

Mixotrophs generate carbon tipping points under warming

Daniel Wieczynski¹, Holly Moeller², and Jean-Philippe Gibert¹

¹Duke University

²University of California Santa Barbara

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Abstract

Mixotrophs are ubiquitous and integral to microbial food webs, but their impacts on the dynamics and functioning of broader ecosystems are largely unresolved. Here, we show that mixotrophy produces a unique, dynamic type of food web module that exhibits unusual ecological dynamics, with surprising consequences for carbon flux under warming. We find that mixotrophs generate alternative stable carbon states across temperatures—including an autotrophy-dominant carbon sink state, a heterotrophy-dominant carbon source state, and cycling between these two. Moreover, warming always shifts this mixotrophic system from a carbon sink state to a carbon source state, but increasing nutrients erases early warning signals of this transition and expands hysteresis. This suggests that mixotrophs can generate critical carbon tipping points under warming that will be more abrupt and less reversible when combined with increased nutrient levels, having widespread implications for ecosystem functioning in the face of rapid global change.

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7 **AUTHORS**

8 Daniel J. Wieczynski^{1,*} (daniel.wieczynski@duke.edu),

9 Holly V. Moeller² (holly.moeller@lifesci.ucsb.edu),

10 Jean P. Gibert¹ (jean.gibert@duke.edu)

11

12 **AFFILIATIONS**

13 ¹Department of Biology, Duke University, Durham NC, 27708, USA

14 ²Department of Ecology, Evolution, and Marine Biology, University of California, Santa

15 Barbara, Santa Barbara, CA 93106, USA

16 *Corresponding author

17

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27 DJW designed the study and performed the mathematical modeling with support from HVM and
28 JPG. DJW wrote the first draft of the manuscript and all authors contributed substantially to
29 revisions.

30

31 **DATA ACCESSABILITY STATEMENT**

32 All software and data from this study are available on GitHub (upon publication).

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41 **ABSTRACT**

42 Mixotrophs are ubiquitous and integral to microbial food webs, but their impacts on the
43 dynamics and functioning of broader ecosystems are largely unresolved. Here, we show that
44 mixotrophy produces a unique, dynamic type of food web module that exhibits unusual
45 ecological dynamics, with surprising consequences for carbon flux under warming. We find that
46 mixotrophs generate alternative stable carbon states across temperatures—including an
47 autotrophy-dominant carbon sink state, a heterotrophy-dominant carbon source state, and cycling
48 between these two. Moreover, warming always shifts this mixotrophic system from a carbon sink
49 state to a carbon source state, but increasing nutrients erases early warning signals of this
50 transition and expands hysteresis. This suggests that mixotrophs can generate critical carbon
51 tipping points under warming that will be more abrupt and less reversible when combined with
52 increased nutrient levels, having widespread implications for ecosystem functioning in the face
53 of rapid global change.

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64 **INTRODUCTION**

65 Microbial organisms play a critical role in ecosystem carbon and nutrient cycling (Kayranli *et al.*
66 2010; Schimel & Schaeffer 2012; Steinberg & Landry 2017; Zhang *et al.* 2018; Geisen *et al.*
67 2020; Rocca *et al.* 2022) that is likely to change with rapidly shifting global conditions (Zhou *et*
68 *al.* 2012; Bradford *et al.* 2019; Smith *et al.* 2019; Geisen *et al.* 2021; Wieczynski *et al.* 2021).
69 Understanding the net impacts of global change on ecosystem flux requires untangling the roles
70 of a diverse assortment of ecological strategies within the microbial world (Bengtsson *et al.*
71 1996; Petchey *et al.* 1999; Gao *et al.* 2019; Thakur & Geisen 2019; Geisen *et al.* 2020;
72 Kuppardt-Kirmse & Chatzinotas 2020).

73

74 Mixotrophy is a common strategy within microbial communities, but its impacts on the dynamics
75 of ecosystem processes remains relatively unresolved (Sanders 1991; Jones 2000; Esteban *et al.*
76 2010; Mitra *et al.* 2014; Jassej *et al.* 2015; Selosse *et al.* 2017; Stoecker *et al.* 2017; Johnson &
77 Moeller 2018; Flynn *et al.* 2019). Mixotrophic organisms use a combination of energy
78 acquisition (or trophic) modes: autotrophy (phototrophy or chemoautotrophy) and heterotrophy
79 (phagotrophy or chemoheterotrophy) (Stoecker 1998; Esteban *et al.* 2010). Although mixotrophy
80 also occurs in plants (Selosse & Roy 2009; Schmidt *et al.* 2013) and animals (Orr 1888; Venn *et*
81 *al.* 2008; Graham *et al.* 2013), the majority of mixotrophs are microorganisms like bacteria,
82 archaea, protists, and fungi (Selosse *et al.* 2017). Mixotrophic microbes are ubiquitous in
83 terrestrial, freshwater, and marine systems (Sanders 1991; Stoecker 1998; Mieczan 2009;
84 Esteban *et al.* 2010; Worden *et al.* 2015; Selosse *et al.* 2017; Stoecker *et al.* 2017; Flynn *et al.*
85 2019), and mixotrophy is increasingly recognized as a dominant nutrient acquisition strategy
86 within microbial food webs (Sanders 1991; Mitra *et al.* 2014; Jassej *et al.* 2015; Selosse *et al.*

87 2017). By acting as both primary producers and consumers, mixotrophs play a unique role in
88 ecosystem carbon and nutrient cycling (Jones 2000; Mitra *et al.* 2014; Jasey *et al.* 2015) that is
89 likely to change with warming (Wilken *et al.* 2013). Elucidating mixotrophic responses to
90 rapidly changing environmental conditions is thus essential for understanding and predicting the
91 impacts of global climate change on ecosystem functioning.

92

93 Mixotrophic strategies may be characterized by differential utilization of three basic resources—
94 light, dissolved nutrients, and prey organisms (Jones 1997; Stoecker 1998; Mitra *et al.* 2016).

95 Multiple schemes have been developed to organize mixotrophs according to their dependencies
96 on these core resources (Jones 1997; Stoecker 1998; Mitra *et al.* 2016). Accordingly, mixotrophs
97 generally fall into one of three basic categories: 1) “ideal” mixotrophs equally balance
98 autotrophy and heterotrophy, 2) “phagotrophic algae” are primarily autotrophs that use
99 heterotrophy to supplement either carbon or nutrient needs, and 3) “photosynthetic protozoa” are
100 primarily heterotrophs that supplement carbon needs with autotrophy. However, these strategies
101 are not fixed. Indeed, changes in the availability of light, nutrients, or prey can cause individual
102 organisms to shift from one mode of energy acquisition to another (Stoecker 1998).

103

104 Consequently, mixotrophy likely represents a unique type of food web module whose structural
105 and dynamical qualities vary in response to shifts between energy acquisition modes in
106 mixotrophs. Under certain conditions, mixotrophs may benefit more from autotrophy, acquiring
107 carbon primarily via photosynthesis rather than predation (Figure 1, left). Under other
108 conditions, heterotrophy may be favored and carbon acquired primarily via consumption of prey
109 (Figure 1, right). Importantly, mixotrophs may dynamically switch between these energy

110 acquisition modes as conditions change across space or time. This dynamic blending of energy
111 acquisition modes could introduce novel dynamical behaviors, altering population dynamics,
112 species interactions, and equilibria in ecological communities in ways that are not fully captured
113 by current theoretical frameworks. Although some studies have investigated mixotrophic
114 dynamics using mathematical models (e.g., Thingstad *et al.* 1996; Jost *et al.* 2004; Moeller *et al.*
115 2016, 2019; Yang *et al.* 2016; Moroz *et al.* 2019), these tend to be tailored to specific systems,
116 organisms, and environmental conditions, potentially missing the full range of dynamical
117 behaviors possible in mixotrophic systems. To begin to explore these possible behaviors—and
118 how they are altered by environmental change—we need a generalizable mixotrophic model that
119 incorporates dynamically shifting energy acquisition modes in response to dynamic changes in
120 the availability of essential resources and variation in environmental conditions. Yet, no single
121 model to date has done this, which precludes us from truly understanding the roles of mixotrophs
122 within food webs and their associated impacts on ecosystem-level functioning.

123

124 Additionally, the processes that control mixotrophic population dynamics—autotrophic
125 production (photosynthesis), heterotrophic production (predation), respiration, mortality, etc.—
126 are expected to be accelerated by warming ((Brown *et al.* 2004; Savage *et al.* 2004; Allen *et al.*
127 2005; Dell *et al.* 2014)), but may exhibit different sensitivities to temperature change.
128 Importantly, autotrophic production exhibits significantly lower sensitivity to increasing
129 temperature than heterotrophic production, as evidenced by temperature sensitivities (in the form
130 of ‘activation energies’) of $\sim 0.32\text{eV}$ and $\sim 0.65\text{eV}$, respectively (Allen *et al.* 2005; López-Urrutia
131 *et al.* 2006; Yvon-Durocher & Allen 2012). Consequently, some empirical (Wilken *et al.* 2013,
132 2018) and theoretical (Yang *et al.* 2016) evidence suggests that mixotrophs will tend to favor

133 heterotrophy over autotrophy with warming. But whether this transition will be sudden or
134 gradual, and whether this will be mediated by other environmental change factors (e.g.,
135 eutrophication), is virtually unknown.

136

137 Here we develop a generalizable mixotrophic food web model to evaluate the impacts of
138 environmental change on mixotrophic dynamics and carbon flux. We address three main
139 questions: 1) Does environmental change (in the form of temperature and nutrient concentration)
140 alter the ecological dynamics and stability of mixotrophic systems?, 2) Does this, in turn, cause
141 shifts in carbon flux states (i.e., carbon sink and carbon source states)?, and 3) Are there early
142 warning signals for tipping points between these states? Our results show that mixotrophic
143 systems undergo complex—but predictable—dynamical transitions between alternative stable
144 carbon states with warming that may be preceded by early warning signals in the form of steady-
145 state cycling behavior. However, these early warning signals disappear and are replaced by an
146 abrupt carbon state shift when warming is accompanied by increasing nutrient levels, which has
147 important implications for ecosystem functioning in a rapidly warming and increasingly
148 anthropogenized world.

149

150 **MATERIAL AND METHODS**

151 *Mixotrophic model*

152 Different types of mixotrophs can be modeled by defining specific dependencies (functional
153 responses) of photosynthesis and consumption on three limiting resources: prey, nutrients, and
154 light (Stoecker 1998). To study the effects of warming, we also incorporate temperature
155 dependence on several rate parameters in our model. We focus our analysis on a model of

156 mixotrophy representing organisms that are primarily phagotrophic but switch to photosynthesis
 157 to obtain carbon when prey are limiting (known as “photosynthetic protozoa”, or Type-III
 158 mixotrophs in the terminology of (Stoecker 1998)). Although we study this particular type of
 159 mixotroph here, our model can be generalized to any other type of mixotrophs by replacing the
 160 functional responses for prey, nutrient, and light dependencies with alternative functional forms.

161

162 Our mixotrophic model consists of two ordinary differential equations (ODEs) that define the
 163 dynamics of a two species system—a mixotroph (M) and its prey (P):

$$\frac{dM}{dt} = M * \left(\varphi(T, N_M, P, M) + \varepsilon \lambda(T, P) - \delta_M(T) - m_M(T) \right) \quad (1a)$$

$$\frac{dP}{dt} = P * \left(\mu_P(T) \frac{N_P}{h_P + N_P} \left(1 - \frac{P}{K_P} \right) - \lambda(T, P) - \delta_P(T) - m_P(T) \right), \quad (1b)$$

164 where M and P are biomass densities in units of nanograms of carbon per liter (Table 1). The
 165 terms φ and $\varepsilon \lambda$ (consumption rate λ multiplied by a conversion efficiency ε) represent the
 166 mixotroph’s per-capita biomass production rates from photosynthesis and consumption,
 167 respectively:

$$\text{Photosynthesis: } \varphi(T, N_M, P, M) = \mu_M(T) \frac{N_M}{h_M + N_M} e^{-dP^2} \left(1 - \frac{M}{K_M} \right) \quad (2a)$$

$$\text{Consumption: } \varepsilon \lambda(T, P) = \frac{\varepsilon \alpha(T) P}{1 + \alpha(T) P} \quad (2b)$$

168 Photosynthetic production rate (φ) follows a modified logistic-growth form that incorporates
 169 dependencies on temperature (T), nutrient concentration (N_M), prey density (P), and mixotroph
 170 density (M). Per-capita photosynthetic production is assumed to decline as mixotroph density
 171 approaches a carrying capacity (K_M), due to limitation of essential resources (e.g., light). Nutrient
 172 uptake follows Michaelis–Menten kinetics where uptake rate saturates with increasing nutrient
 173 concentrations to a maximum rate ($\mu_M(T)$) according to a half-saturation constant (h_M , i.e., the

174 nutrient concentration at which the foraging rate is half the maximum possible rate). To capture a
175 reduction in photosynthetic investment when prey are abundant, the dependence of
176 photosynthetic production rate on prey density is defined by a logistic decay function (e^{-dP^2})
177 that declines with increasing prey density at a rate determined by d and saturates at a maximum
178 value as prey density approaches zero (Figure S1a). Consumption rate (λ) follows a type-II
179 functional response that saturates with increasing prey density and has an attack rate of $\alpha(T)$.
180 Biomass loss is accounted for through the parameters δ_M and m_M , which represent respiration and
181 mortality, respectively. The percentage of total production that comes from photosynthesis was
182 calculated as $\varphi/(\varphi + \varepsilon \lambda) \cdot 100$.

183
184 Prey are assumed to be exclusively chemoheterotrophic and also follow a modified logistic form,
185 with dependencies on temperature (T), nutrient concentration (N_P), and prey density (P) defined
186 by Michaelis–Menten kinetics with maximum uptake rate $\mu_P(T)$, a half-saturation constant h_P ,
187 and carrying capacity K_P . Prey biomass declines through consumption by the mixotroph (λ),
188 respiration (δ_P), and mortality (m_P).

189

190 *Temperature dependence*

191 Maximum uptake, attack, mortality, and respiration rates are all assumed to be temperature-
192 dependent (explicitly written as a function of T in Eqns. 1&2) and follow the common Arrhenius
193 form:

$$rate(T) = b_0 e^{-\frac{E_a}{k} \left(\frac{1}{T} - \frac{1}{T_{ref}} \right)} \quad (2)$$

194 where b_0 is a normalization constant, E_a is activation energy, k is Boltzmann's constant ($8.6 \cdot 10^{-5}$
195 $\text{eV} \cdot \text{K}^{-1}$), and T_{ref} is a reference temperature at which the given rate is equal to b_0 ($T_{ref} = 20^\circ\text{C}$ for
196 all parameters in our model). The temperature sensitivities of each rate are controlled by the
197 activation energies (E_a), which were empirically estimated elsewhere: $E_a = 0.32$ for
198 photosynthetic production (Allen *et al.* 2005) and $E_a = 0.65$ for heterotrophic production and
199 respiration (Brown *et al.* 2004; Dell *et al.* 2011).

200

201 *Carbon dynamics*

202 To track carbon dynamics, we calculated net CO_2 flux as total system respiration rate minus total
203 system photosynthetic rate:

$$CO_2 \text{ flux} = 3.67 * (\delta_M(T)M + \delta_P(T)P - \varphi(T, n, P, M)M) \quad (4)$$

204 where $\delta_M(T)M + \delta_P(T)P$ is total system respiration rate and the third term $\varphi(T, n, P, M)M$
205 represents the rate of carbon uptake for use in photosynthesis. The coefficient 3.67 converts
206 grams of carbon (C) to grams of carbon dioxide (CO_2) ($\text{gCO}_2 / \text{gC} = 44/12 = 3.67$).

207

208 *Equilibria and stability analysis*

209 We quantified equilibria by numerically solving the system (using Mathematica V13.0.0
210 (Wolfram Research, Inc. 2021)) across a range of temperatures ($19\text{--}23^\circ\text{C}$) and nutrient
211 concentrations ($0.45\text{--}0.95 \text{ ng L}^{-1}$). The stability and dynamical behavior of equilibria were
212 determined through local stability analysis, i.e., by calculating the eigenvalues (for our system
213 there are two, one for each state variable) of the Jacobian matrix evaluated at equilibrium in each
214 environmental state (i.e., combination of temperature and nutrient concentration), then using
215 those eigenvalues to characterize the stability of equilibria. For eigenvalues with only real parts,

216 if the dominant (largest) eigenvalue is negative, then the equilibrium is a stable node (non-
217 oscillatory), otherwise, it is an unstable node (non-oscillatory). For complex eigenvalues (with
218 imaginary parts), there were three possible equilibrium behaviors: i) a negative dominant
219 eigenvalue produces a stable focus (damped oscillations), ii) if only one eigenvalue has a
220 positive real part and it is not equal to the conjugate of the other eigenvalue, then this produces
221 an unstable focus (outward spiral), and iii) the existence of all positive real parts when one
222 eigenvalue is equal to the conjugate of the other produces a limit cycle (sustained oscillations).
223 Steady-state dynamics were identified by numerically solving the system for 100,000 time steps
224 and recording the maximum and minimum densities of each species for the last 10,000 time
225 steps. We repeated this process for all equilibria in each environmental state, initializing each
226 simulation with small perturbations from each equilibrium point (equilibrium values + 0.001).
227 This allowed us to calculate long-term stationary dynamics created by limit cycles and
228 degenerate limit cycles.

229

230 **RESULTS**

231 *Effects of temperature on mixotrophic dynamics*

232 Increasing temperature reshapes the dynamical landscape of this mixotrophic system (Figure 2).

233 At low temperatures, a single, stable equilibrium exists where mixotrophs are at an intermediate

234 density and their prey are at very low (or zero) density (Figure 2a, green). At intermediate

235 temperatures, three stable equilibria appear: i) one stable point where both species are at

236 relatively low densities (green), ii) one high-density stable point (red), and iii) a stationary cycle

237 that orbits these two stable points (blue; Figure 2b). At higher temperatures, only one equilibrium

238 exists where both species coexist at relatively high densities (Figure 2c).

239

240 These transitions between stable states are produced by a progression of bifurcations across
241 temperatures (Figure 2d). Multiple equilibria exist across a range of intermediate temperatures
242 (20.06–21.99°C) whose stability and dynamical behavior change as temperature increases. First,
243 a limit cycle appears at 20.06°C (black dashed line), but its stability is disrupted by an unstable
244 node (gray dotted line) separating it from the original stable point (green line), and the long-term
245 dynamics approach the stable point regardless of initial conditions (Figure 2d). Next, at 20.7°C
246 multiple stable states cooccur—one is the original stable point (green) and the other is a
247 stationary cycle (blue lines, gray shading) that orbits the stable point and the unstable limit cycle.
248 The high-density stable point (red) appears at 20.79°C, producing a unique form of tri-stability
249 including all three of the alternative stable states described above (Figure 2b). The stationary
250 cycle disappears at 21.05°C, leaving two alternative, static, stable points, but the low-density
251 stable point (green) quickly becomes a limit cycle that does not sustain cycling at 21.1°C.
252 Instead, the long-term trajectory here always approaches the high-density stable point (red).
253 Eventually, as temperature increases to 22°C, only one, high-density stable point (red) remains
254 (Figure 2c).

255

256 *Effects of temperature on carbon flux*

257 Increasing temperature shifts this mixotrophic system from a net carbon sink (dominated by
258 photosynthesis; Figure 2e), to alternative carbon states (sink and source; Figure 2f), to a net
259 carbon source (dominated by predation; Figure 2g). This sequence of carbon state transitions
260 corresponds with changes in the long-term carbon dynamics of the system due to shifts in the
261 dominant carbon acquisition strategy of the mixotroph. At low temperatures, most of the

262 mixotroph's biomass production comes from photosynthesis and, after accounting for carbon
263 uptake for use in photosynthesis and carbon release through respiration by both species, the net
264 flux of carbon dioxide (CO₂) in the system is negative (i.e., a net carbon sink) (Figure 2e). At
265 intermediate temperatures, three stable carbon states coexist: i) one carbon-sink state (green), ii)
266 one stationary cycle where production fluctuates between photosynthesis and predation and the
267 system cycles between a carbon-sink state and carbon-source state, respectively (blue), and iii)
268 one carbon-source state where production is dominated by predation and carbon flux is positive
269 (Figure 2f). At high temperatures, predation takes over as the sole form of production and the
270 system becomes a net carbon source (Figure 2g). Because stationary cycles span a range of
271 temperatures separating carbon sink and source states (Figure 2d), this cycling behavior can be
272 considered an early warning signal of this transition.

273

274 *Combined effects of temperature and nutrient concentration*

275 The temperature-driven progression through alternative stable states is mediated by nutrient
276 concentration (Figure 3). Changes in temperature and nutrient levels leads to a complex
277 equilibrium landscape that produces a rich assortment of behaviors (Figure 3a). Within this
278 landscape, the range of temperatures producing multiple nontrivial equilibria widens with
279 increasing nutrient concentration (Figure 3a; region inside solid black line), creating upper and
280 lower equilibrium planes in three-dimensional space (Figure 3b) consisting of various
281 combinations of stable points and limit cycles that are separated by an interior plane of unstable
282 points (Figure 3a).

283

284 The carbon-flux behavior of a mixotrophic system in any given environmental state (i.e.,
285 combination of temperature and nutrients) depends on the arrangement of these equilibria
286 (Figure 3a). A static carbon sink state can occur within a region of low temperatures and high
287 nutrient concentrations, where either a single, stable point equilibrium exists (at low mixotroph
288 density) or a stable point in the lower plane is accompanied by a limit cycle in the upper plane
289 that cannot sustain cycling (Figure 3a, green). Conversely, a static carbon source state occurs
290 when temperatures are higher and nutrient concentrations are lower, associated with either a
291 single, stable, high-mixotroph-density equilibrium point or a stable point in the upper plane that
292 is accompanied by a limit cycle in the lower plane (Figure 3a, red). Interestingly, stationary
293 cycling can occur under any combination of equilibrium points, producing fluctuations in carbon
294 flux between carbon sink and source states (Figure 3a, gray). In some cases, stationary cycling
295 can occur around fixed, stable points, even without limit cycle present (see Discussion section
296 for more information). At high temperatures and nutrient concentrations, hysteresis can occur at
297 temperatures for which both static, stable carbon sink and source states occur (Figure 3a, purple).

298

299 *Early warning signals for transitions between carbon flux states*

300 Interestingly, increasing nutrient loads erases early warning signals of a shift between carbon
301 sink to carbon source states with warming (Figure 3c–e). Early warning signals come in the form
302 of stationary fluctuations between carbon sink and source states that precede the transition to a
303 static carbon source state as temperature increases (gray region in Figure 3a). Indeed, at low
304 nutrient concentrations ($N_M = 0.6 \text{ ng L}^{-1}$), increasing temperatures produces a large temperature
305 window over which stationary cycling and fluctuations in carbon flux dynamics occur before the
306 system eventually locks in to a static carbon source state (Figure 3e). As nutrient concentration

307 increases, the range of temperatures that produce fluctuations shrinks (gray region in Figure 3a)
308 and alternative stable point equilibria begin to overlap at intermediate temperatures (Figure 3d).
309 When nutrient concentrations become high enough, stationary cycles completely disappear and
310 alternative, static point equilibria overlap across a wide range of temperatures (Figure 3a&c). In
311 this case, the warming-induced tipping point to a static carbon source state is abrupt and occurs
312 without warning. Additionally, once warming has shifted the system to a carbon source state, a
313 significant reduction in temperature ($>1^{\circ}\text{C}$) would be required to revert the system back to the
314 carbon sink state (hysteresis; Figure 3c).

315

316 Generally speaking, although warming always leads to a transition from a carbon sink state to a
317 carbon source state, whether this transition is preceded by a period of fluctuating carbon flux
318 dynamics (early warning signal) depends on nutrient concentrations. Moreover, increasing
319 nutrients reduces the temperature range over which fluctuating carbon flux dynamics occur
320 (shortening early warning signals) while also increasing the temperature range over which static
321 carbon sink and source states overlap (widening hysteresis) (Figures 3a & 4).

322

323 **DISCUSSION**

324 Mixotrophic organisms and their prey can be considered a unique type of food web module that
325 dynamically transitions between autotrophy (single-species or competitive dynamics) and
326 heterotrophy (consumer-resource dynamics), generating surprising dynamical behaviors that can
327 have important—albeit largely unknown—impacts on ecosystem functioning in novel
328 environments. Here we show how warming can shift mixotrophic systems from a
329 photosynthesis-dominant net carbon sink (Figure 2a&e), through alternative stable carbon states

330 (Figure 2b&f), and ultimately to a predation-dominant net carbon source (Figure 2c&g). These
331 transitions are preceded by early warning signals in the form of fluctuations between carbon
332 source and sink states when nutrient concentrations are low (Figure 3a&e). But increasing
333 nutrient levels erases these early warning signals by replacing cyclic behavior with alternative,
334 static carbon sink and source states (hysteresis; Figure 3a&c). Taken together, this suggests that
335 mixotrophic systems will tend to shift from carbon sinks to carbon sources with warming and
336 this transition will be more abrupt and less reversible when combined with increased nutrient
337 levels. Given the ubiquity of mixotrophs across all types of ecosystems (Sanders 1991; Stoecker
338 1998; Mieczan 2009; Worden *et al.* 2015; Selosse *et al.* 2017; Stoecker *et al.* 2017; Flynn *et al.*
339 2019), our results uncover a potentially crucial but previously unknown aspect of ecosystem
340 responses to global change.

341
342 Ecologists have been concerned about identifying how changing environmental conditions might
343 produce tipping points and abrupt regime shifts for decades (Holling 1973; May 1977; Scheffer
344 *et al.* 2001; Folke *et al.* 2004; Dakos & Hastings 2013; Dakos *et al.* 2019). Our study exposes a
345 new mechanism by which abrupt regime shifts may occur—through the unique dynamics of
346 mixotrophic organisms. We find that early warning signals of such shifts may occur in the form
347 of fluctuating dynamics that bridge a transition between static carbon sink and carbon source
348 states. However, we also find that these early warning signals may be environmentally context-
349 dependent—the nature of regime shifts across one environmental gradient might depend on the
350 state of separate environmental factors (as is also evident in some empirical examples of regime
351 shifts (Folke *et al.* 2004)). In our system, the window of early warning signals with warming
352 (e.g., fluctuations spanning temperature changes of $\sim 0.25^{\circ}\text{C}$ vs $\sim 1.5^{\circ}\text{C}$ in Figures 2d and 2e,

353 respectively), and indeed their very existence (e.g., the lack of fluctuations in Figure 2c),
354 depends on coordinated changes along multivariate environmental gradients (temperature and
355 nutrient concentrations in our case). This finding that specific, multivariate environmental
356 contexts control the nature of regime shifts could also shed light on why tipping points are so
357 elusive in nature (Connell & Sousa 1983; Dudley & Suding 2020; Hillebrand *et al.* 2020). We
358 propose that mixotrophs are not only integral to ecosystem responses to climate change (Jassey
359 *et al.* 2015), but also provide an early warning for carbon tipping points and an opportunity to
360 study complex regime shifts and variation in early warning signals across multivariate
361 environmental gradients.

362

363 There is growing recognition that temperature and nutrients interact to impact the structure and
364 dynamics of ecological communities (Binzer *et al.* 2012, 2016; Gilbert *et al.* 2014; Sentis *et al.*
365 2014; Han *et al.* 2022). Discovering conditions under which temperature-nutrient interactions
366 occur and which properties of ecological systems are affected (e.g., species extinction risk, food
367 web structure and stability, etc.) is of particular interest. Our results show that increasing
368 temperature leads to important dynamical shifts across alternative stable states in mixotrophic
369 systems, but whether this change involves stationary cycling (fluctuating alternative stable states)
370 or hysteresis (static alternative stable states) is controlled by nutrients (Figures 3 & 4). As a
371 result, nutrient levels mediate the impacts of warming on carbon flux dynamics and also
372 determine our ability to predict abrupt transitions between alternative carbon flux states. The
373 critical condition producing this previously unrecognized temperature-nutrient interaction in our
374 model is the dynamic balancing of carbon uptake (via photosynthesis) and carbon release (via
375 respiration) due to flexible energy acquisition strategies in mixotrophs. However, it is possible

376 that the temperature-nutrient interaction studied here might extend beyond mixotrophic systems
377 to other multispecies systems that also dynamically balances carbon uptake and carbon release
378 (i.e., systems that include both autotrophs and heterotrophs). Determining the generality of this
379 type of temperature-nutrient interaction is an interesting question and area for future research.

380

381 The mixotrophic system studied here produces some highly unusual behaviors that have rarely—
382 if ever—been described in ecological systems. Specifically, our model produces a strange and
383 unique form of tri-stability—two alternative stable foci and stable cycling around these points
384 (Figures 2b & 3d)—with important associated impacts on carbon flux dynamics. Another
385 example of unusual behavior occurs when nutrient concentration is low (Figure 3e): some
386 temperatures (19.24–19.65°C) produce stationary cycling around a single fixed-point equilibrium
387 (i.e., a single stable focus that is encircled by two limit cycles—one outer, stable cycle and one
388 inner, unstable cycle). In this situation, the system can produce two possible long-term
389 behaviors: i) dampened oscillations toward the stable focus point when initial conditions are
390 inside the inner, unstable limit cycle or ii) stationary cycling around this stable point when initial
391 conditions are outside the unstable limit cycle. This specific arrangement of coexisting attractors
392 has been observed before in non-ecological systems (De Carvalho Braga & Mello 2013), but to
393 our knowledge, it has yet to be described in an ecological system. The dynamics in each of these
394 examples are a direct result of the flexible carbon acquisition strategies of mixotrophs and
395 variation in environmental conditions, suggesting that other unusual dynamics are possible, or
396 even common, in mixotrophic systems and probably vary across environments. Hence,
397 investigating the dynamical behaviors of mixotrophic systems could fundamentally change our

398 understanding about the dynamics and structure of microbial communities as well as ecosystem
399 responses to global change.

400

401 Our study focuses on a specific type of mixotrophic organism—a primarily predatory organism
402 that uses photosynthesis to supplement energy needs when prey densities are low. But several
403 different types of mixotrophic organisms exist, exhibiting a wide range of mixotrophic strategies
404 and responses to changes in light, nutrient concentrations, and prey densities (Jones 1997;
405 Stoecker 1998; Mitra *et al.* 2016). Each type of mixotroph is likely to produce unique dynamical
406 responses to changes in environmental conditions with different associated impacts on carbon
407 flux. As such, mixotrophs may cause a rich array of novel dynamics that have yet to be
408 uncovered either theoretically or empirically. Although our analysis is based on one specific type
409 of mixotroph, we designed our modeling framework so that it can easily be extended to
410 incorporate the specific resource dependencies of any type of mixotroph simply by defining
411 functional responses for light availability, nutrient concentrations, and prey densities as desired
412 (see Supporting Information for details). In addition, our analysis makes several other
413 assumptions regarding the particular sort of mixotrophic system studied here: two-species
414 system, heterotrophic prey, static nutrient concentrations, single limiting nutrient, fixed
415 stoichiometry, static environments, etc. For example, our analysis considered static nutrient
416 concentrations, but we find that our results are robust to the inclusion of nutrient dynamics
417 (Figure S2). In addition, we focused only on the effects of variation in nutrients utilized by the
418 mixotroph species, however, increasing prey nutrients may mitigate, or even reverse, the
419 transitions between carbon flux states with warming (Figure S3). Furthermore, it remains unclear
420 how explicit competition for resources between a mixotroph and its prey might impact carbon

421 flux. Relaxing these assumptions could have myriad consequences for dynamics that should be
422 explored in future studies.

423

424 Overall, we show that these globally distributed (Sanders 1991; Stoecker 1998; Mieczan 2009;
425 Esteban *et al.* 2010; Worden *et al.* 2015; Selosse *et al.* 2017; Stoecker *et al.* 2017; Flynn *et al.*
426 2019) and massively abundant (Bar-On *et al.* 2018) mixotrophic microbes exhibit a rich array of
427 dynamical responses to joint changes in temperature and nutrient levels, leading to
428 fundamentally important tipping points between carbon flux states. We also show that nutrient
429 levels determine whether these carbon tipping points are abrupt or accompanied by early
430 warning signals, which is of paramount importance in a rapidly warming and increasingly
431 human-influenced world.

432

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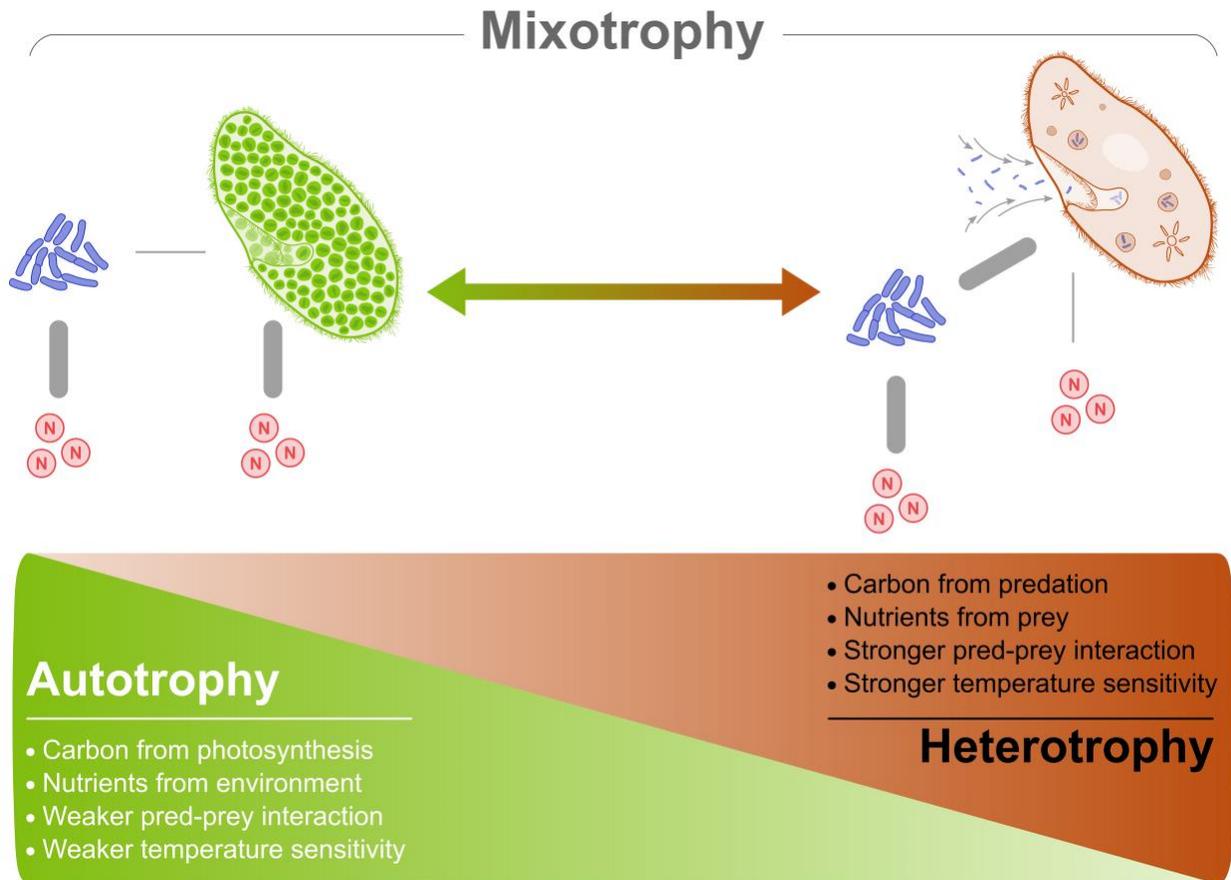
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609 **Figure 1.** Mixotrophs move dynamically along a spectrum of energy acquisition modes between
 610 autotrophy and heterotrophy according to changes in the environment and three essential
 611 resources: nutrients, prey, and light. A mixotrophic protist is shown here with its prey (bacteria;
 612 blue) and their respective essential nutrients (N). When autotrophy dominates, carbon is obtained
 613 primarily via photosynthesis, nutrients come from the environment, and the mixotroph occupies
 614 the same trophic level as its prey. When heterotrophy dominates, carbon and nutrients are
 615 obtained primarily via predation and the mixotroph occupies a higher trophic level than its prey.
 616 As mixotrophs switch between autotrophy and heterotrophy, the mixotrophic food web module
 617 shifts between single-species dynamics (or competition, if the mixotroph shares a resource with
 618 its prey) and predator-prey dynamics, respectively. The dynamic nature of the mixotrophic food

619 web module likely impacts the structure and dynamics of food webs as well as the flux of matter
620 and energy in broader ecosystems.

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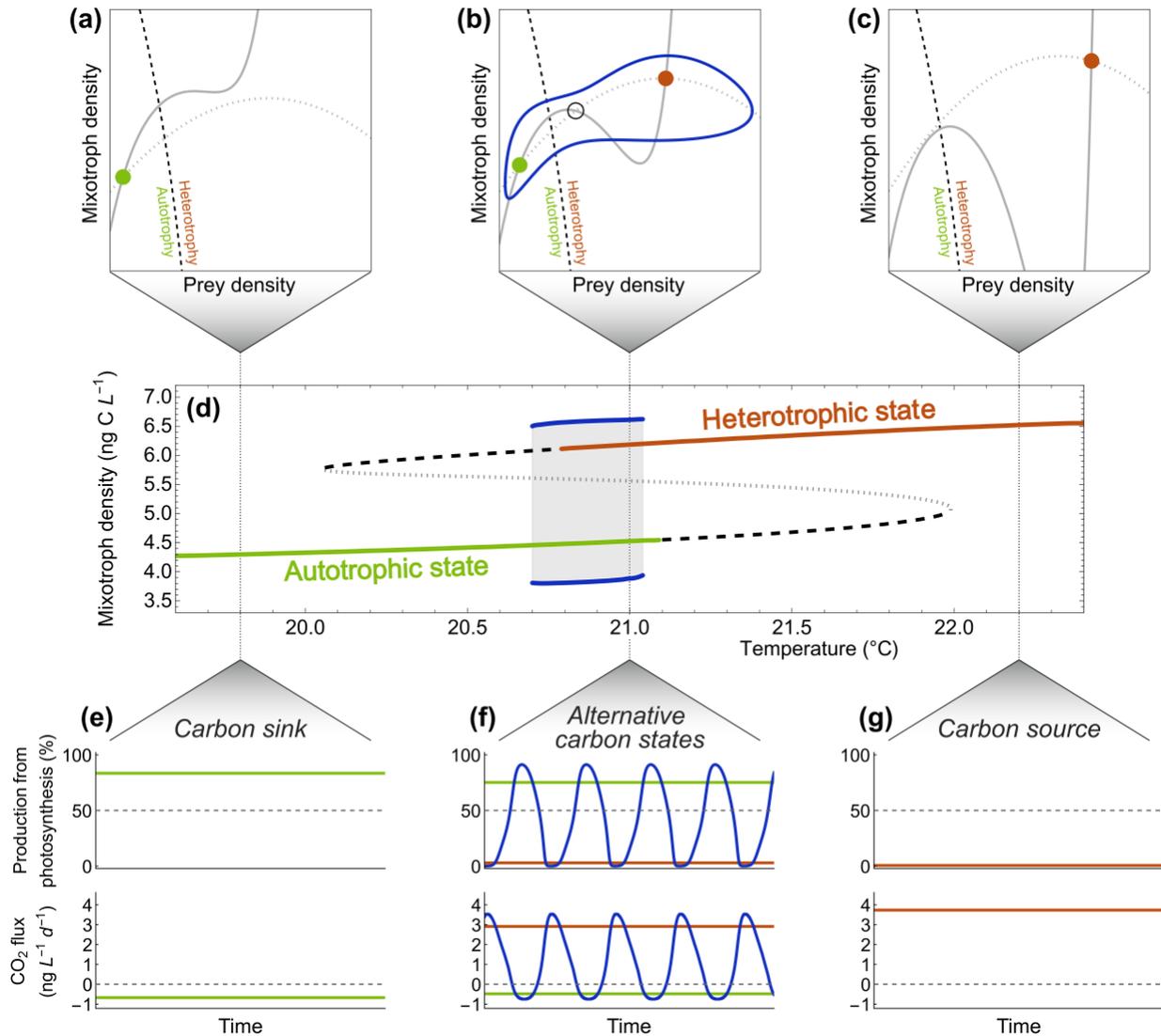
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643 **Figure 2.** Increasing temperature shifts equilibrium densities, the balance between

644 photosynthesis and consumption, and net CO₂ flux. (a–c) Phase portraits displaying null clines

645 (gray lines) for the prey species (dotted) and the mixotrophic species (solid). Intersections of

646 these null clines represent equilibrium points that are either stable (solid green and red dots) or

647 unstable (open circle). The blue lines indicate stable limit cycles that orbit the three interior

648 equilibria. The black dashed line separates a region where photosynthesis dominates production

649 (left) from a region where predation dominates production (right). (d) A bifurcation diagram

650 displaying transitions between equilibrium scenarios as a function of increasing temperature. (a),

651 (b), and (c) correspond to temperatures of 19.8°C, 21.0°C, and 22.2°C, respectively. (e–g) Long-
652 term dynamical behavior of the percentage of production from photosynthesis in the mixotroph
653 and the total system net CO₂ flux at 19.8°C, 21.0°C, and 22.2°C, respectively. Colors correspond
654 to stable equilibria and limit cycles in (a–d).

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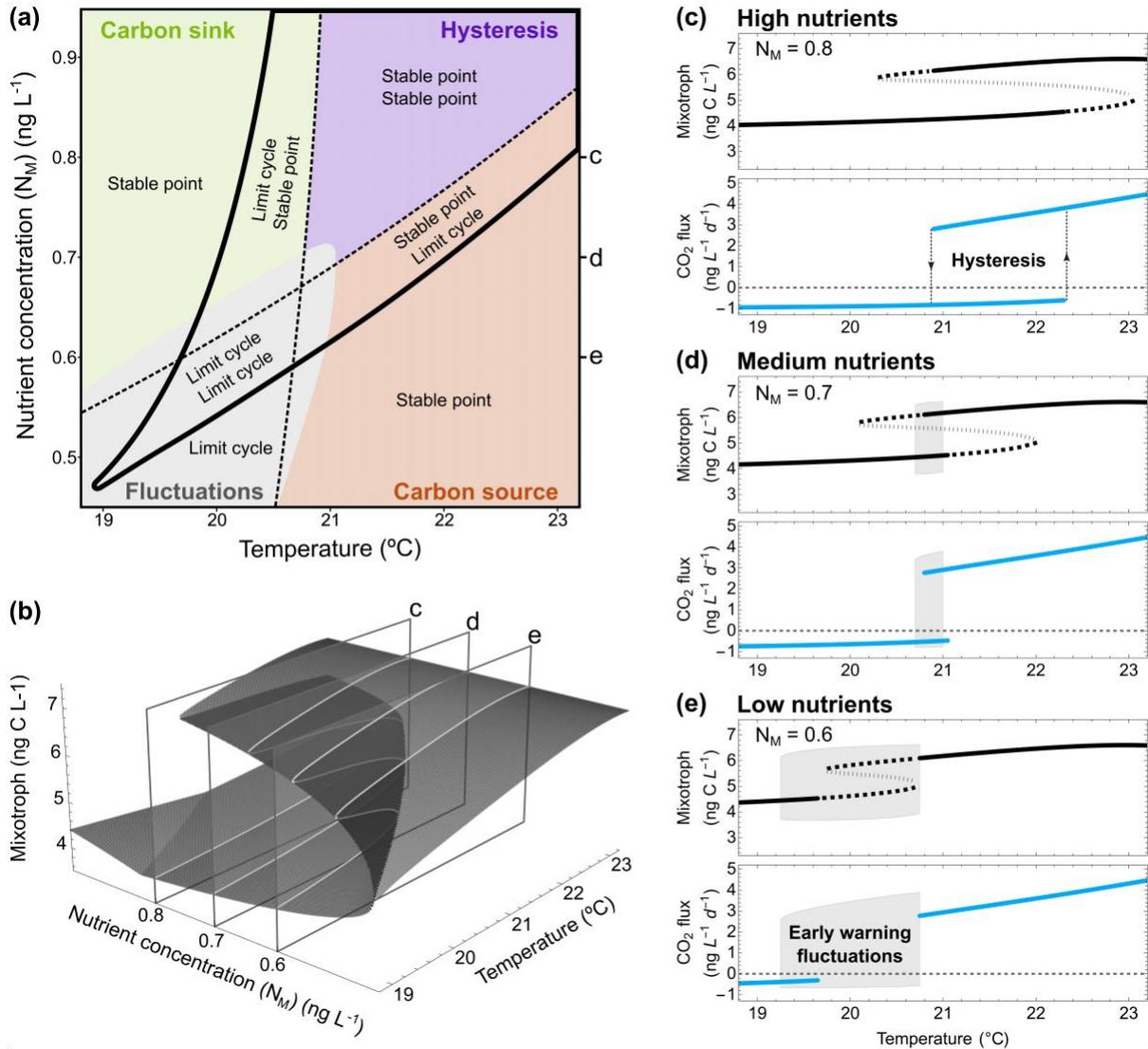
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675 **Figure 3.** Gradients in temperature and nutrient concentrations produce a rich landscape of
 676 equilibrium behaviors. (a) Different environmental conditions produce different equilibrium
 677 scenarios with one or two stable points or limit cycles (the solid black line delineates regions
 678 with one (outside) or two (inside) equilibria and black dashed lines further subdivide these
 679 regions). For regions with two equilibria, the upper and lower text correspond to the orientation
 680 of upper and lower equilibria in three-dimensional space (b). The steady-state carbon-flux
 681 behaviors of each equilibrium scenario are shown in colored regions: static carbon sink (green),
 682 static carbon source (red), fluctuations between carbon sink and source states (gray), and

683 hysteresis with overlapping carbon sink and source states (purple). (b) In three-dimensional
684 space, equilibria create a folded landscape where the upper and lower planes are either stable
685 points or limit cycles and are separated by an interior plane of unstable equilibria. (c–e) show
686 bifurcation diagrams of equilibrium densities (upper panels) and steady-state CO₂ flux (lower
687 panels) across temperatures for three different nutrient concentrations (indicated by “c”, “d”, and
688 “e” in panels (a) and (b)). Solid lines (black and blue) denote fixed point equilibria, dashed lines
689 denote unstable limit cycle equilibria, gray regions denote stationary cycling (fluctuations), and
690 dotted lines denote unstable equilibria (i.e., the interior plane in (b)).

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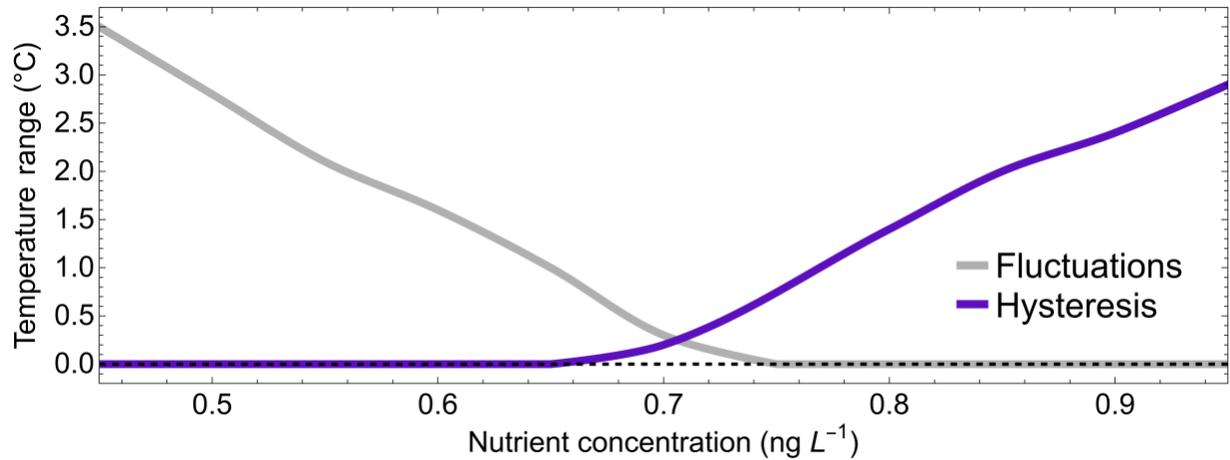
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707 **Figure 4.** The effect of nutrient concentration on the range of temperatures over which
 708 fluctuations in carbon flux (an early warning signal of a carbon tipping point) and overlapping
 709 static carbon sink and source states (hysteresis) occur. The decline in fluctuations with increasing
 710 nutrients (gray) indicates a reduction in the temperature window producing early warning
 711 signals. Increases in the range of temperatures where stable carbon states overlap (purple)
 712 indicates increasing hysteresis.

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723 **Table 1.** Variables and parameters used in the mixotrophy model.

Variable/Parameter	Definition	Units	Value
M, P	Biomass density	ng C L ⁻¹	na
N_i	Nutrient concentration	ng L ⁻¹	$N_M = [0.4, 1.0]$ $N_P = 0.7$
h_i	Half-saturation constant	ng	$h_M = 0.8$ $h_P = 0.3$
d	Photosynthesis prey dependence decline rate	n/a	0.072
K_i	Carrying capacity	ng C L ⁻¹	$K_M = 10$ $K_P = 19$
ε	Max conversion efficiency	n/a	0.25
Temperature-dependent parameters (following Eqn. 2)			
$\mu_i(T)$	Max production rate	t ⁻¹	$\mu_M(T): b_0 = 0.45; E_a = 0.32$ $\mu_P(T): b_0 = 1.35; E_a = 0.65$
$\alpha(T)$	Attack rate	t ⁻¹	$b_0 = 0.21; E_a = 0.65$
$\delta_i(T)$	Respiration rate	t ⁻¹	$\delta_M(T): b_0 = 0.07; E_a = 0.65$ $\delta_P(T): b_0 = 0.05; E_a = 0.65$
$m_i(T)$	Mortality rate	t ⁻¹	$m_M(T): b_0 = 0.072; E_a = 0.45$ $m_P(T): b_0 = 0.052; E_a = 0.45$

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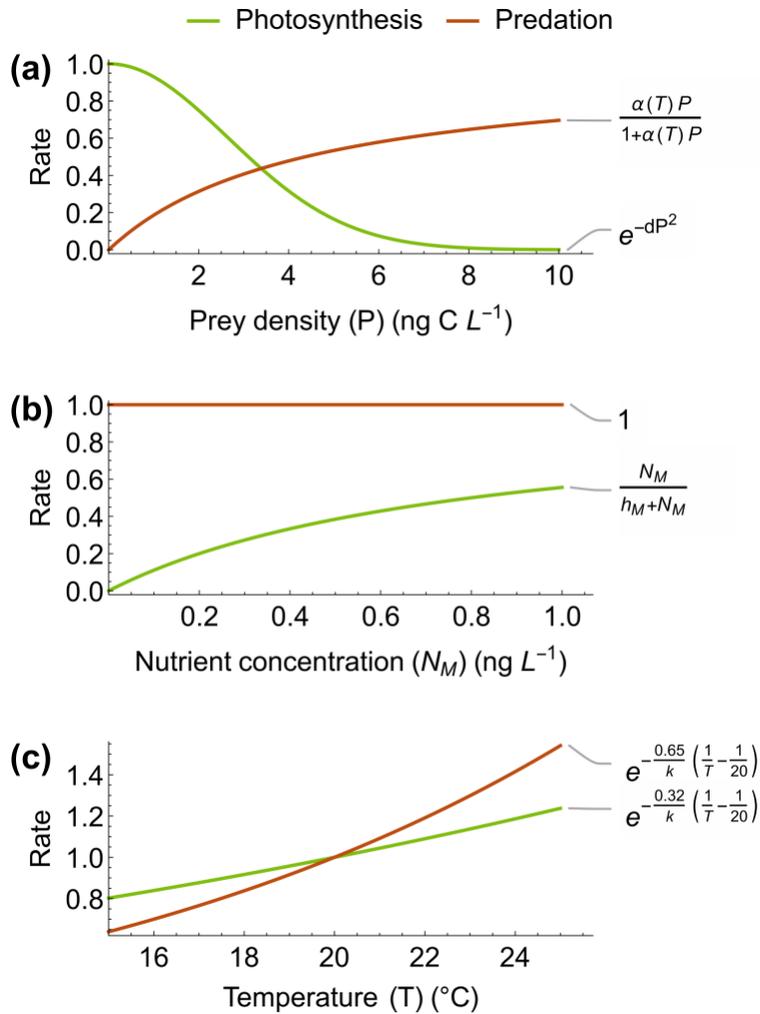
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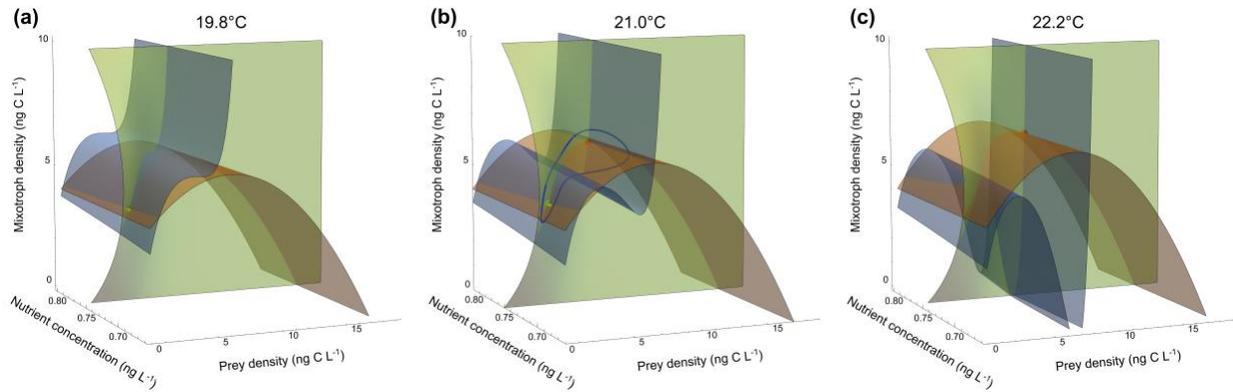
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739 **Figure S1.** Functional forms of functional responses for mixotroph photosynthesis and predation
 740 rates across (a) prey densities, (b) nutrient concentrations, and (c) temperatures.



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747 **Figure S2.** Equilibrium phase space shown for a variation of our mixotrophy model that includes

748 dynamic nutrients for the mixotroph. Orange, blue, and green planes correspond to prey,

749 mixotroph, and nutrient null clines. Intersections of these null clines represent equilibrium points

750 (solid green and red dots) and the blue line indicates stable limit cycles that orbit the interior

751 equilibria (as in Figure 2 in the main text). Although the nutrient dimension introduces more

752 complex equilibria that included changes in nutrient concentrations, the results here are

753 qualitatively the same as in the static nutrient model. In this version of the model, nutrients

754 utilized by the mixotroph follow chemostat dynamics, with reduction do to mixotroph

755 photosynthetic production: $\frac{dN_M}{dt} = \tau(N_{M,feed} - N_M) - M * \varphi(T, N_M, P, M)$, where N_M is the

756 concentration of nutrients utilized by the mixotroph, τ is a dilution rate, $N_{M,feed}$ is a feed

757 concentration for mixotroph nutrients, M is mixotroph density, and φ is the per-capita

758 photosynthetic production rate of the mixotroph. Nutrient model parameters used for the results

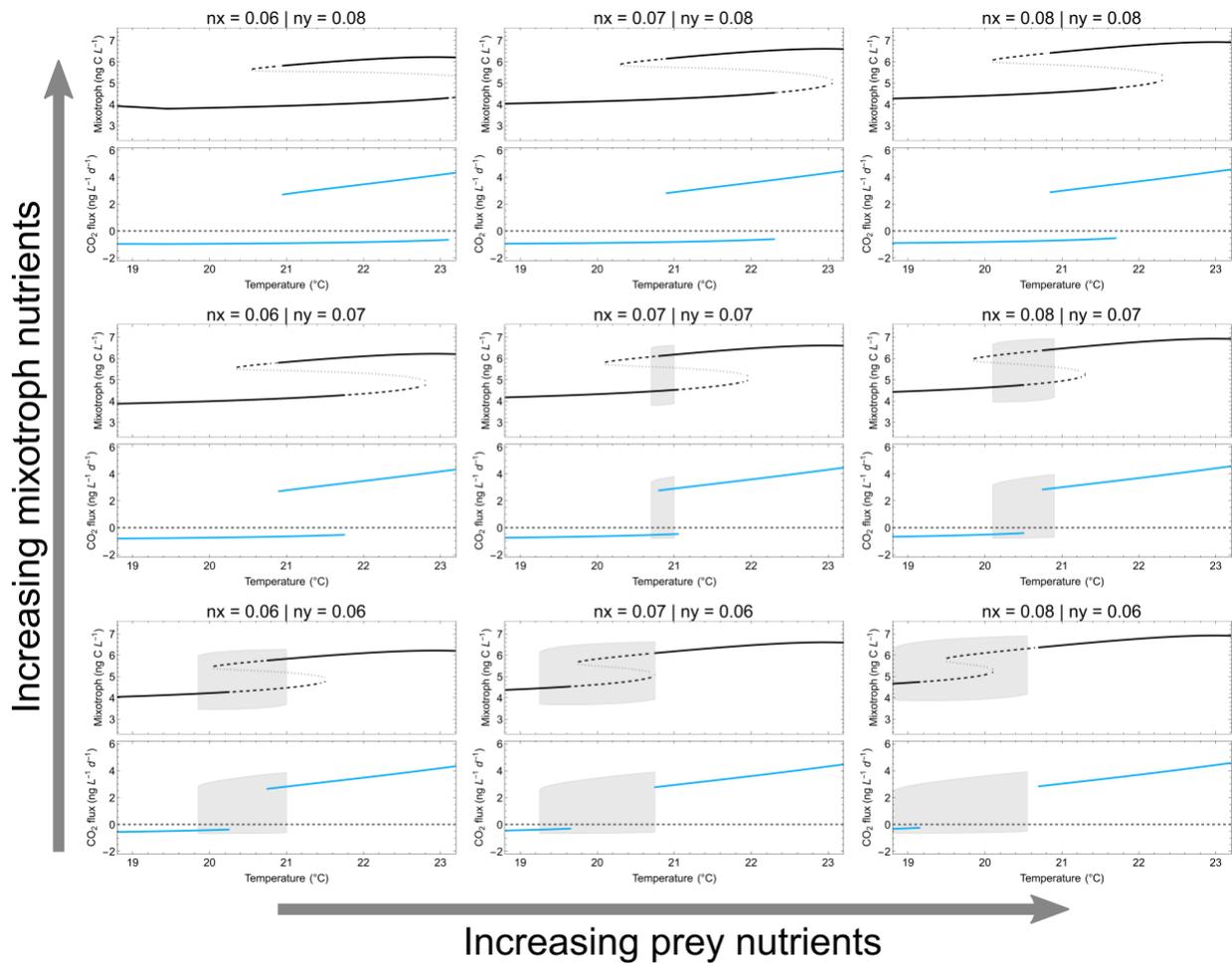
759 shown here were $\tau = 10 \text{ ng L}^{-1} \text{ t}^{-1}$ and $N_{M,feed} = 0.75 \text{ ng L}^{-1}$. All other model parameters were the

760 same as used in the main results (Table 1).

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765 **Figure S3.** Equilibrium mixotroph densities and net CO₂ flux shown across temperatures and
 766 across gradients of nutrient concentrations for nutrients utilized by a mixotroph (vertical) and its
 767 prey (horizontal). The center column corresponds to panels c–e in Figure 3 in the main text.

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