

1 **Influence of macrophytes on stratification and dissolved oxygen dynamics in ponds**

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20 **ABSTRACT**

21 1. Small waterbodies are sensitive to stressors such as nutrient enrichment and heatwaves.

22 However, when present, macrophytes may mediate these compounding stressors through their
23 influence on water column thermal structure. Canopy-forming macrophyte beds can induce
24 thermal stratification, which may limit the depth and degree of water column warming during
25 heatwaves.

26 2. We leveraged an ecosystem experiment and hydrodynamic model to evaluate how macrophyte
27 biomass, thermal structure, and dissolved oxygen (DO) responded to the interaction of
28 episodic nutrient loading and periods of high temperatures in two shallow, temperate ponds
29 (mean depth 0.8 m, maximum depth 2 m). We added nutrients to one pond, simulating storm-
30 driven loading, while the other pond served as an unmanipulated reference. Following the first
31 nutrient addition both ponds experienced a 5-day period of high surface water temperatures.

32 3. Submersed macrophytes in the nutrient addition pond began to senesce mid-summer, likely a
33 result of phytoplankton shading from the nutrient addition and heat stress, while macrophytes
34 in the reference pond followed expected seasonal patterns, senescing in early autumn.

35 4. We found that macrophytes structured the thermal environment in the ponds through vertical
36 attenuation of turbulent kinetic energy and light. Macrophytes reduced the vertical extent of
37 water column warming during the sustained heat event by 0.25-0.5 m and maintained cooler
38 bottom temperatures (up to 2.5 °C cooler) throughout the summer, suggesting that
39 macrophytes may buffer small waterbodies from heatwaves. Seasonal patterns in DO
40 saturation also followed trends in macrophyte biomass; however, during the heat event, DO
41 saturation fell sharply (declined by 22.4 to 50.4%) in both ponds and remained depressed
42 through the remainder of the summer.

43 5. *Synthesis*: Our findings reveal that canopy-forming aquatic plant beds can buffer ponds from
44 brief aquatic heat events but also that the plants themselves are sensitive to nutrient loading
45 and temperature extremes. These results contribute to our mechanistic understanding of the
46 effects of compound, extreme events in small waterbodies and the role aquatic plants can play
47 in mediating these stressors. This understanding is necessary for adaptive management of
48 small waterbodies such that these systems will continue to support freshwater biodiversity.

49
50 **Keywords:** dissolved oxygen, heatwaves, hydrodynamics, macrophytes, phenology, ponds,
51 water temperature

52 53 **INTRODUCTION**

54 Small lakes and ponds are the most abundant lentic ecosystems worldwide and sites of
55 critical importance for freshwater biodiversity and biogeochemical cycles (Scheffer et al., 2006;
56 Cole et al., 2007; Messenger et al., 2016). Small waterbodies are active components of regional-
57 and global-scale carbon cycling, both as sites of burial and storage (Wilkinson et al., 2018;
58 Taylor et al., 2019) and as sources of atmospheric carbon dioxide and methane (Holgerson &
59 Raymond, 2016; Peacock et al., 2021). Small lakes and ponds enhance local and regional
60 biodiversity, due in part to abundant macrophytes (Van Geest et al., 2003). Robust macrophyte
61 communities support greater diversity of other taxa by providing food resources and habitat
62 structure (Williams et al., 2004; Scheffer et al., 2006). In addition to these biotic interactions,
63 macrophytes play a pivotal role in the ecosystem structure and function of small waterbodies.
64 Interactions between hydrodynamics and the physical structure of macrophyte beds influence
65 turbulent mixing, water column thermal structure, and sediment stability as well as whole-lake

66 productivity and nutrient cycling (Carpenter & Lodge, 1986; Scheffer et al., 1993; Licci et al.,
67 2019). Despite the prevalence and importance of small, vegetated lakes and ponds, these
68 ecosystems remain understudied (Downing, 2010; Biggs et al., 2017), particularly with regards
69 to the role of macrophyte-hydrodynamic interactions in mediating ecosystem response to
70 stressors.

71 The characteristic morphometry of small waterbodies makes these ecosystems more
72 susceptible to stressors. For example, limited water volume and shallow mean depth reduce both
73 dilution potential and thermal inertia making small waterbodies sensitive to nutrient loading and
74 intense heatwaves (Biggs et al., 2017; Woolway et al., 2021a; Polazzo et al., 2022). As a result,
75 eutrophication and aquatic heatwaves may act as compounding stressors for small lakes and
76 ponds. Aquatic heatwaves are periods of sustained high surface water temperatures relative to
77 local and seasonal baseline conditions (Hobday et al., 2016; Tassone et al., 2021; Woolway et
78 al., 2021a). The frequency, duration, and intensity of aquatic heatwaves have increased over the
79 past century with climate change, a trend which is anticipated to continue (Oliver et al., 2018;
80 Woolway et al., 2021a, Woolway et al., 2022). Aquatic heatwaves may lower dissolved oxygen
81 (DO) concentrations directly through reduced gas solubility in warmer waters and indirectly
82 through increased ecosystem respiration (Tassone et al., 2021). Low DO concentrations and
83 thermal stress have been linked to coral bleaching, declines in kelp forests, and mass seagrass
84 mortality in marine systems (Wernberg et al., 2016; Hughes et al., 2017; Strydom et al., 2020).
85 Although heatwaves are well-studied in coastal and marine systems, research on the effects of
86 heatwaves in lakes and ponds is relatively nascent (Woolway et al., 2021a; Woolway et al.,
87 2022). As such, it remains unclear how aquatic heatwaves affect the structure and function of
88 lentic ecosystems, especially small, vegetated waterbodies.

89 Macrophytes may buffer small lakes and ponds from extreme temperatures by restricting
90 vertical heat transfer. Canopy-forming macrophyte beds attenuate incoming solar radiation and
91 dissipate wind-driven turbulence, thus limiting the depth of heat transport and creating a shallow
92 mixed surface layer separated from cooler bottom waters (Herb & Stefan, 2004; Andersen et al.,
93 2017a; Sand-Jensen et al., 2019). In short, macrophytes can induce thermal stratification in small
94 waterbodies, which is frequently observed (Holgerson et al., 2022). Stratification may isolate
95 bottom waters from rising surface temperatures during an aquatic heatwave and is expected to
96 restrict dissolved gas transport, producing a vertically heterogenous chemical environment
97 within macrophyte beds (Andersen et al., 2017b; Vilas et al., 2017). As a result, macrophyte
98 structure could also influence how DO responds to aquatic heatwaves. Rising surface water
99 temperatures during a heatwave are expected to lower DO concentrations; however, in
100 maintaining cooler bottom water temperatures, macrophyte-induced stratification may dampen
101 the effects of a heatwave on gas solubility and DO depletion. Altogether, macrophytes play a key
102 role in the thermal structure and DO environment of small waterbodies, potentially mediating the
103 effect of temporary, external stressors on these ecosystems' structure and function. However, an
104 explicit evaluation of how macrophyte-flow interactions alter ecosystem response to stressors is
105 lacking.

106 The abundance and growth of macrophytes in small waterbodies is driven by temperature
107 and the availability of light and nutrients (Carpenter & Lodge, 1986; Phillips et al., 2016),
108 making them sensitive to environmental stressors such as nutrient loading and heatwaves (Wu et
109 al., 2021). Warmer water temperatures can enhance macrophyte growth (Olesen & Madsen,
110 2001; Bertani et al., 2016; Hansson et al., 2020); however, high temperatures may also induce
111 heat stress, impeding reproduction (Li et al., 2017) or prompting senescence (Hao et al., 2018).

112 Similarly, nutrient loading may have either a positive or negative effect on macrophyte growth.
113 Although macrophytes can benefit from enhanced nutrient availability through foliar uptake
114 (DeMarte & Hartman, 1974; Twilley et al., 1977), nutrient loading can also increase algal
115 biomass and subsequently reduce light availability through shading (Scheffer et al., 1993; Short
116 et al., 1995). The effect of shading from phytoplankton-associated turbidity is expected to vary
117 across macrophyte growth forms, as submersed species are more vulnerable to light limitation
118 than emergent or floating-leaf species (Szabo et al., 2010). While the individual responses of
119 macrophytes to nutrient enrichment and heat stress have been established, it remains unclear how
120 macrophytes respond to simultaneous, compounding stressors and what the consequences are for
121 the thermal and chemical structure of small waterbodies.

122 We experimentally evaluated the interacting effects of episodic nutrient loading and
123 extreme heat on macrophyte biomass and the subsequent changes in the physicochemical
124 environment in two shallow, vegetated ponds. One of the ponds was pulsed with nutrients to
125 simulate typical storm-driven loading, while the other pond served as a reference. Both ponds
126 experienced a mid-summer heat event following the first experimental nutrient pulse. We
127 hypothesized that the combined stress of the nutrient addition and heat event would induce early
128 senescence of submersed macrophytes, reducing macrophyte biomass and subsequently altering
129 temperature and DO gradients. We asked (Q1) How does macrophyte biomass and canopy height
130 influence pond thermal structure, and does macrophyte presence control vertical heat transfer
131 during a heat event? (Q2) How does spatiotemporal variation in DO relate to macrophyte
132 biomass and canopy height, and does macrophyte presence alter the DO response to a heat
133 event? We evaluated these mechanisms using empirical data and developed a hydrodynamic
134 model of one-dimensional heat transport in vegetated waters. We predicted that greater

135 macrophyte biomass and canopy height in the reference pond would be associated with stronger
136 thermal stratification and greater differences in DO between surface and bottom waters. We also
137 predicted that macrophyte presence would maintain cooler temperatures and stable DO
138 concentrations during a heat event. Through the combination of our experiment, observations,
139 and modeling, we found that macrophytes structure spatial and seasonal variation in the thermal
140 and chemical environments of shallow waterbodies.

141

142 **MATERIALS AND METHODS**

143 *Study Site and Field Measurements*

144 We studied how macrophyte structure influences spatiotemporal variation in thermal and
145 DO dynamics of small waterbodies in two ponds at Iowa State University's Horticulture
146 Research Station (42.110005, -93.580454) during the summer of 2020. The ponds are
147 rectangular, relatively small (surface area 400 m², dimensions 10 m × 40 m; Figure 1), and
148 shallow (2 m maximum depth, 0.8 m mean depth). These waterbodies fit morphometric and
149 functional definitions of ponds (Richardson et al., 2022). The mean depth is below the theoretical
150 threshold distinguishing polymictic from stratifying waterbodies, such that the water column of
151 these ponds is expected to be well-mixed in the absence of internal structure (Kirillin &
152 Shatwell, 2016). Water balance is driven solely from precipitation and evaporation as there are
153 no surface inflows or outflows and the ponds are lined with bentonite to prevent groundwater
154 interactions. As there is effectively no watershed for these ponds, we controlled all external
155 nutrient inputs.

156 The macrophyte community was composed of longleaf pond weed (*Potamogeton*
157 *nodosus*; hereafter “floating-leaf”) and leafy pondweed (*P. foliosus*; hereafter “submerged”),

158 established from a natural seedbank. The typical phenology of these species follows a pattern of
159 spring emergence from over-wintering structures, growth throughout the summer, and then
160 senescence of the leaves and stems in early autumn. The ponds were initially filled with water
161 from the research station's irrigation reservoir to seed natural phytoplankton and zooplankton
162 communities. There were six total ponds at the site with varying fish communities as a part of a
163 larger experiment (Butts, 2023). Two of the ponds with the same fish community were selected for
164 this study. The ponds were stocked in early spring (15-45 days before measurements began) with
165 yellow perch (*Perca flavescens*) and bluegill (*Lepomis macrochirus*) at roughly 20 kg ha⁻¹ for
166 both species, which is within the range of densities for temperate waterbodies (Carlander, 1977;
167 Schneider, 1999).

168 External nutrient loading typical of a discrete storm event in this agricultural region was
169 simulated by adding nitrogen and phosphorus to one of the ponds (hereafter “nutrient addition
170 pond”) twice over the course of the summer. Nitrogen as ammonium nitrate (NH₄NO₃), and
171 phosphorus as sodium phosphate monobasic dihydrate ((NaH₂PO₄•(H₂O)₂) in a molar ratio of
172 24N:1P was mixed with pond water in a slurry and then distributed across the surface of the
173 pond. The first nutrient addition (day of year, DOY 176, 24 June 2020) raised the mass of
174 phosphorus in the water column by 3%, and the second addition (DOY 211, 29 July 2020) raised
175 the mass by 5%. These mass increases and stoichiometry are consistent with the expected
176 nutrient loading from a large precipitation event in an agricultural landscape (Vanni et al., 2001;
177 Lürding et al., 2018). Nutrients were not added to the other pond (hereafter “reference pond”). To
178 assess and compare the effects of the nutrient additions on algal biomass, we measured
179 chlorophyll-*a* concentrations daily at the deep site of each pond (Figure 1) using the Total Algae
180 sensor with the YSI ProDSS multiparameter sonde (Yellow Springs Instruments, Yellow
181 Springs, Ohio USA). The sensor logged continuously and was slowly lowered through the water

182 column. Surface concentrations of chlorophyll-*a* were averaged over 10-30 cm depth. To
183 minimize drift, the sensor was calibrated regularly against laboratory standards. Secchi depth
184 was also measured daily at the same location.

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

196 We quantified variation in macrophyte structure over space and time by monitoring
197 community composition, canopy height, and biomass from May to August (DOY 143-233).
198 Species presence-absence and canopy height were measured weekly at three sites from the deep
199 to shallow ends of the ponds (Figure 1). Canopy height was measured as the distance from the
200 sediment-water interface to the top of the canopy. Water depth was also measured at these points
201 so that canopy height could be expressed as a proportion of the water column. Macrophyte dry
202 biomass was sampled biweekly (6 sampling events) on the same grid of 18 sites in each pond
203 (Figure 1). Profiles of DO were always taken before biomass was sampled. To collect a biomass
204 sample, a two-side rake was lowered to the bottom and used to remove submersed and floating-

205 leaf macrophytes from an approximately 0.4 m² area (Mikulyuk et al., 2011). The above-
206 sediment tissue was collected from the rake to determine dry biomass by drying the material at
207 60°C to a constant mass before weighing.

208

209 *Hydrodynamic Modeling*

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

221 For boundary conditions, the model requires an initial water temperature profile from the
222 temperature chain at the deep site, timeseries of meteorological drivers (air temperature, relative
223 humidity, air pressure, wind speed, and short-wave radiation), and waterbody hypsography
224 (depth-area relationship). We obtained the necessary meteorological data at an hourly timestep
225 from a weather station located approximately 550 meters from the study ponds (ISU Soil
226 Moisture Network, 2021). Macrophyte data were incorporated into the model through weekly
227 measurements of canopy height at the deep site of the reference pond as well as biomass density

228 (g m⁻³) measurements taken every two weeks. We used the average biomass density from the two
229 sampling sites closest to the deep site temperature sensors.

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

241

242 *Statistical Methods*

243 An aquatic heatwave occurs when daily mean surface water temperatures exceed the
244 seasonal 90th percentile of historical measurements for at least 5 days, without dropping below
245 that threshold for two or more days (Hobday et al., 2016; Tassone et al., 2021). Since we did
246 not have long-term data on summer water temperatures in the experimental ponds, we elected to
247 use temperature data for the study period only, but a more stringent definition of a heat event
248 threshold (i.e., 95th percentile) for the same extended period (5 days). To determine the 95th
249 percentile of surface water temperature for each pond, we calculated the mean daily temperature
250 from loggers placed at 0 and 0.25 m at the middle at deep sites. A period with at least five days

251 above the 95th percentile threshold was classified as a sustained aquatic heat event (hereafter,
252 “heat event”).

253 To visualize thermal structure, we used linear interpolation at depths where temperature
254 measurements were missing. Only empirical measurements of water temperature were used in
255 the model and to calculate indices of pond thermal structure (i.e., thermocline depth).
256 Thermocline depth was calculated using the rLakeAnalyzer package. (Winslow et al., 2019). We
257 defined surface water temperature as the average from the 0 and 0.25m sensors. We used the
258 deepest sensors (2m at the middle and deep sites, 1.5m at the shallow sites) for measurements of
259 bottom water temperature. To compare macrophyte biomass between the study ponds, we
260 calculated a weighted average across the 18 sampling sites such that sites around the perimeter of
261 the pond (n=14) had the same weight as sites in the interior of the pond (n=4; Figure 1) so that
262 the mean value would not be biased by water depth.

263 The vertical distribution of DO saturation was either uniform, highest in the surface
264 waters, or highest in the bottom waters depending on the site and time of year (Figure S1).
265 Therefore, we focused on DO dynamics in the surface and bottom layers by averaging values
266 from 0-0.25 m and the sediment-water interface to 0.25 m above this depth, respectively. To
267 quantify spatial variation within each pond, we calculated the coefficient of variation for DO for
268 each sampling day, standardized by the pond-mean for that day. Temporal variation was
269 calculated based on the coefficient of variation in DO for each sampling site, so variation was
270 due to differences in DO over time, standardized by the summer mean DO saturation for that
271 site. A generalized additive model (GAM) was used to smooth the daily time series of
272 chlorophyll-*a* concentrations and weekly time series of DO saturation to visually highlight
273 differences in trends over time between the nutrient addition and reference pond. All statistical

274 analyses were completed in R version 4.1.2 (R Core Team 2021) using mgcv (Wood, 2017) in
275 addition to packages used for the hydrodynamic model.

276

277 **RESULTS**

278 *Environmental Stressors and Macrophyte Response*

279 Both ponds experienced a heat event from DOY 185-190 (3-8 July 2020, Figure 2A).
280 During this period, mean daily surface water temperatures exceeded the seasonal 95th percentile
281 in each pond (greater than 29.8 °C in the nutrient addition pond and 30.6 °C in the reference
282 pond). Maximum daily surface water temperatures ranged from 34.5-37.1 °C in the nutrient
283 addition pond and 33.7-38.6 °C in the reference pond during the heat event. Although both ponds
284 had brief periods of high surface water temperatures at other times, DOY 185-190 was the only
285 sustained period of high surface water temperatures (≥ 5 days). The heat event began 9 days after
286 the first nutrient pulse, creating multiple stressors (i.e., eutrophication and prolonged heat stress)
287 for macrophytes in the nutrient addition pond.

288 At the time of the first nutrient addition (DOY 176), chlorophyll-*a* concentrations were
289 the lowest in the amended pond (Figure 2B). Prior to and during the heat event chlorophyll-*a*
290 concentrations increased in the nutrient addition pond and continued to increase after the second
291 nutrient addition (DOY 211) until the early fall. At that time the reference pond (which did not
292 receive any nutrient input) had higher chlorophyll-*a* concentrations that declined prior to and
293 during the heat event, remaining low until early fall. Heightened chlorophyll-*a* concentrations in
294 the nutrient addition pond are indicative of higher algal biomass and thus phytoplankton-
295 associated turbidity, potentially reducing light availability for submerged macrophytes.

296 Spatiotemporal patterns in macrophyte biomass differed between the ponds and
297 macrophyte growth forms (Figure 2C, 3). The floating-leaf species was largely limited to a ring
298 around the shallower edges of the ponds where the leaves reached the surface. The submerged
299 species was present across a range of depths but reached the highest biomass in the deeper,
300 central region of each pond. In the reference pond, macrophytes followed expected phenology
301 with increasing biomass through to the late summer and then senescence of both growth forms
302 beginning in early autumn (Figure 2C, 3). However, in the nutrient addition pond, submerged
303 macrophytes growing in 1.75 to 2 m of water began to senesce following the initial nutrient
304 addition and heat event such that no plants were present in the deepest areas of the nutrient
305 addition pond nine days (DOY 199) after the heat event and the middle site of the pond 23 days
306 (DOY 213) after the heat event ended (Figure 3). Early senescence of submersed macrophytes in
307 the nutrient addition pond resulted in rapid declines in both biomass (Figure 2C) and canopy
308 height (Figure 4). Both submersed and floating-leaf plants persisted around the shallow edge of
309 the pond through the end of the summer. The temporal mismatch in the senescence of submersed
310 macrophytes between the reference and nutrient addition ponds was qualitatively consistent with
311 the other ponds at the site with differing fish communities, half of which received nutrient
312 additions and the other half were reference ecosystems (Figure S2).

313

314 ***Spatiotemporal Variation in Pond Thermal Structure***

315 Both ponds experienced intermittent thermal stratification and similar spatial patterns of
316 stratification from shallow to deep sites (Figure 4). Early in the summer, both ponds were cool
317 and isothermal. As surface waters warmed, the middle and deep sites of both ponds began to
318 stratify while the shallow sites remained well-mixed. At the center of the reference pond, a stable

319 thermocline developed between 1-1.5 m from DOY 155-161, and almost constantly between
320 DOY 168-210 (Figure S3A). In the nutrient addition pond, thermocline formation was
321 intermittent (DOY 157-159, 170-174, and 178-192; Figure S3B). When present, thermoclines
322 cooled and deepened slightly (0.05-0.1 m) during the night, indicating nighttime convective
323 cooling and mixing in the surface layer. During periods of thermal stratification, differences
324 between surface and bottom water temperatures ranged from 4.3 to 11.5 °C (mean 8.1°C) in the
325 reference pond and 4.5 to 10.2 °C (mean 7.3 °C) in the nutrient addition pond.

326 During times of similar macrophyte coverage, either early or late in the season (Figure 4),
327 the thermal structure between the two ponds was similar. For example, both ponds experienced a
328 strong mixing event in late summer (DOY 216-220), driven by cooler air temperatures and
329 higher wind speeds (Figure 5) and facilitated by declining macrophyte structure (Figure 4).
330 Similar meteorological conditions occurred in early summer and resulted in some water cooling
331 (e.g., DOY 164, 175); however, the ponds did not mix as fully during these events, likely due to
332 the macrophyte structure present in both ponds in early summer (Figure 4).

333 Although the overall seasonal pattern in thermal structure was similar between the ponds,
334 there were brief dissimilarities related to differences in the timing of macrophyte senescence
335 (Figure 4). There was a prolonged period of stable thermal stratification from DOY 180-210 in
336 the reference pond, with only brief instability between DOY 198-199. A stable thermocline set-
337 up in the nutrient addition pond around the same time (DOY 178); however, stratification began
338 to break down approximately 18 days earlier (DOY 192) than in the reference pond (Figure S3).
339 The timing of destratification in the nutrient addition pond coincides with declines in macrophyte
340 biomass (Figure 3) and canopy height (Figure 4) in the center of the pond. Destratification in the
341 reference pond also followed patterns of biomass senescence and loss of canopy height, although

342 this did not occur until later in the summer in this pond. In both ponds, there was strong temporal
343 coherence between the timing of destratification and macrophyte senescence, with the difference
344 in timing from typical phenology likely spurred by the compounding stress of eutrophication and
345 a heat event in the nutrient addition pond. The declines in canopy cover and biomass were
346 associated with increased thermal mixing and warmer bottom water temperatures, especially in
347 the nutrient addition pond.

348 During the aquatic heat event (DOY 185-190), both ponds experienced a similar increase
349 in surface water temperatures (Figure 2A, Figure 4). However, there was a stark difference in the
350 response of bottom water temperatures at the deep and middle sites between the ponds (Figure
351 S4). During the heat event, bottom water temperature remained consistent at both the deep and
352 middle sites of the reference pond. The daily mean bottom water temperature increased by only
353 0.2 °C at both sites. In contrast, bottom water temperatures in the nutrient addition pond
354 increased by 1.3 °C and 1.4 °C at the deep and middle sites, respectively. At the shallow site in
355 both ponds bottom water temperatures increased 1-1.3 °C during the heat event. The different
356 response of bottom water temperature to the heat event related to differences in macrophyte
357 biomass between the two ponds. At the time of the heat event, macrophyte biomass (Figure 2C)
358 and canopy height (Figure 4) were greater in the reference pond as submersed macrophytes in
359 the center of the nutrient addition pond had begun to senesce (Figure 3). Overall, bottom waters
360 in the nutrient addition pond warmed during the heat event while cooler temperatures were
361 maintained in the reference pond, despite significant warming in the surface layers.

362

363 *Spatiotemporal Variation in Dissolved Oxygen*

364 In both ponds, DO varied across sites, vertically in the water column, and over time. Weekly DO
365 profiles captured broad seasonal trends in surface DO saturation as similar temporal patterns
366 were observed between weekly and high-frequency values (Figure 2D, S5A). Early in the
367 summer, both ponds were supersaturated in DO in the surface waters (Figure 2D), and mean
368 surface saturation was similar between the ponds before the initial nutrient addition. In the
369 reference pond, DO remained stable until the heat event (DOY 185-190), after which saturation
370 declined and generally remained below saturation for the remainder of the study period. In
371 contrast, mean surface DO saturation in the nutrient addition pond declined immediately
372 following the first nutrient addition (DOY 176) and remained below saturation for the rest of the
373 summer. Over a one-week period during and after the heat event (DOY 186-192), DO saturation
374 fell 22.4% and 50.4% in the nutrient addition and reference ponds, respectively. Bottom water
375 DO saturation mirrored the seasonal pattern in surface DO (Figure S5B). For both ponds the
376 difference in DO saturation between surface and bottom waters was most pronounced early in
377 the season (reference pond range of difference from -87.1 to 23.3%; nutrient addition pond range
378 -56.8 to 58.8%). Surface and bottom water DO saturation became more similar throughout the
379 summer, especially following the heat event (reference pond range of difference -2.1 to 17.0%;
380 nutrient addition pond range -0.3 to 12.2%; Figure S5C).

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

386 biomass, with lower DO following the nutrient addition; however, the heat event had the most
387 pronounced effect on seasonal DO saturation (Figure 6).

388

389 *Mechanisms of Macrophyte Control of Pond Hydrodynamics*

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

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

409 macrophytes were entirely absent from the pond, the model predicted complete water column
410 mixing and much warmer temperatures throughout the water column, especially following the
411 aquatic heat event (Figure 7B). This is consistent with the prediction of polymixis for these
412 ponds based on bathymetry (Kirillin & Shatwell, 2016). When macrophytes were present, the
413 increase in bottom water temperatures in the simulation over the course of the aquatic heat event
414 was negligible (0.01 °C), and bottom temperatures remained below 25 °C for the remainder of
415 the summer (maximum temperature 23.3 °C; Figure 7A, C). However, in the no-macrophyte
416 scenario bottom water temperatures increased by 1.92 °C during the heat event and had much
417 higher maximum water temperatures later in the summer (33.4 °C; Figure 7B, C). The model
418 simulations demonstrate that macrophyte structure increases thermal stratification and promotes
419 cooler bottom water temperatures, even during an extreme heat event.

420

421 **DISCUSSION**

422 We applied an ecosystem experiment and complimentary hydrodynamic model to test
423 macrophyte response to compounding stressors (i.e., stochastic external nutrient loading and an
424 extreme heat event) and evaluate the consequences for the physicochemical environment in two
425 shallow, vegetated ponds. The type and severity of environmental stressors differed between the
426 study ponds, producing distinct seasonal trajectories in macrophyte assembly and coverage. In
427 the reference pond, both submersed and floating-leaf macrophytes followed the expected
428 phenology of spring emergence, summer growth, and then senescence beginning in late summer.
429 This natural phenology may be attributed to the lack of severe physiological stressors in the
430 reference pond. Although this pond did experience a sustained heat event, dense macrophyte
431 beds limited the depth of water column warming and maintained stable bottom water

432 temperatures. In contrast, submersed macrophytes in the nutrient addition pond began to senesce
433 in mid-summer after the first experimental nutrient pulse and continuing after the heat event.
434 This early senescence was limited to submersed macrophytes in the deeper, central region of the
435 pond, suggesting that light limitation from phytoplankton shading may have been a contributing
436 stressor (Scheffer et al., 1993; Short et al., 1995; Phillips et al., 2016). Unlike the reference pond,
437 chlorophyll-*a* concentrations continued to rise in the nutrient addition pond, maintaining any
438 light limitation stress. The combination of some light limitation with more acute heat stress due
439 to higher bottom temperatures during the heat event may have been the catalyst for the observed
440 early senescence trajectory in the nutrient addition pond. Qualitatively, the same declining
441 trajectory in macrophyte biomass was observed in the other ponds on site that received nutrient
442 additions but had higher resilience to the nutrient pulses (Butts, 2023), supporting our hypothesis
443 that the combined stressors of eutrophication and intense heat led to the early decline. Regardless
444 of the driver, the observed early senescence of submersed macrophytes in the nutrient addition
445 pond altered the thermal environment and DO saturation in relation to the reference pond.

446

447 *Thermal Structure*

448 On a seasonal scale, we observed strong temporal coherence between macrophyte growth
449 and senescence and pond thermal structure. Both ponds experienced intermittent thermal
450 stratification in early summer, once macrophytes had grown to fill most of the water column.
451 This threshold is consistent with the expectation that canopy-forming macrophyte beds may
452 induce diurnal stratification once they occupy at least 50% of the water column (Vilas et al.,
453 2017) as well as observed stratification in many small waterbodies (Holgerson et al., 2022). By
454 mid-summer we observed strong thermal stratification, characterized by large differences

455 (around 10°C) between surface and bottom water temperatures, comparable to stratification
456 reported in other small, vegetated ponds (Andersen et al., 2017a; Vilas et al., 2017). However, in
457 other studies, macrophyte-induced stratification followed diel cycles, with daytime stratification
458 and nighttime overturn due to convective mixing (Martinsen et al., 2019; Sand-Jensen et al.,
459 2019), which can move through macrophyte structure more readily than external turbulent
460 mixing (Herb & Stefan, 2005; Andersen et al., 2017b). Although we observed nighttime
461 convective cooling and mixing in the surface of the ponds during periods of macrophyte-induced
462 stratification, convective cooling did not mix the entire water column. Stable thermal
463 stratification began to break down in both study ponds during macrophyte senescence, the timing
464 of which varied between ponds. Declining macrophyte canopy height and biomass removed
465 structural barriers that had previously attenuated incoming solar radiation and dissipated wind-
466 driven turbulent mixing (Herb & Stefan, 2004), allowing the warm, mixed surface layer to
467 deepen and eventually mix the entire water column.

468 Macrophyte presence mediated how pond thermal structure responded to a sustained heat
469 event by restricting vertical heat transfer and maintaining cooler bottom water temperatures.
470 Both study ponds saw high surface water temperatures during the heat event. However, as
471 anticipated, the depth of water column warming and change in bottom water temperature differed
472 between the reference and nutrient addition ponds (+0.2°C vs. 1.3°C, respectively) due to
473 differences in macrophyte coverage and canopy height (75% of the water column vs. 0%,
474 respectively). Macrophyte structure clearly played a pivotal role in mediating the depth and
475 degree of water column warming during the aquatic heat event. Nevertheless, thermal structure
476 in small waterbodies arises from complex interactions between waterbody features and external
477 forces (Herb & Stefan, 2004; Branco & Torgersen, 2009, Holgerson et al., 2022). To explicitly

478 test the mechanisms underpinning the differences we observed between the study ponds, we used
479 a hydrodynamic model for heat transport in vegetated waterbodies and simulated water column
480 thermal structure during a heat event under contrasting scenarios of macrophyte density.

481 The model results support our field observations and demonstrate that macrophytes
482 structure the thermal environment in shallow ponds through vertical attenuation of both light and
483 turbulent kinetic energy. Simulations of seasonal patterns in pond thermal structure under
484 variable macrophyte density revealed that macrophyte presence reduces the depth and degree of
485 water column warming both during and after an aquatic heat event and maintains cooler bottom
486 water temperatures throughout the course of the summer. Our experimental observations and
487 model simulations provide evidence that macrophytes can buffer small waterbodies from brief
488 aquatic heatwaves, particularly the vertical extent of warming in the water column. As aquatic
489 heatwaves become more widespread and frequent, heatwaves in small waterbodies are expected
490 to be shorter but more intense compared to larger waters (i.e., due to lower thermal inertia;
491 Woolway et al., 2021a; Polazzo et al., 2021; Woolway et al., 2022). Our findings suggest that
492 macrophytes can mediate the intensity of brief aquatic heatwaves in small lakes and ponds. As
493 such, managing for robust macrophyte communities is a potential tool for mitigating heatwave
494 effects in small waterbodies and promoting thermal habitat refugia to support other aquatic
495 organisms (Carpenter & Lodge, 1986; Till et al., 2019).

496 Our hydrodynamic model enhances a conceptual and analytical framework for
497 incorporating biological structure into one-dimensional heat transport models (Herb & Stefan,
498 2004). Overall, there was good agreement between the model predictions and observed
499 temperature profiles. The largest discrepancies between the model output and field data occurred
500 in early summer when the model underpredicted temperature. The early summer period

501 coincides with rapid growth in macrophyte biomass and canopy height. It is possible that the
502 temporal resolution of our weekly canopy height measurements and biweekly biomass
503 collections was insufficient to accurately capture biological dynamics in the pond for the model
504 input. There was also a one-day period of high air temperatures (DOY 155) early in the summer
505 that caused a rapid change in water temperature. Other one-dimensional heat transport models
506 (i.e., Simstrat; General Ocean Turbulence Model, GOTM; General Lake Model, GLM) have
507 greater error during extreme meteorological events, including atmospheric heatwaves (Mesman
508 et al., 2020), which may be the case with our model as well during this early-season, rapid warm-
509 up. Our model would benefit from additional testing across a variety of vegetated waterbodies to
510 inform best practices for applying this approach in other waters. However, our findings support
511 the need to incorporate vegetation dynamics into hydrodynamic models, and our modeling
512 approach provides an example of how to do so.

513

514 *Dissolved Oxygen*

515 There was little spatial variability in DO saturation, contrary to observations in other
516 vegetated waterbodies (Andersen et al., 2017a; Vilas et al., 2017). Most of the DO profiles were
517 uniform with depth, suggesting that the balance of processes releasing and consuming oxygen
518 (e.g., gross primary production versus respiration) were similar throughout the ponds or that DO
519 was able to readily equilibrate. Temporal variation in surface and bottom water DO saturation
520 were connected to the compounding environmental stressors, beginning with the initial
521 experimental nutrient addition. Following the first nutrient pulse, DO remained stable in the
522 reference pond but declined below 100% saturation in the experimental pond. The observed
523 decline in DO suggests that ecosystem respiration was stimulated by the added nutrients,

524 outweighing any increases in gross primary production. Stress-induced macrophyte senescence
525 likely contributed to this pattern through reduced macrophyte photosynthesis and increased
526 macrophyte decomposition. Although we observed differences in DO saturation between the
527 ponds after the nutrient addition, the most pronounced temporal shift in DO occurred following
528 the heat event.

529 During the heat event, DO saturation fell sharply in both ponds and remained depressed
530 through the remainder of the study period. We had expected macrophyte structure to maintain
531 stable DO saturation in the bottom waters of the reference pond; however, both ponds responded
532 similarly, regardless of the differences in macrophyte coverage and thermal structure. In addition
533 to lower oxygen solubility in warmer water, organic carbon mineralization in sediments is
534 strongly controlled by temperature (Gudasz et al., 2010; Cardoso et al., 2014) and the heat event
535 likely spurred microbial respiration, which has been observed during heatwaves in both
536 freshwater and marine ecosystems. (Ali et al., 2016; Brauko et al., 2020; Tassone et al., 2021).
537 Reduced DO availability has consequences for nutrient and carbon cycling as well as greenhouse
538 gas production. For example, anoxia within the sediment profile can mobilize phosphorus from
539 redox-sensitive minerals (Orihel et al., 2017). Anoxic conditions and increased organic matter
540 availability (e.g., from macrophyte necromass) also promote denitrification and methanogenesis,
541 two microbial processes that produce potent greenhouse gases (Bastviken et al., 2005; Fonseca et
542 al., 2017; Ming et al., 2022). Although both ponds likely experienced shifts in other elemental
543 cycles as a result of reduced DO saturation during the aquatic heat event, the consequences for
544 greenhouse gas production were likely more pronounced in the nutrient addition pond due to
545 early macrophyte senescence increasing organic matter inputs to the sediments. This difference

546 between the study ponds highlights the complex interactions that may arise from compounding
547 stressors in lentic ecosystems.

548

549 *Conclusions*

550 Aquatic heatwaves are becoming more frequent and are likely to co-occur with
551 chlorophyll-*a* maxima in lakes and ponds (i.e., algal blooms; Woolway et al., 2021b). As such,
552 there is a growing need to understand how lentic ecosystems will respond to compound, extreme
553 events to inform water resource and biodiversity management. Our findings illuminate the role
554 that macrophyte-hydrodynamic interactions play in mediating aquatic ecosystem response to
555 multiple stressors. Overall, dense macrophyte structure can buffer small waterbodies from brief
556 aquatic heat events by restricting vertical heat transport. However, declines in DO saturation are
557 expected during heat events, regardless of macrophyte coverage. Our findings also reinforce that
558 macrophyte communities in small lakes and ponds are vulnerable to nutrient loading and
559 temperature extremes. Given the prevalence and importance of small waterbodies for
560 biodiversity and biogeochemical processing, it is essential that we continue to build our
561 mechanistic understanding of the effects of compound, extreme events in these ecosystems and
562 the role that macrophytes can play in ameliorating stressors.

563

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576

577 *Author Contributions*

578 Ellen Albright and Grace Wilkinson designed the study. Ellen Albright collected and analyzed
579 the field data. Robert Ladwig and Ellen Albright designed and ran the hydrodynamic model.
580 Ellen Albright wrote the first draft. Grace Wilkinson provided substantial feedback on the first
581 draft. All authors assisted with later revisions.

582

583 *Data Availability Statement*

584 The data supporting the conclusions and the analysis code are available in the Github repository
585 https://github.com/AlbrightE/pond_physics_2022. Following manuscript acceptance, the Github
586 repository will be archived on Zenodo, and the data will be published on the Environmental Data
587 Initiative's online repository under a Creative Commons Attribution license (CC-BY).

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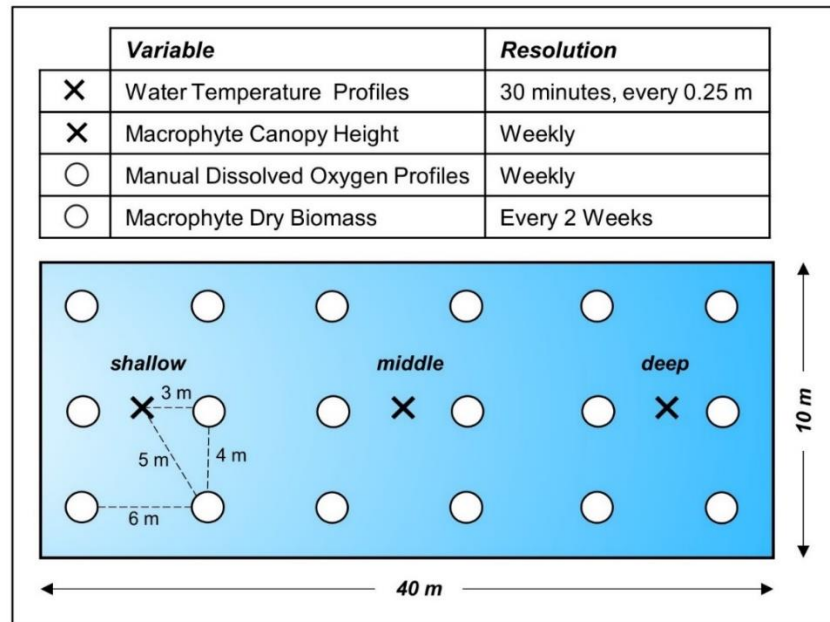
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782 **TABLES**783 **Table 1.** *Temporal versus spatial variation in DO saturation*

<i>Layer</i>	<i>Pond</i>	<i>Temporal Variation in DO Saturation</i>		<i>Spatial Variation in DO Saturation</i>	
		<i>CV (%) Mean</i>	<i>CV (%) Range</i>	<i>CV (%) Mean</i>	<i>CV (%) Range</i>
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

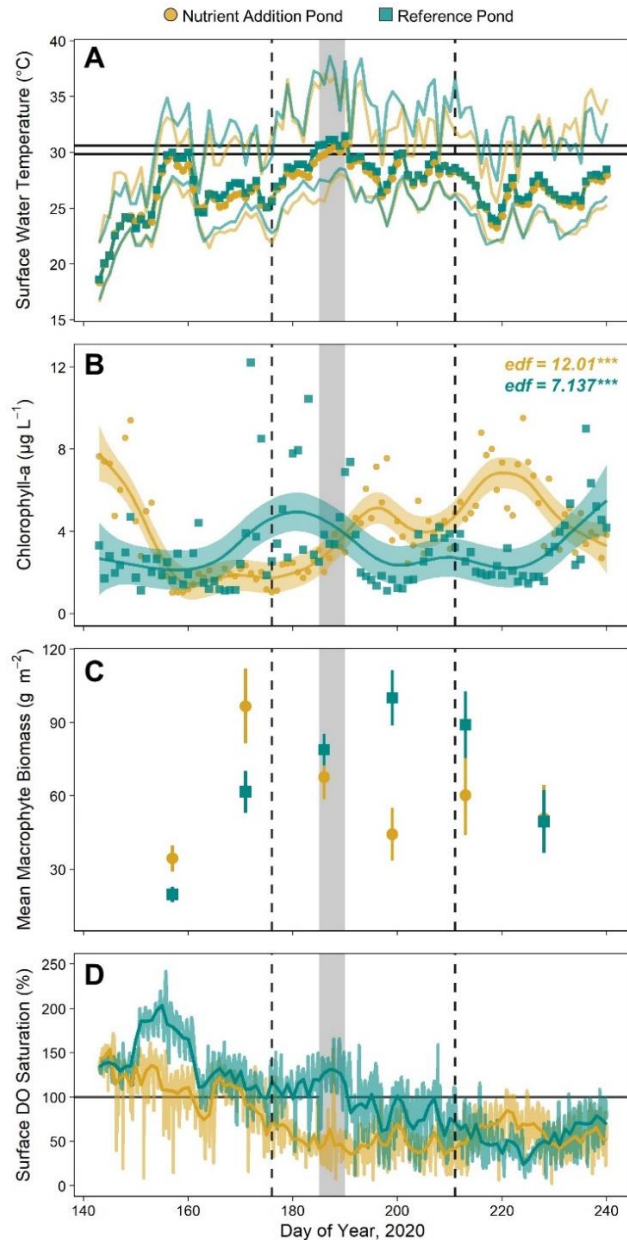
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785 **FIGURES**

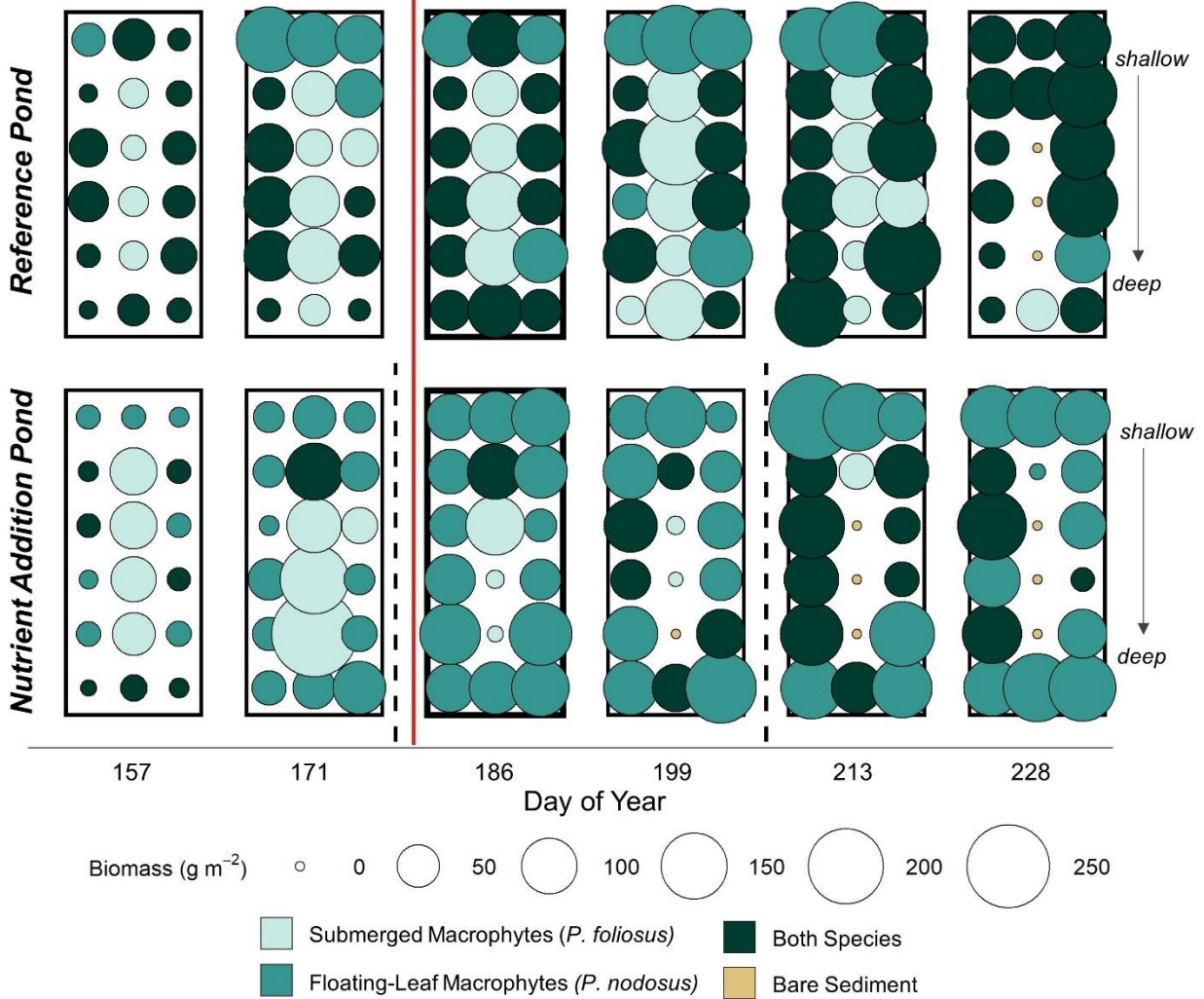


786

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

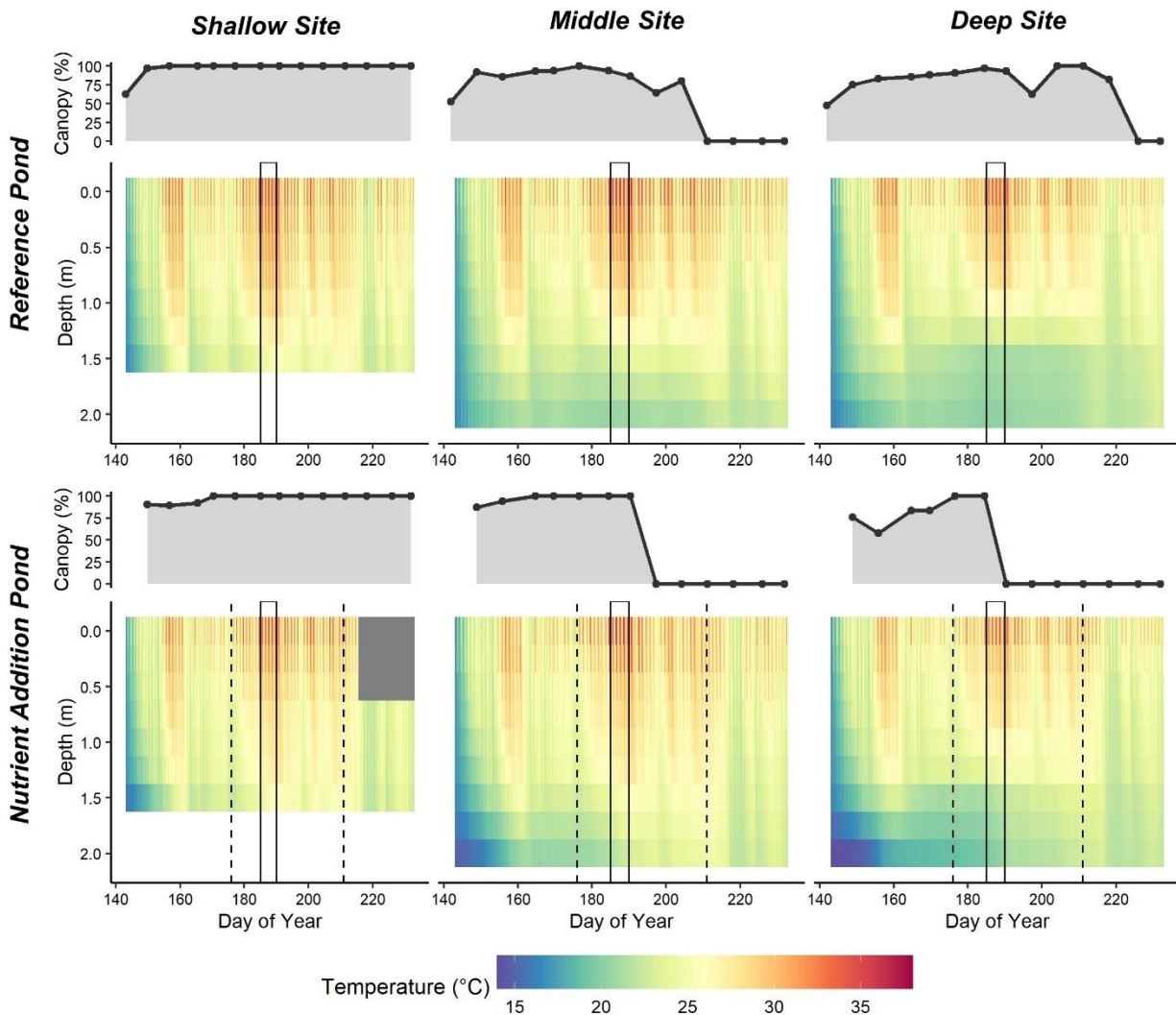


788
 789 **Figure 2.** (A) Timeseries of daily surface (0-0.25 m) water temperature. Points represent daily
 790 means while lighter lines above and below mark daily maximum and minimum temperatures.
 791 Surface sensors at the deep and middle sites were used for pond averages. The 95th percentile
 792 surface water temperature thresholds are noted in horizontal lines for the nutrient addition
 793 (29.8°C) and reference (30.6°C) ponds. (B) Timeseries of surface (0.1-0.3 m) chlorophyll-*a*
 794 concentrations. Values from each pond were GAM-fit to highlight temporal trends. The asterisks
 795 next to the effective degrees of freedom (edf) correspond to the estimated p-value, with ***
 796 denoting $p < 0.001$. (C) Time series of mean macrophyte biomass in each pond. Error bars mark
 797 the standard error of the mean. (D) High frequency surface (0.25 m) dissolved oxygen (DO)
 798 saturation at the deep site of each pond over the study period. The daily mean is plotted in a
 799 darker line. The horizontal grey line marks 100% saturation. Across all panels, dashed lines mark
 800 the timing of nutrient additions (DOY 176, 211) and the grey box highlights the aquatic heat
 801 event (DOY 185-190).

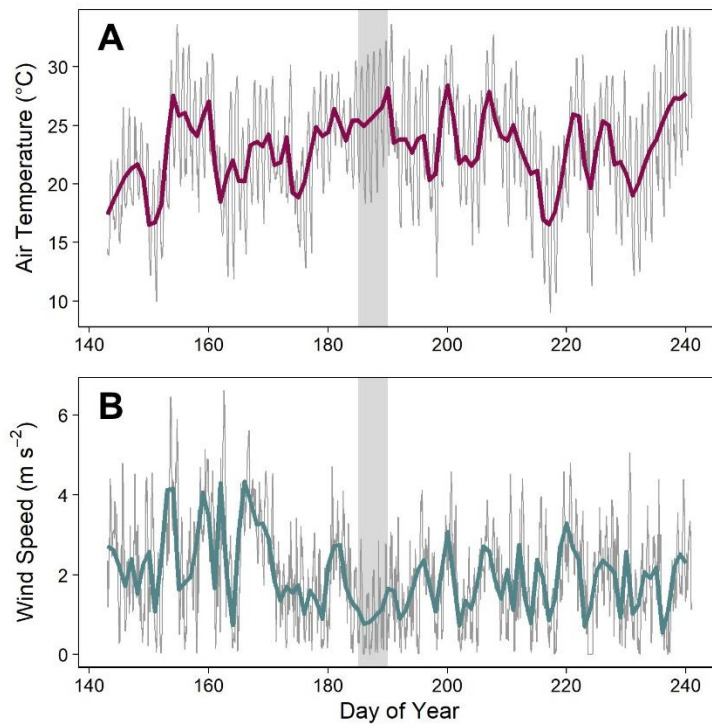


803

804 **Figure 3.** Spatial and temporal variation in macrophyte dry biomass density and species
 805 assemblage in the reference (top) and nutrient addition (bottom) ponds. The start of the aquatic
 806 heat event on DOY 185 is noted with the solid, red line. Dashed lines mark nutrient additions on
 807 DOY 176 and 211.

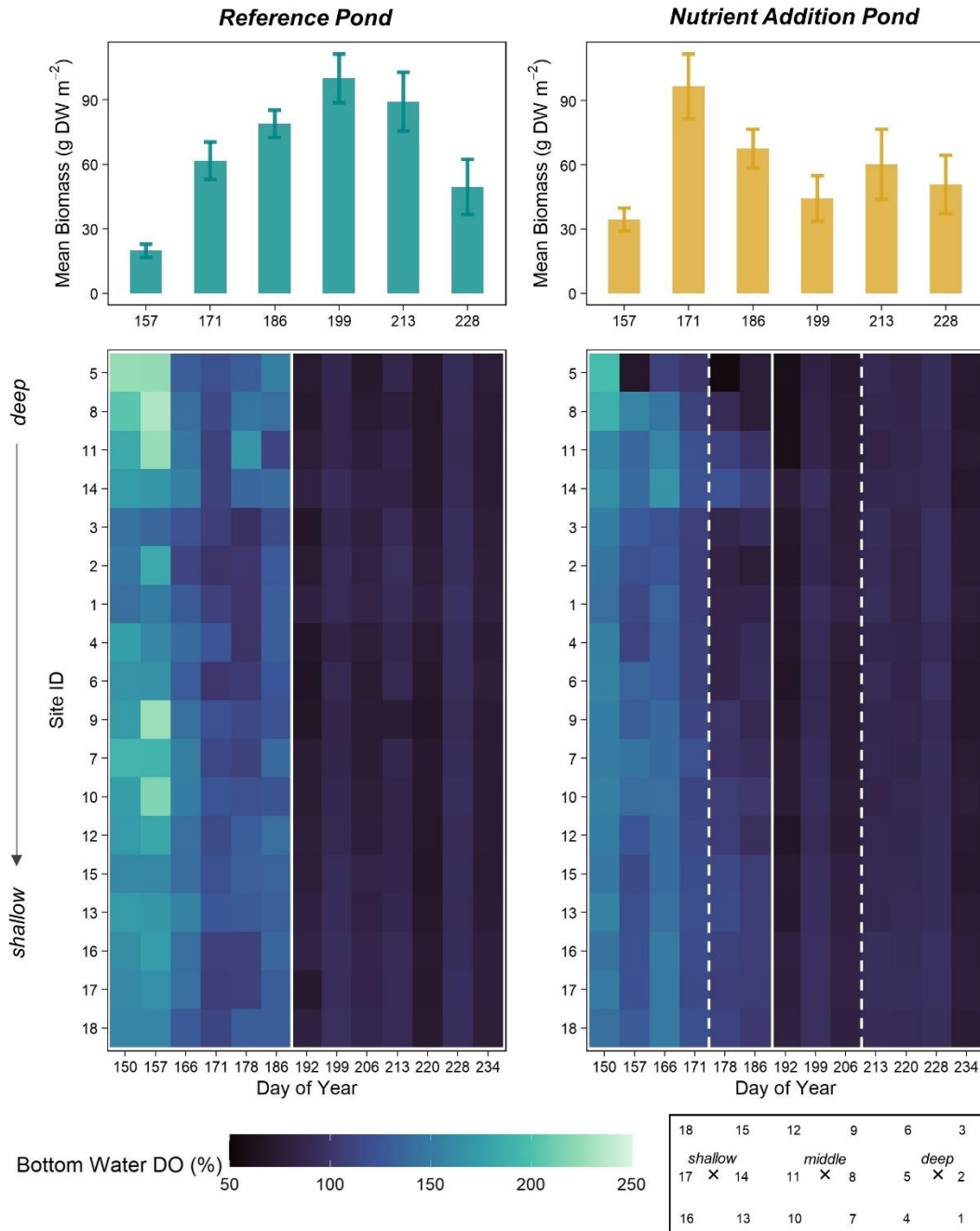


808
 809 **Figure 4.** Spatial and temporal variation in macrophyte canopy height and temperature profiles.
 810 Canopy height is plotted as a percent of the water column height. Boxes mark an aquatic heat
 811 event from DOY 185-190 while dashed lines indicate nutrient additions on DOY 176 and 211.
 812 High frequency water temperature loggers were placed every 0.25 m of the water column up to
 813 1.5 m and then every 0.5 m to the bottom. Temperature values at 1.75 m were interpolated for
 814 the sake of visualization. Several surface sensors (0-0.5 m) from the shallow site of the nutrient
 815 addition pond lost power following DOY 215 (grey box). An initial canopy height measurement
 816 is missing from the nutrient addition pond because the water was not sufficiently clear to see and
 817 measure the top of the canopy. The canopy height was less than 1 m or 50-75% of the water
 818 column at this time.



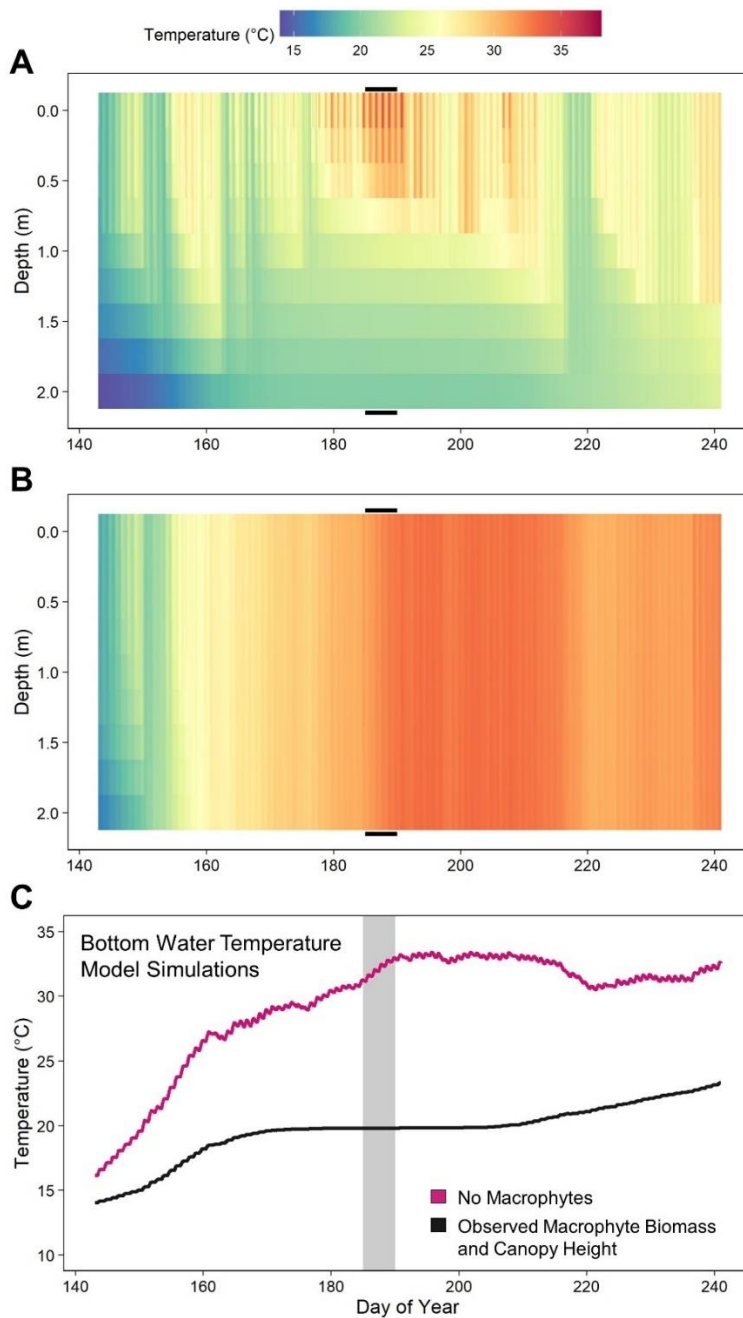
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822

Figure 5. Air temperature (A) and windspeed (B) over the study period. Daily mean values are plotted in thicker lines over the hourly time series. The light grey box marks the aquatic heat event from DOY 185-190.



823

824 **Figure 6.** Pond averaged macrophyte dry biomass density over time (top panels) and
 825 spatiotemporal variation in bottom water dissolved oxygen (DO) saturation (bottom panels).
 826 Error bars on the macrophyte biomass bar plots represent the standard error of the mean. DO
 827 saturation values are reported by sampling site and date. Sampling sites are ordered
 828 approximately from the deepest to most shallow locations. A simplified map of sampling site
 829 location is displayed in the lower righthand corner of the figure. Data from sampling events
 830 before and after the end of the heat event (DOY 185-190) are separated with a solid white line.
 831 Dashed white lines indicate nutrient additions on DOY 176 and 211.



833
 834 **Figure 7.** Model simulation results contrasting a no macrophyte scenario with the observed
 835 macrophyte coverage from the reference pond. Predicted temperature profiles with (A) observed
 836 macrophyte biomass and canopy height versus (B) no macrophytes present in the reference pond.
 837 Black, horizontal lines mark an aquatic heat event (DOY 185-190). (C) Timeseries of predicted
 838 daily mean bottom water (2m) temperatures with no macrophytes versus observed macrophyte
 839 biomass and canopy height.

841 **Appendix S1**

842

843 **Influence of macrophytes on stratification and dissolved oxygen dynamics in ponds**

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848

849 **Introduction**

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

854 Figure S1 provides an example of individual dissolved oxygen (DO) saturation profiles to
855 illustrate there was no evidence of mid-depth DO maxima. Figure S2 shows a qualitative
856 comparison in the change in macrophyte biomass during the summer at the central sampling sites
857 across all the experimental ponds at the Iowa State University Horticulture Research Station. We
858 provide time series of estimated thermocline depth (Figure S3) for each pond as well as time
859 series of bottom water temperatures at each sampling site on each pond (Figure S4). We further
860 provide supplementary visualizations for our DO data, specifically time series of surface and
861 bottom water DO saturation and the difference between surface and bottom water DO (Figure
862 S5).

863

864 **Model Equations and Approach**

865 *Model description*

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

869

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

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

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

875

876 Internal heat generation is implemented as:

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

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

880

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

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

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

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

886 where H_{lw} is the incoming long-wave radiation, H_{lwr} is emitted radiation from the water column,
887 H_v is the latent heat flux, and H_c is the sensible heat flux. Implementations to estimate the
888 respective heat fluxes were taken from Livingstone and Imboden (1989) and Goudsmit et al.
889 (2002). Latent and sensible heat fluxes were calculated using the atmospheric stability
890 algorithms from Verburg and Antenucci (2010).

891

892 The lower, sediment boundary condition was prescribed as:

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

894

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

898

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

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

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

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

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

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

909 and N^2 is the squared buoyancy frequency:

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

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

912

913

914 *(b) Mixed layer depth*

915 In the third step, we quantified the depth where the amount of external kinetic energy by wind
916 shear stress equals the internal potential energy of the water column. Up to this mixed layer
917 depth, z_{ml} , adjacent layers are subsequently mixed to account for a wind shear stress acting over
918 the vertical water column. Here, the kinetic energy KE is described as:

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

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

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

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

925

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

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

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

930

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

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

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

935

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

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

940

941

942

943 (c) *Convective overturn*

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

947

948 *Appendix References*

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976

977

978 **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

979

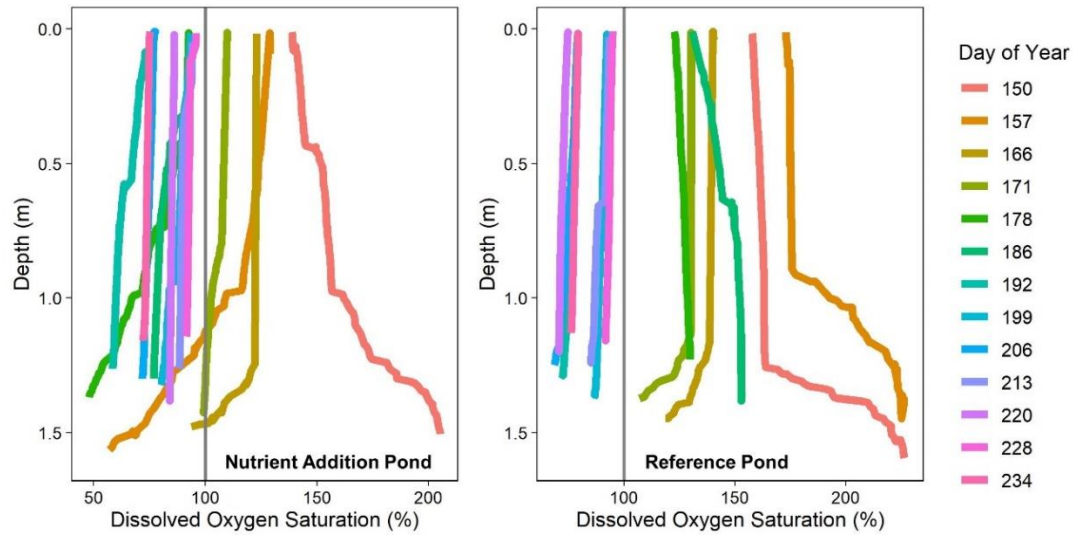
980 **Table S2.** *Input macrophyte data for model simulations*

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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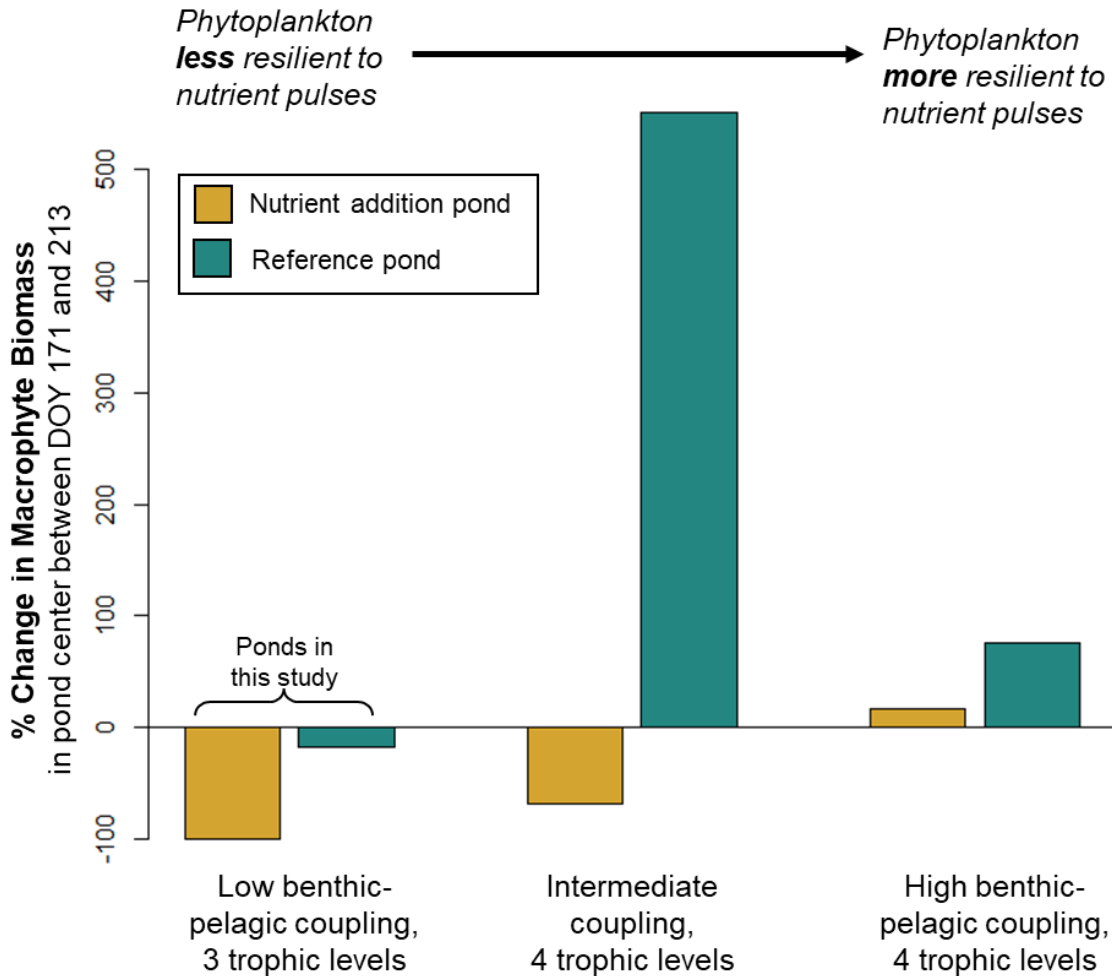
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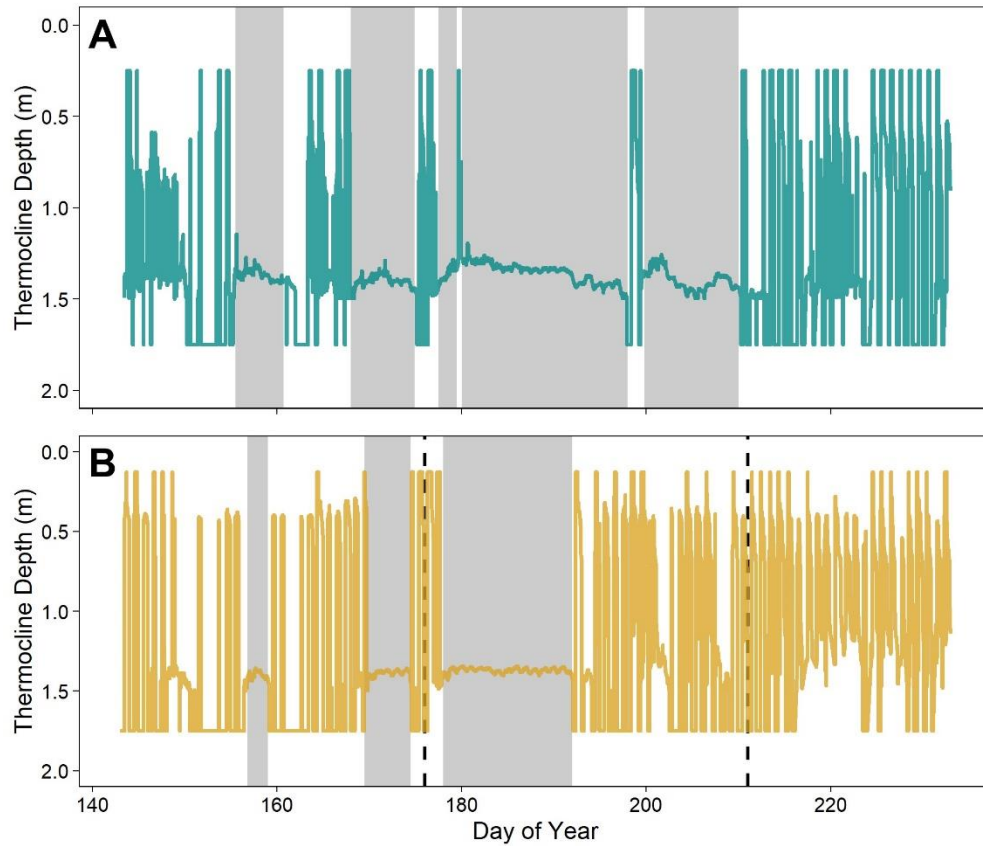
985 **Figure S1.** Examples of individual dissolved oxygen (DO) saturation profiles near the deep site
 986 of each study pond. Profiles are color-coded by the day of year (DOY) of the sampling event.
 987 These profiles were taken in the deeper, central region of the ponds. The vertical grey line
 988 highlights 100 percent saturation. The vertical distribution of DO saturation was either uniform,
 989 highest in the surface waters, or highest in the bottom waters depending on the time of year.
 990 There was no evidence of mid-depth DO maxima.

991



992

993 **Figure S2.** Change in macrophyte biomass during the summer at the central sampling sites in all
 994 the experimental ponds at the Iowa State University Horticulture Research Station. A total of six
 995 ponds are located at the station and were used in an experiment examining the role of food web
 996 structure (i.e., benthic-pelagic coupling) on phytoplankton resilience to nutrient pulses (Butts,
 997 2023). One pond in each food web pair received a nutrient addition on DOY 176 and 211, as
 998 described in the main text. The two ponds with low benthic-pelagic coupling food webs were
 999 instrumented with high frequency sensors and used in this study. Less frequent macrophyte
 1000 biomass measurements were taken in the other four ponds, all of which were more resilient to
 1001 nutrient additions (i.e., less susceptible to light limitation stress). All ponds experienced the heat
 1002 event from DOY 185-190. The pattern of early macrophyte senescence was most pronounced in
 1003 this low coupling pond that received the nutrient additions, but also occurred in the intermediate
 1004 coupling pond. In the high coupling pond, macrophyte biomass increased slightly in the nutrient
 1005 addition pond (there was minimal increase in phytoplankton biomass in this treatment; Butts,
 1006 2023), but the increase was small, especially compared to the reference ecosystem during the
 1007 same period. In summary, the ponds that received nutrient additions had a similar trajectory of
 1008 declining or lower biomass earlier in the summer as compared to the reference ponds.



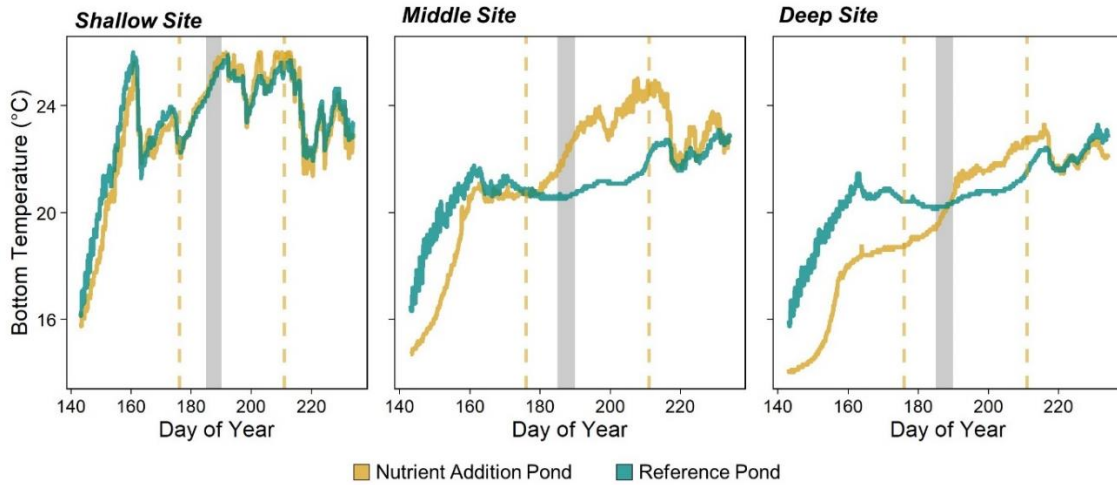
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1011 **Figure S3.** Estimated thermocline depth by pond. Periods of a stable thermocline for the
 1012 reference (A) and nutrient addition (B) ponds are noted in the grey shaded boxes. Nutrient
 1013 additions are noted with dashed black lines. Estimated thermocline values oscillating rapidly
 1014 between the near surface and bottom waters were not considered time of stable stratification. A
 1015 stable thermocline between 1-1.5 m depth was observed in the reference pond between DOY
 1016 155-161, and almost constantly between DOY 168-210, while the thermocline in the nutrient
 1017 addition pond was more intermittent (DOY 157-159, 170-174, and 178-192).

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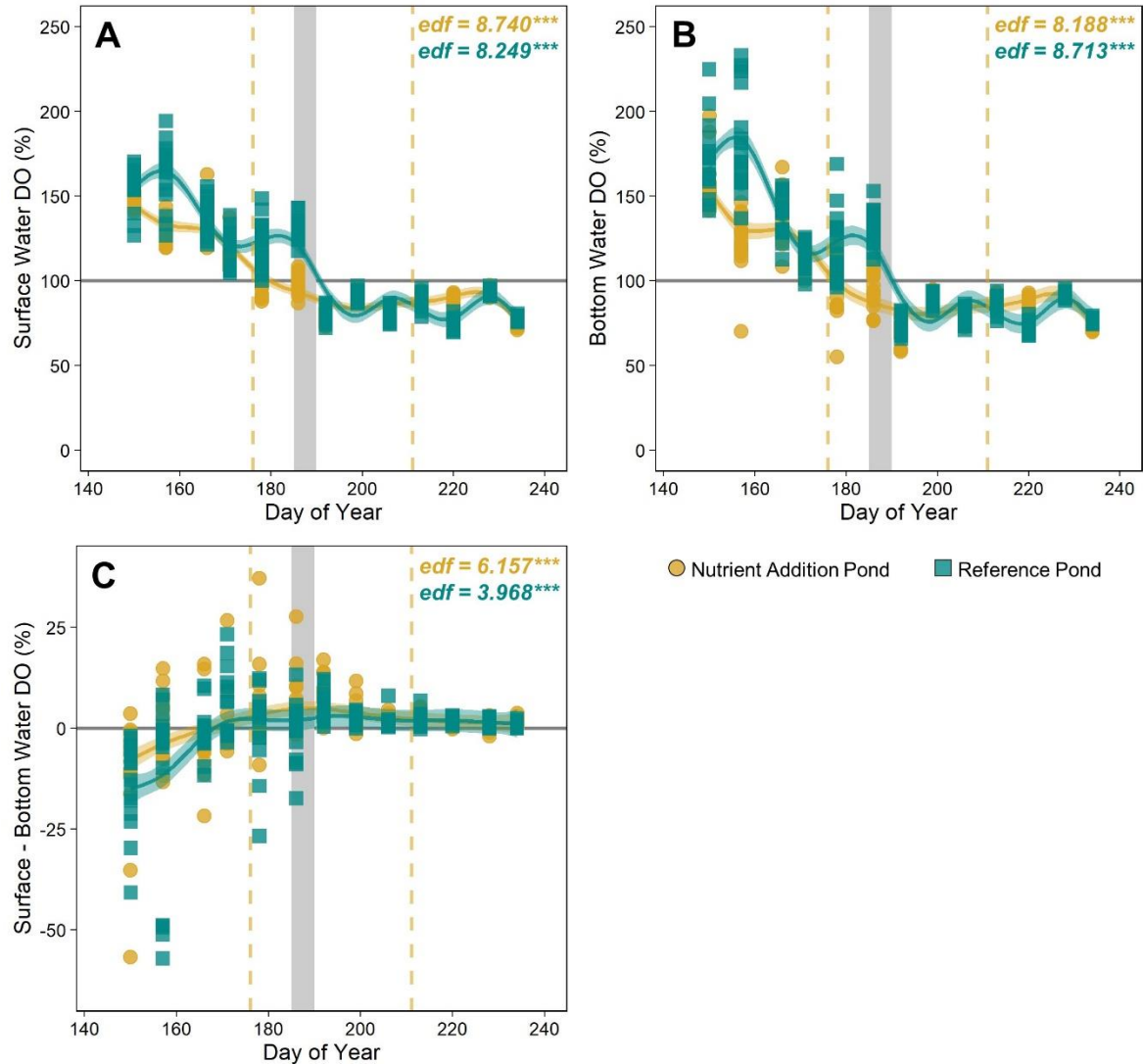
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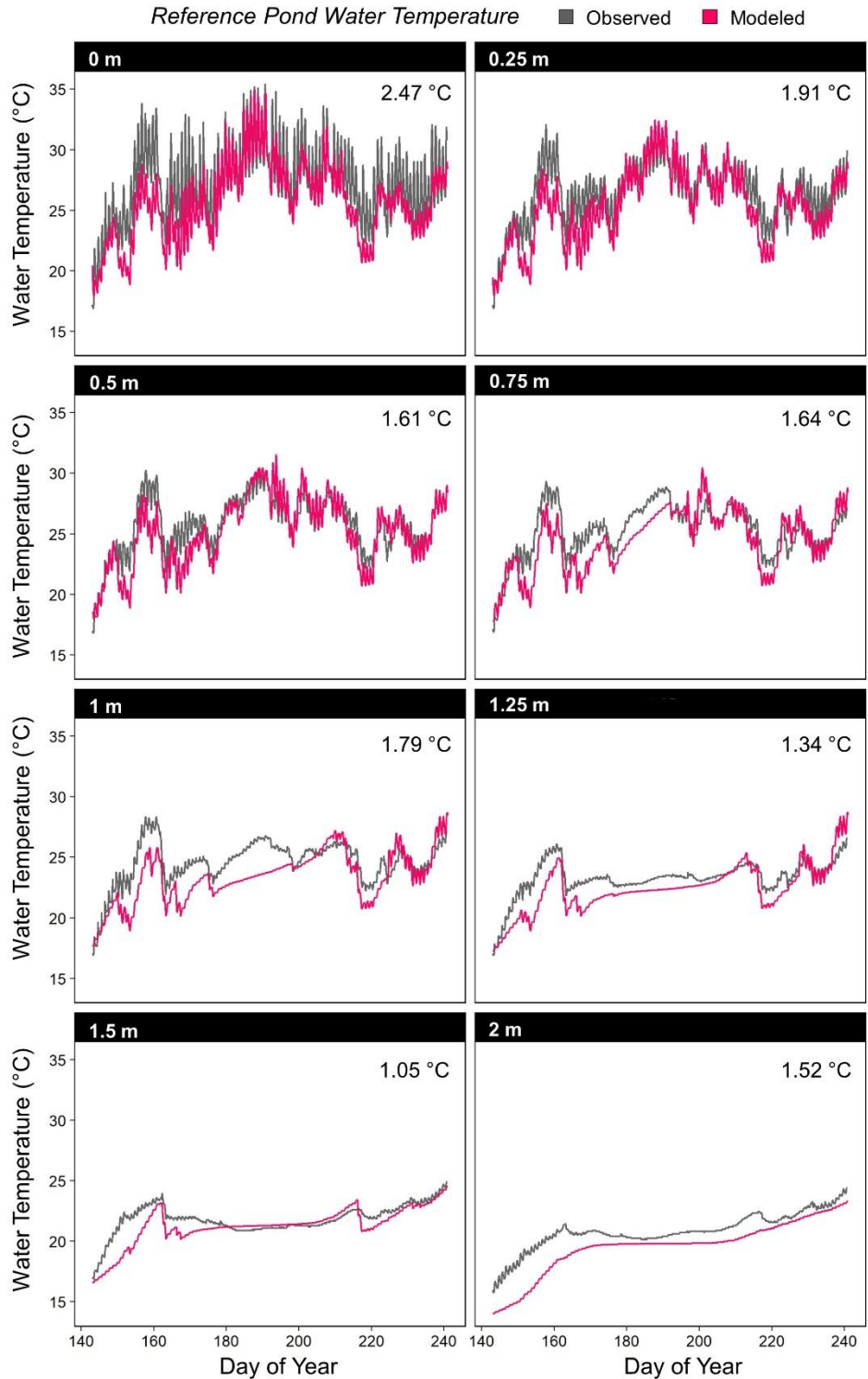
1022 **Figure S4.** Bottom water temperature time series across sampling sites in each study pond. The
1023 grey box marks an aquatic heat event from DOY 185-190. Yellow dashed lines indicate the
1024 nutrient additions on DOY 176 and 211. During the heat event, bottom water temperature
1025 remained consistent at both the deep and middle sites of the reference pond. In contrast, bottom
1026 water temperatures in the nutrient addition pond increased by 1.3 °C and 1.4 °C at the deep and
1027 middle sites, respectively. At the shallow site in both ponds bottom water temperatures increased
1028 1-1.3 °C during the heat event, suggesting an interaction with water depth.



1029

1030 **Figure S5.** Time series of (A) surface and (B) bottom water dissolved oxygen (DO) saturation
 1031 and (C) the difference between surface and bottom water DO saturation is based on weekly
 1032 profiles at 18 sites across each pond. Negative values of delta-DO indicate greater DO saturation
 1033 in the bottom waters while positive values mean DO is higher in the surface layer. Values from
 1034 each pond were GAM-fit to highlight temporal trends. The asterisks next to the effective degrees
 1035 of freedom (edf) correspond to the estimated p-value, with *** denoting $p < 0.001$. Across both
 1036 panels, dashed yellow lines mark the timing of nutrient additions and the grey box highlights the
 1037 aquatic heat event.

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Figure S6. Model fit at discrete depths in the reference pond. Predicted water temperatures from the hydrodynamic model (pink lines) are plotted against the observed temperatures at the same depths (grey lines). The root mean square error (RMSE, °C) are noted in the top right corner of each panel. Overall, the model captured the magnitude and seasonal trends of temperature throughout the water column.