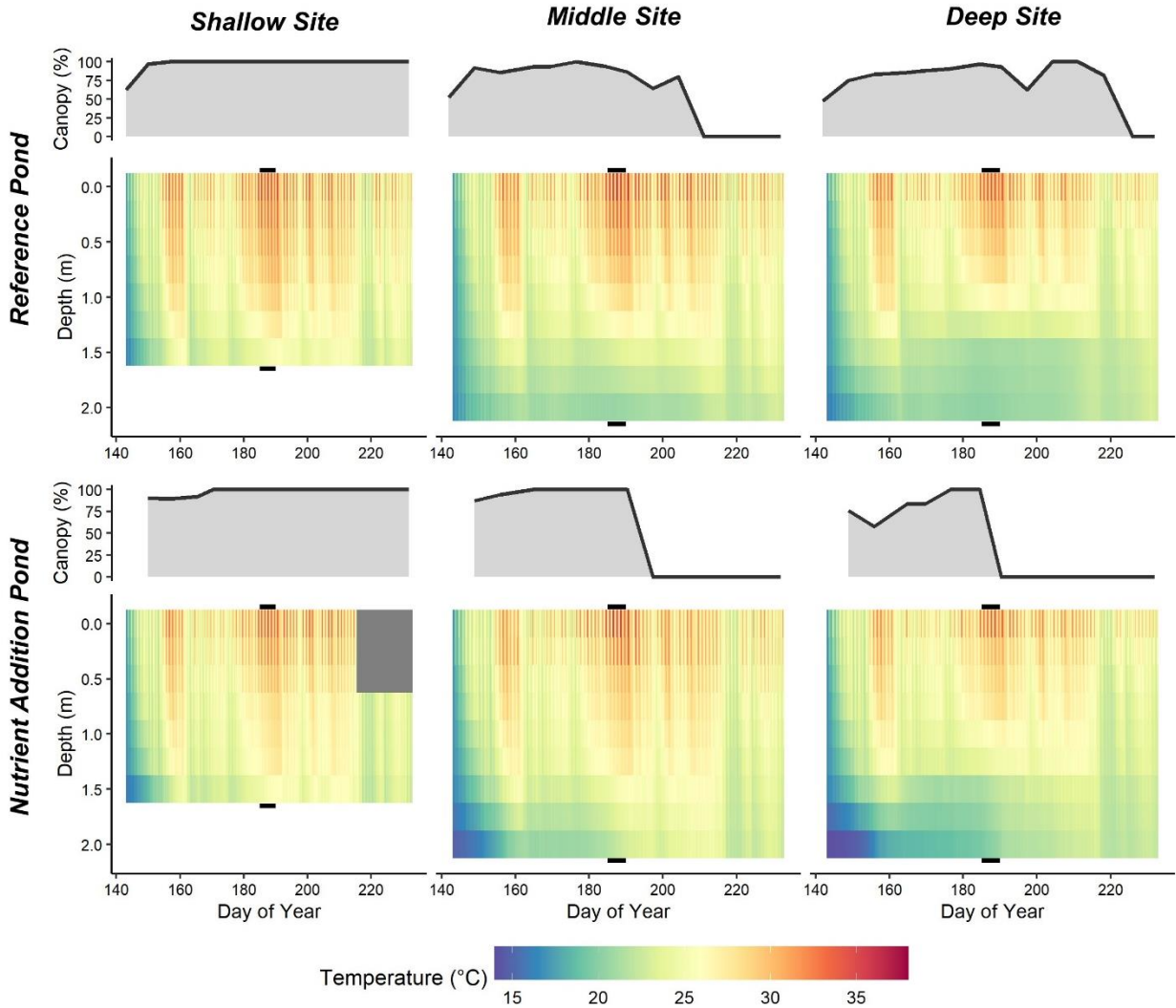
834 **Figure 2.** (A) Timeseries of surface chlorophyll-*a* concentrations. Values from each pond were
 835 GAM-fit to highlight temporal trends. The asterisks next to the effective degrees of freedom

836 (edf) correspond to the estimated p-value, with *** denoting $p < 0.001$. (B) Timeseries of mean
 837 daily surface (0-0.25 m) water temperature. Surface sensors at the deep and middle sites were
 838 used for these pond averages. The 95th percentile surface water temperature thresholds are noted
 839 in horizontal lines for the nutrient addition (29.8°C) and reference (30.6°C) ponds. (C) High
 840 frequency surface dissolved oxygen (DO) saturation at the deep site of each pond over the study
 841 period. The daily mean is plotted in a darker line. The horizontal grey line marks 100 percent
 842 saturation. (D) Time series of mean macrophyte biomass in each pond. Error bars mark the
 843 standard error of the mean. Across all panels, dashed lines mark the timing of nutrient additions
 844 (DOY 176, 211) and the grey box highlights the aquatic heat event (DOY 185-190).
 845

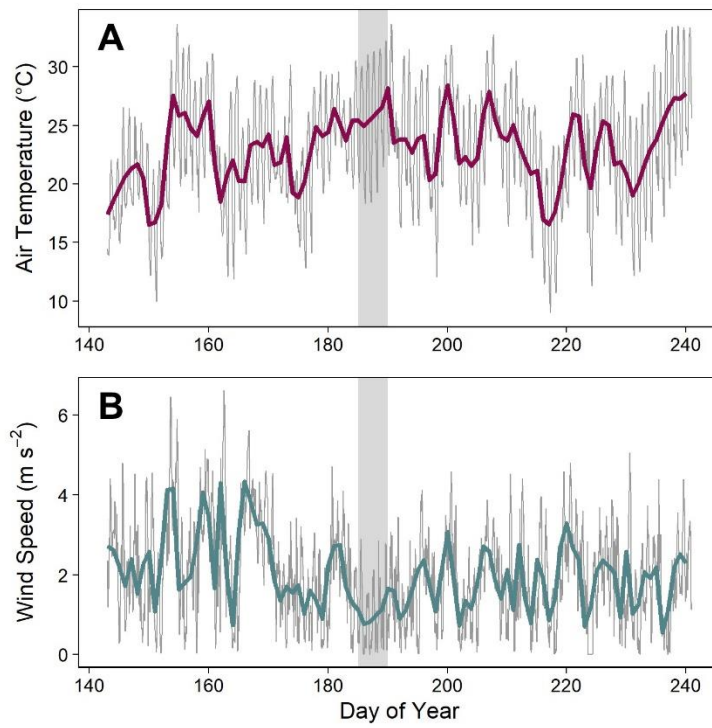


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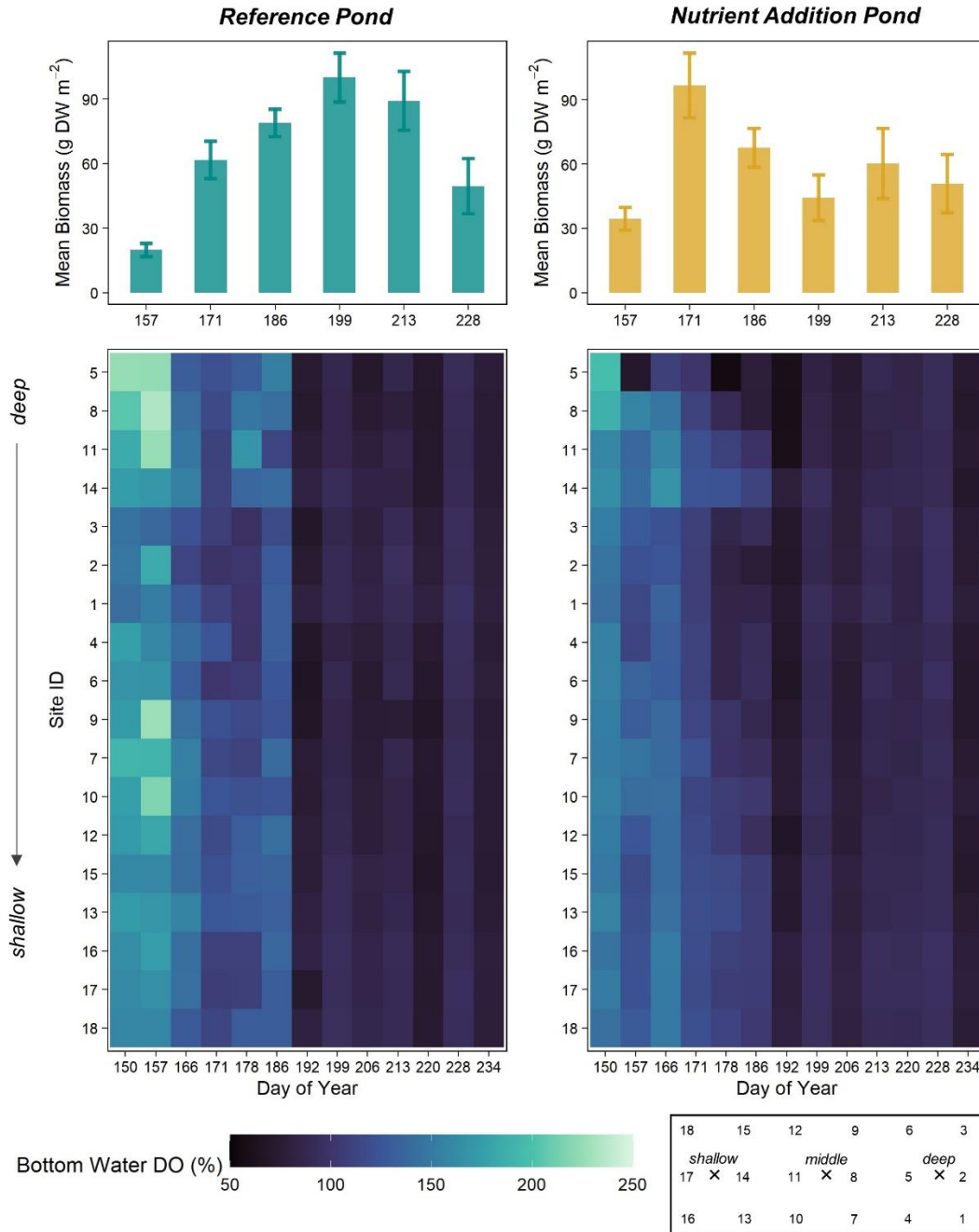
847 **Figure 3.** Spatial and temporal variation in macrophyte dry biomass and community
 848 composition. Biomass was sampled during the aquatic heat event (DOY 185-190), as noted by
 849 the thicker, red pond outline on DOY 186. Dashed lines mark nutrient additions on DOY 176
 850 and 211.



851
 852 **Figure 4.** Spatial and temporal variation in macrophyte canopy height and temperature profiles.
 853 Canopy height is plotted as a percent of the water column height. Black, horizontal lines mark an
 854 aquatic heat event from DOY 185-190. High frequency water temperature loggers were placed
 855 every 0.25 m of the water column up to 1.5 m and then every 0.5 m to the bottom. Temperature
 856 values at 1.75 m were interpolated for the sake of visualization. Several surface sensors (0-0.5 m)
 857 from the shallow site of the nutrient addition pond lost power following DOY 215 (grey box).
 858 An initial canopy height measurement is missing from the nutrient addition pond because the
 859 water was not sufficiently clear to see and measure the top of the canopy. The canopy height was
 860 less than 1 m or 50-75% of the water column at this time.



861
862 **Figure 5.** Air temperature (A) and windspeed (B) over the study period. Daily mean values are
863 plotted in thicker lines over the hourly time series.

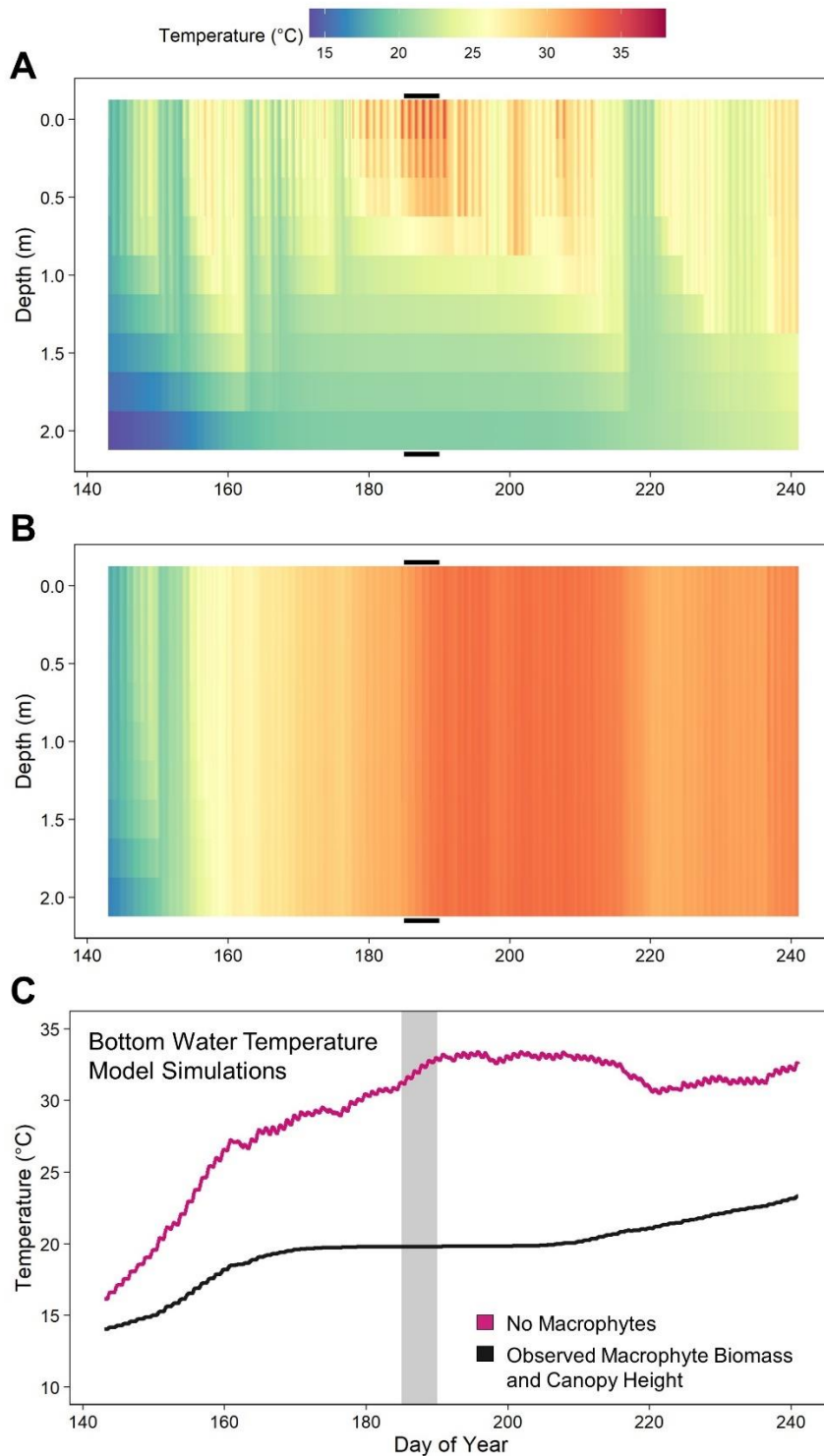


864

865 **Figure 6.** Spatiotemporal variation in bottom water dissolved oxygen (DO) saturation. Sampling
 866 sites are ordered approximately from the deepest to most shallow locations. The pond-averaged
 867 biomass values are plotted above. Error bars represent the standard error of the mean.

868

869



870
 871 **Figure 7.** Model simulation results contrasting a no macrophyte scenario with the observed
 872 macrophyte coverage from the reference pond. Predicted temperature profiles with (A) observed
 873 macrophyte biomass and canopy height versus (B) no macrophytes present in the reference pond.
 874 Black, horizontal lines mark an aquatic heat event (DOY 185-190). (C) Timeseries of predicted
 875 daily mean bottom water (2m) temperatures with no macrophytes versus observed macrophyte
 876 biomass and canopy height.

877 **Appendix S1**

878

879 **Macrophyte-hydrodynamic interactions mediate stratification and dissolved oxygen**
880 **dynamics in ponds**

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885

886 **Introduction**

887 The appendix contains additional methods text detailing the algorithms and approach for
888 the hydrodynamic model as well as an overview of calibrated model parameters (Table S1) and
889 simulation input data for testing heat event response under variable macrophyte scenarios (Table
890 S2). Model fit is illustrated in Figure S6.

891 Figure S1 provides an example of individual dissolved oxygen (DO) saturation profiles to
892 illustrate there was no evidence of mid-depth DO maxima. We provide time series of estimated
893 thermocline depth (Figure S2) and Lake Number (Figure S3) for each pond as well as time
894 series of bottom water temperatures at each sampling site on each pond (Figure S4). We further
895 provide supplementary visualizations for our DO data, specifically time series of surface and
896 bottom water DO saturation and the difference between surface and bottom water DO (Figure
897 S5).

898

899

900

901 **Model Equations and Approach**

902 *Model description*

903 A one-dimensional, integral energy model was developed to simulate the temperature, heat flux
904 and stratification dynamics in small, vegetated ponds. Model algorithms are based on the
905 MINLAKE model (Ford and Stefan 1980; Riley and Stefan 1988; Herb and Stefan 2004).

906

907 Heat transport is implemented through the one-dimensional temperature diffusion equation:

908 (Eq. 1)
$$\frac{\partial T}{\partial t} = \frac{\partial}{\partial z} \left(K_z \frac{\partial T}{\partial z} \right) + \frac{H(z)}{\rho_w c_p} + \frac{H_{geo}(z)}{\rho_w c_p}$$

909 where K_z is the vertical turbulent diffusion coefficient, H is internal heat generation due to
910 incoming solar radiation, ρ_w is water density, c_p is specific heat content of water, and H_{geo} is
911 internal geothermal heat generation.

912

913 Internal heat generation is implemented as:

914 (Eq. 2)
$$H(z) = (1 - \alpha) I_s^{-((k_d + k_m)z)}$$

915 where α is the water albedo, I_s is total incident short-wave radiation, and k_d is a light attenuation
916 coefficient and k_m is the light attenuation coefficient due to macrophyte biomass.

917

918 For the upper, surface boundary condition we assume a Neumann type for the temperature
919 diffusion equation:

920 (Eq. 3)
$$H_{net} = \rho_w c_p \left(K_z \frac{\partial T}{\partial z} \right)$$

921 where H_{net} is the net heat flux exchange between atmosphere and water column:

922 (Eq. 4)
$$H_{net} = H_{lw} + H_{lwr} + H_v + H_c$$

923 where H_{lw} is the incoming long-wave radiation, H_{lwr} is emitted radiation from the water column,
924 H_v is the latent heat flux, and H_c is the sensible heat flux. Implementations to estimate the
925 respective heat fluxes were taken from Livingstone and Imboden (1989) and Goudsmit et al.

926 (2002). Latent and sensible heat fluxes were calculated using the atmospheric stability
927 algorithms from Verburg and Antenucci (2010).

928

929 The lower, sediment boundary condition was prescribed as:

930 (Eq. 5)
$$\left(K_z \frac{\partial T}{\partial t}\right) = 0$$

931

932 The model algorithm is modularized into three components: (a) heat generation from boundary
933 conditions and vertical diffusion, (b) turbulent mixing up to the mixed layer depth, and (c)
934 convective overturn to account for density instabilities.

935

936 *(a) Heat generation from boundary conditions and vertical diffusion*

937 In the first step the heat fluxes H and H_{geo} are applied over the vertical water column. The
938 atmospheric heat flux, H_{net} , is applied to the surface layer.

939 Simultaneously, vertical turbulent diffusion between adjacent grid cells is calculated. Here, we
940 applied the implicit Crank-Nicholson scheme. The vertical turbulent diffusion coefficient K_z is
941 calculated based on the empirical equations by Hondzo and Stefan (1993) for lakes in Minnesota
942 as a function of the buoyancy frequency:

943 (Eq. 6)
$$K_z = a_k(N^2)^{-0.43}$$

944 where a_k is an empirical factor accounting for the surface area of the lake A_s :

945 (Eq. 7)
$$a_k = 0.00706(A_s)^{0.56}$$

946 and N^2 is the squared buoyancy frequency:

947 (Eq. 8)
$$N^2 = \frac{g}{\rho_w} \frac{\partial \rho_w}{\partial z}$$

948 Values of N^2 less than $7.0 \cdot 10^{-5} \text{ s}^{-2}$ were set to $7.0 \cdot 10^{-5} \text{ s}^{-2}$.

949

950

951 *(b) Mixed layer depth*

952 In the third step, we quantified the depth where the amount of external kinetic energy by wind
953 shear stress equals the internal potential energy of the water column. Up to this mixed layer

954 depth, z_{ml} , adjacent layers are subsequently mixed to account for a wind shear stress acting over
955 the vertical water column. Here, the kinetic energy KE is described as:

956 (Eq. 9) $KE = \tau u^* \Delta t$

957 where τ is the surface turbulent shear stress, and u^* is the surface shear velocity, which was
958 calculated from wind velocity as:

959 (Eq. 10) $u^* = \sqrt{\frac{C_{10}\rho_a}{\rho_w}} U_2$

960 where C_{10} is the wind stress coefficient dependent on the measured wind speed U_2 at 2 m height
961 above the water surface, and ρ_a is the density of air, respectively (Herb and Stefan, 2005).

962

963 The dissipation of turbulent kinetic energy (DKE) by macrophytes is incorporated as:

964 (Eq. 11) $DKE_z = (\rho_w \hat{a} C_d)(u^*)^3 \Delta z \Delta t$

965 where \hat{a} is the ratio of plant surface area per unit volume water and C_d is the plant form drag
966 coefficient.

967

968 The potential energy of the water column for each layer over the depth is calculated as:

969 (Eq. 12) $PE_z = g Z_z (Z_{z+1} - Z_{cv}) \Delta \rho$

970 where g is gravitational acceleration, z_{cv} is the center of volume depth, and $\Delta \rho$ is a density
971 change from the current layer to the next layer below.

972

973 The mixed layer depth z_{ml} is calculated by incrementally increasing the comparison between the
974 difference of total kinetic energy KE to the dissipation of KE by macrophytes, and the internal
975 potential energy PE as:

976 (Eq. 14) $z_{ml} \rightarrow PE_{z+1} > (KE - DKE_z)$

977

978

979

980 (c) *Convective overturn*

981 In the fourth step, any density instabilities over the vertical water column are mixed with a first
982 stable layer below an unstable layer. Here, we applied an area weighed mean of temperature
983 between adjacent layers to calculate the temperature of the mixed layer.

984

985 *Appendix References*

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1013

1014

1015 **Table S1.** *Calibrated model parameter values*

Variable	Description	Value	Parameterization
C_d	Wind momentum drag	0.0013	Calibration to field data
K_m	Macrophyte light extinction coefficient	0.04	Calibration to field data
$C_d\ plant$	Macrophyte momentum drag coefficient	1	Herb and Stefan (2004)
\hat{a}	Macrophyte area to volume	0.5	Calibration to field data
<i>Wind factor</i>	Wind speed multiplier	1	Best field calibration did not use wind speed or diffusion rate multipliers (so factors equal to 1)
<i>Diffusion factor</i>	Diffusion rate multiplier	1	
<i>Short wave factor</i>	Shortwave radiation multiplier	0.9	Calibration to field data
H_{geo}	Sediment heating coefficient	0	Calibration to field data
ρ_m	Macrophyte density	30	Calibration to field data

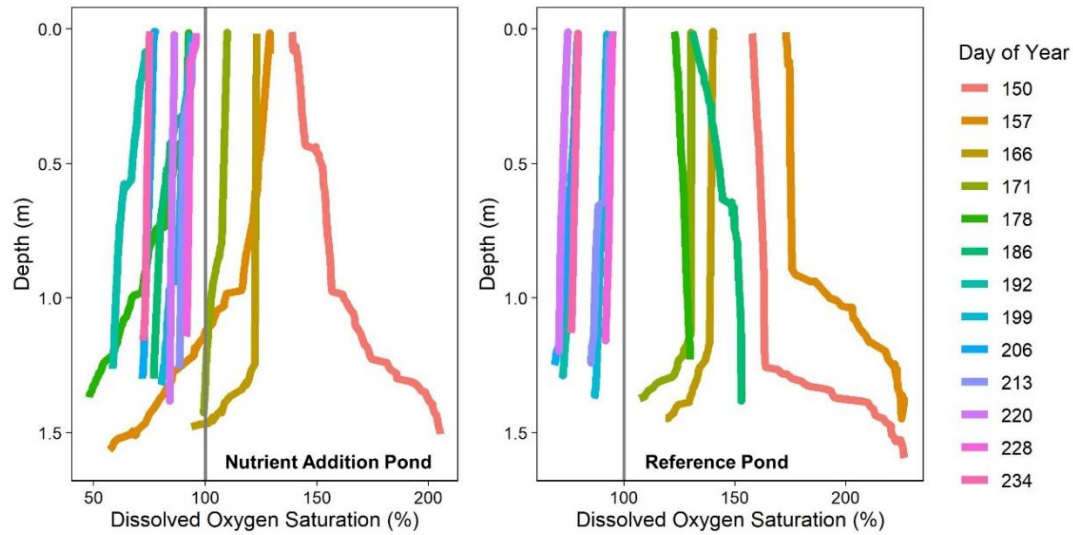
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1017 **Table S2.** *Input macrophyte data for model simulations*

Field Data – Reference Pond															
DOY	129	143	150	157	166	171	178	186	192	199	206	213	220	228	234
Canopy (m)	0.5	1	1.5	1.6	1.6	1.6	1.6	1.6	1.45	1	1.6	1.8	1/4	0.5	0.5
Biomass (g m ²)	NA	NA	NA	14.5	NA	48.4	NA	101.4	NA	73.6	NA	15.4	NA	22.4	NA
Simulation – No Macrophytes in Reference Pond															
DOY	129	143	150	157	166	171	178	186	192	199	206	213	220	228	234
Canopy (m)	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Biomass (g m ²)	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

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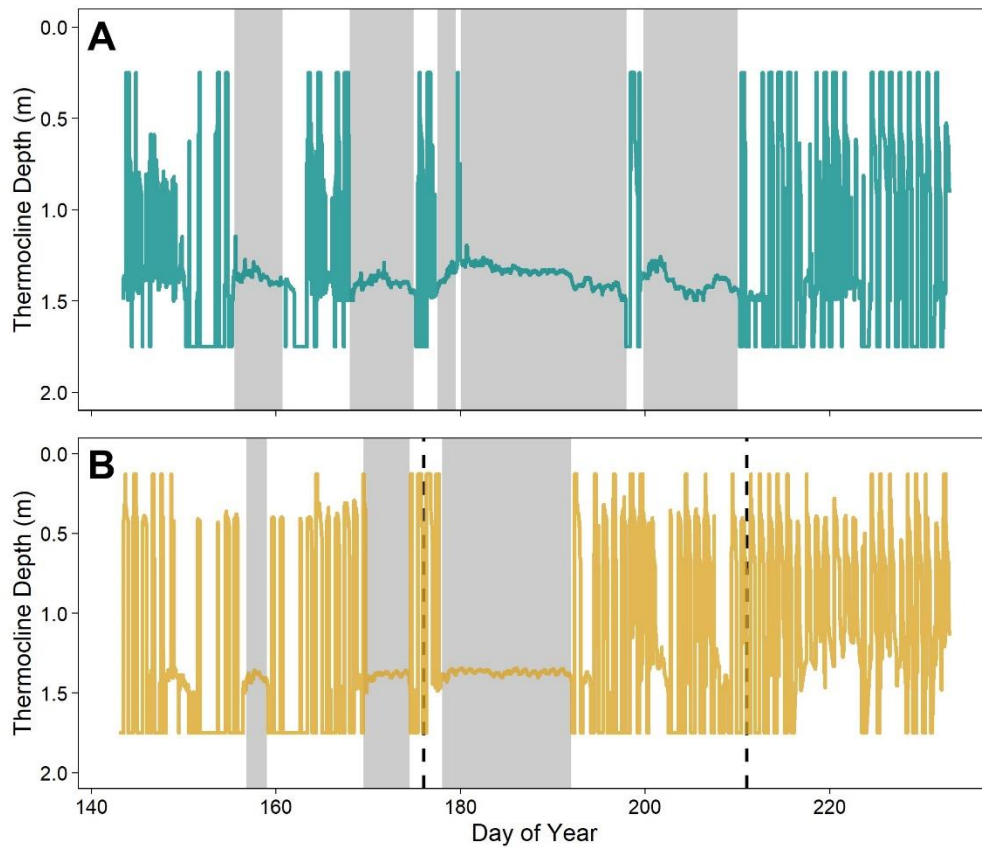
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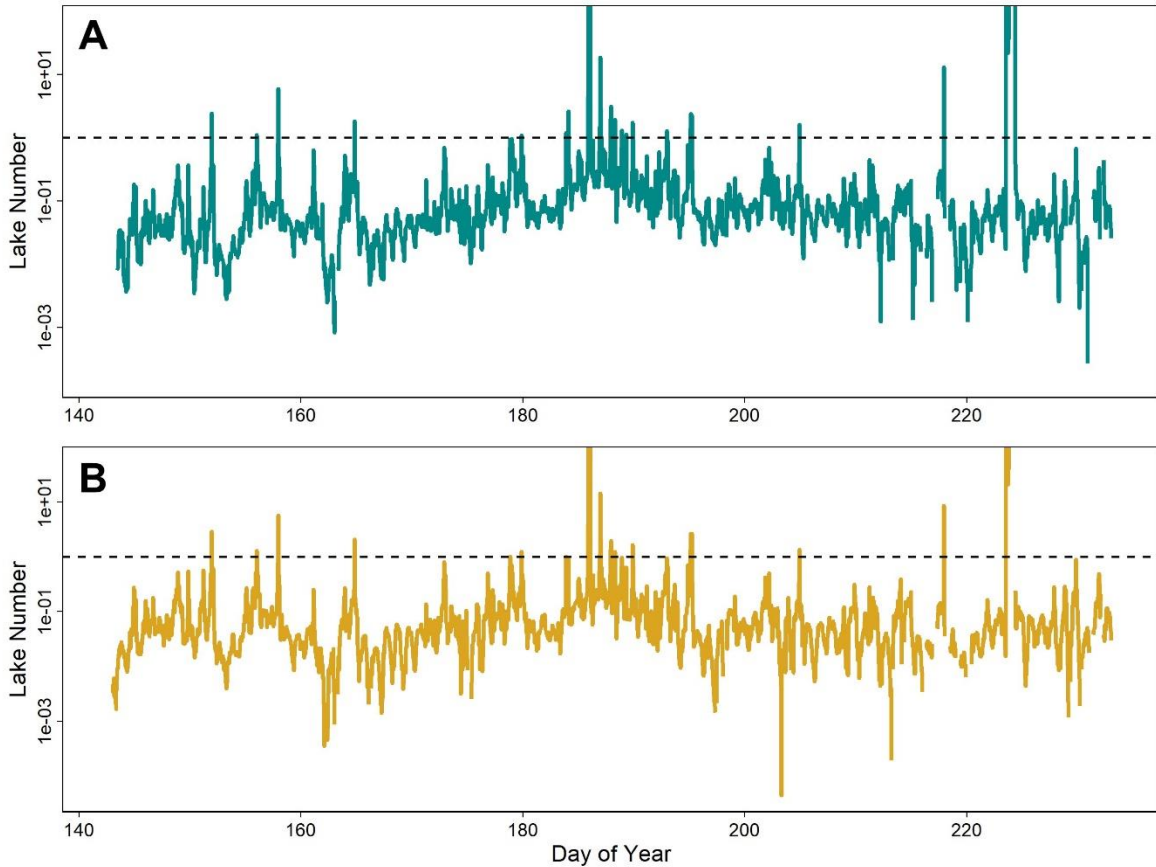
1021 **Figure S1.** Examples of individual dissolved oxygen (DO) saturation profiles near deep sites.
 1022 Profiles are color-coded by the DOY of the sampling event. These profiles were taken in the
 1023 deeper, central region of the ponds. The vertical grey line highlights 100 percent saturation.

1024



1025

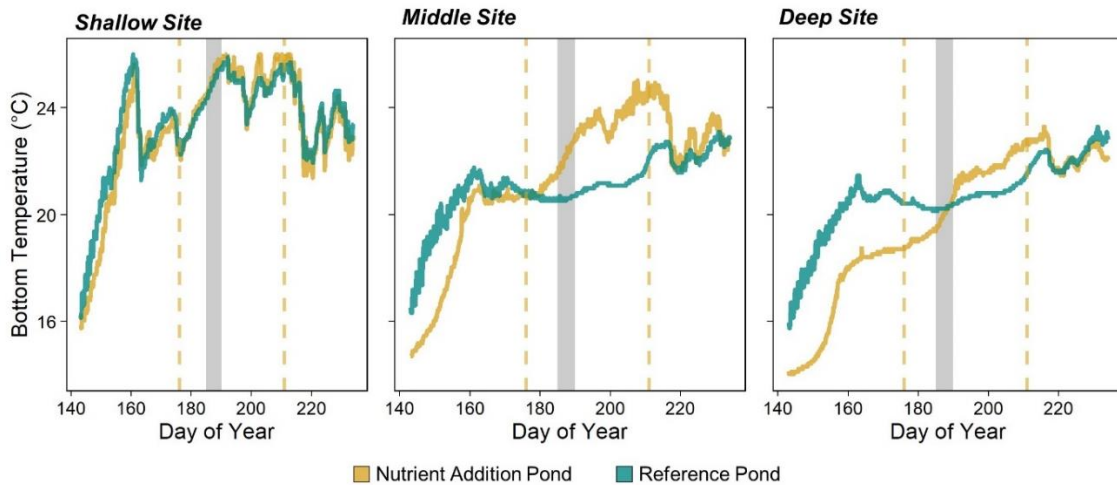
1026 **Figure S2.** Estimated thermocline depth by pond. Periods of a stable thermocline for the
 1027 reference (A) and nutrient addition (B) ponds are noted in the grey shaded boxes. Nutrient
 1028 additions are noted with dashed black lines. Estimated thermocline values oscillating rapidly
 1029 between the near surface and bottom waters were not considered time of stable stratification.



1030

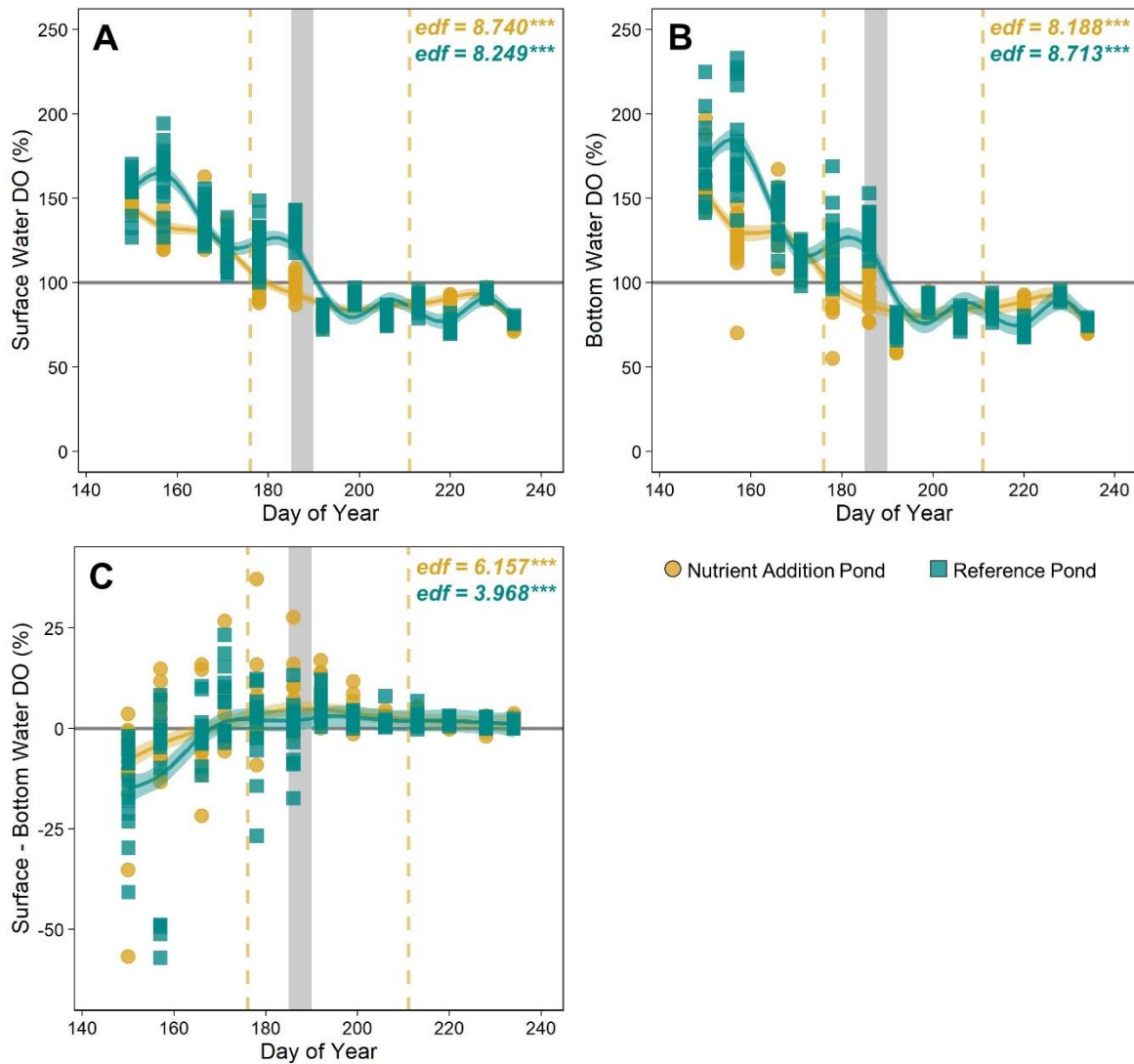
1031 **Figure S3.** Time series of Lake Number for the reference (A) and nutrient addition (B) ponds. A
 1032 value of one (horizontal dashed line) indicates that the internal stability of the water column is
 1033 balanced with wind stress. Values greater than one indicate that the water column is strongly
 1034 stratified while values less than one mean that stratification is weak compared to wind shear
 1035 stress.

1036



1037

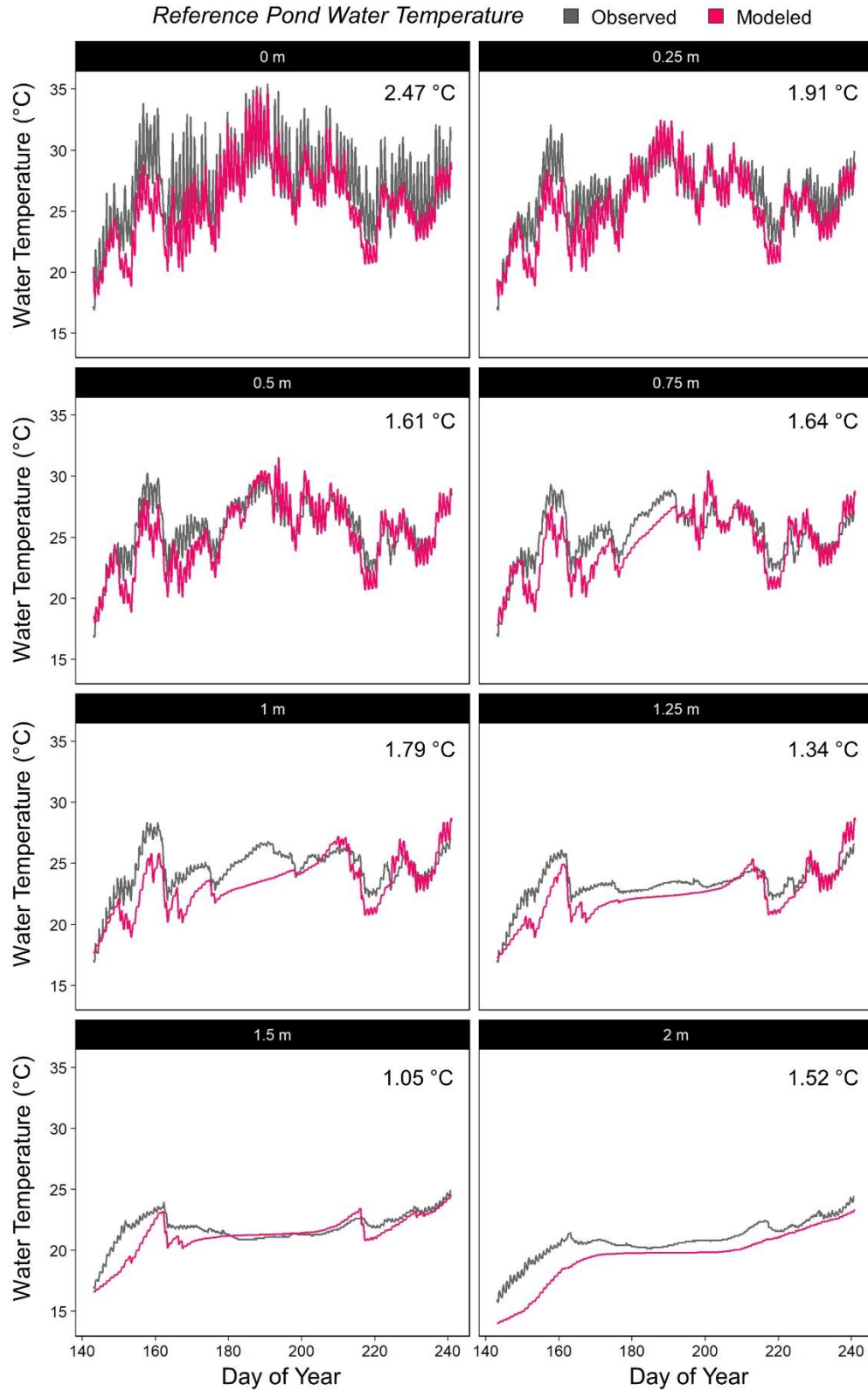
1038 **Figure S4.** Bottom water temperature time series across sampling sites. The grey box marks an
 1039 aquatic heat event from DOY 185-190. Yellow dashed lines indicate the nutrient additions on
 1040 DOY 176 and 211.



1041

1042 **Figure S5.** Time series of surface and bottom water dissolved oxygen (DO) saturation and the
 1043 difference between the two. (A) Surface water DO saturation, (B) bottom water DO saturation,
 1044 and (C) the difference between surface and bottom water DO saturation is based on weekly
 1045 profiles at 18 sites across each pond. Negative values of delta-DO indicate greater DO saturation
 1046 in the bottom waters while positive values mean DO is higher in the surface layer. Values from
 1047 each pond were GAM-fit to highlight temporal trends. The asterisks next to the effective degrees
 1048 of freedom (edf) correspond to the estimated p-value, with *** denoting $p < 0.001$. Across both
 1049 panels, dashed yellow lines mark the timing of nutrient additions and the grey box highlights the
 1050 aquatic heat event.

1051



1052

1053

1054

1055

Figure S6. Model fit at discrete depths in the reference pond. Predicted water temperatures from the hydrodynamic model are plotted against the observed temperatures at the same depths. The root mean square error (RMSE, °C) are noted in the top right corner of each panel