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3	Direct and context-dependent effects of light, temperature, and
4	phytoplankton shape bacterial community composition
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18	Running head: Direct and context-dependent effects

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19 Abstract

Species interactions, environmental conditions, and stochastic processes work in concert to bring 20 about changes in community structure. However, the relative importance of specific factors and 21 22 how their combined influence affects community composition remain largely unclear. We conducted a multi-factorial experiment to 1) disentangle the direct and interaction-mediated 23 effects of environmental conditions and 2) augment our understanding of how environmental 24 context modulates species interactions. We focus on a planktonic system where interactions with 25 phytoplankton effect changes in the composition of bacterial communities, and light and 26 temperature conditions can influence bacteria directly as well as through their interactions with 27 phytoplankton. Epilimnetic bacteria from two humic lakes were combined with phytoplankton 28 assemblages from each lake ("home" or "away") or a no-phytoplankton control and incubated 29 30 for 5 days under all combinations of light (surface, ~25% surface irradiance) and temperature (5 levels from 10°C to 25°C). Observed light effects were primarily direct while phytoplankton and 31 temperature effects on bacterial community composition were highly interdependent. The 32 influence of temperature on aquatic bacteria was consistently mediated by phytoplankton and 33 most pronounced for bacteria incubated with "away" phytoplankton treatments, likely due to the 34 availability of novel phytoplankton-derived resources. The effects of phytoplankton on bacterial 35 community composition were generally increased at higher temperatures. Incorporating 36 mechanisms underlying the observed interdependent effects of species interactions and 37 38 environmental conditions into modeling frameworks may improve our ability to forecast ecological responses to environmental change. 39

Keywords: community succession, algal-bacterial interactions, context-dependence, microcosm,
bacterioplankton, lake

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43 Introduction

Species interactions, environmental conditions, and stochastic processes affect the 44 abundance, diversity, and distribution of organisms in the environment. Determining the relative 45 importance of specific factors and how they interact to structure communities remains a major 46 challenge in ecology (Agrawal et al. 2007, Sutherland et al. 2012). Effects of environmental 47 conditions on a population include direct effects that can potentially be characterized by single-48 species observations and experiments as well as indirect effects that depend on species 49 interactions and require more community-focused approaches to detect and predict (Gilman et al. 50 2010). Concurrently, the outcome (e.g., mutualism, parasitism) and strength of interspecific 51 interactions can change depending on environmental conditions as well as community 52 composition (He et al. 2013, Chamberlain et al. 2014). To predict how communities will respond 53 54 to seasonal environmental fluctuations and long-term changes in climate, it is necessary to account for species interactions that mediate and are affected by environmental conditions. 55 The signal of changing environmental conditions can be muted, augmented, or relayed by 56 interactions with other species including mutualism, competition, parasitism, and food web 57 interactions (Gilman et al. 2010). Interaction-mediated effects of environmental change have 58 consequences for local abundances and geographic distribution as well as phenology (Miller-59 Rushing et al. 2010, Wisz et al. 2012, HilleRisLambers et al. 2013). For example, competitive 60 interactions, predators, consumers, disease agents, or the absence of a mutualist partner may 61 62 counteract environmental changes expected to expand the geographic range or increase the local abundance of a species (Gilman et al. 2010, HilleRisLambers et al. 2013). Additionally, shifts in 63 the phenology of one species can result in temporal mismatch and population declines in a 64 65 mutualist partner or consumer species (Winder and Schindler 2004, Edwards and Richardson

66 2004). In many cases, effects of environmental changes mediated by species interactions are
67 expected to be more consequential than direct effects of change (Davis et al. 1998).

In addition to species interactions modulating the signal of environmental change, 68 environmental conditions can alter the outcome and strength of context-dependent species 69 interactions. A change in outcome occurs when the effect of an interaction on a given species 70 (positive, negative, or neutral) depends on the environment. For example, the net costs of plants 71 associating with mycorrhizae can exceed the net benefits when nutrient availability is high or 72 light availability is low (Johnson et al. 1997). Context-dependent outcomes have been observed 73 for competition, mutualism, and predation, but are most frequently detected for mutualisms 74 (Chamberlain et al. 2014). In contrast, variation in interaction strength, or the magnitude of effect 75 sizes, across contexts is relatively consistent among interaction types (Chamberlain et al. 2014). 76 77 Population-level interactions can cause shifts in community structure, making it important to investigate the consequences of context-dependence at the community level. 78 Planktonic communities from temperate lakes are well suited for investigating 79

interactions between biotic and environmental factors. These communities undergo annually 80 repeated seasonal succession driven both by species interactions and environmental conditions, 81 and respond rapidly to experimental treatments (Kent et al. 2006, 2007, Sommer et al. 2012, 82 Weisse et al. 2016). Moreover, lakes have been described as sentinels of climate change, in part, 83 because they respond rapidly to changes in the environment (Adrian et al. 2009). Light 84 85 availability and temperature are two ecologically important factors that respond to environmental change in lakes. Increases in dissolved organic carbon concentration, as have been observed in 86 lakes across parts of North America and Europe (Monteith et al. 2007), decrease light availability 87 88 due to more rapid light attenuation and shift the vertical distribution of heat towards the surface

(Bukaveckas and Robbins Forbes 2000, Read and Rose 2013). Global mean lake surface
temperature has been increasing 0.34°C per decade in response to climate forcing (O'Reilly et al.
2015). Long-term climate changes are overlaid on a dynamic system where light availability and
temperature change with depth and over the course of a year, especially in lakes that stratify in
the summer and mix in the fall and spring.

Light and temperature can influence bacterial community composition directly as well as 94 through bacterial interactions with phytoplankton. Wavelength-specific light attenuation in the 95 water column creates a spectrum of niches for phototrophs (Stomp et al. 2007), including 96 97 photosynthetic organisms as well as photoheterotrophs (Martínez-García et al. 2011, Evans et al. 2015). Additionally, bacterial growth is affected in a strain-specific manner by light in the 98 ultraviolet range (Agogue et al. 2005, Hörtnagl et al. 2010). Bacteria have diverse optimal 99 100 growth temperatures and ranges, such that temperature can determine outcomes of competition between bacterial populations (Upton et al. 1990, Hall et al. 2008). Phytoplankton interact with 101 bacterioplankton through mechanisms that include selective grazing by mixotrophic 102 phytoplankton (Flynn et al. 2012), serving as a habitat for bacterial epiphytes (Jasti et al. 2005), 103 and providing species-specific resources as detritus (Van Hannen et al. 1999) or labile exudates 104 released by living cells (Teeling et al. 2012, Sarmento and Gasol 2012). As temperature 105 increases, mixotrophic phytoplankton are theorized to become more heterotrophic; this has been 106 experimentally demonstrated with the chrysophyte Ochromonas sp. (Wilken et al. 2012). 107 108 Additionally, the concentration and composition of extracellular organic carbon excreted by 109 phytoplankton depend on light and temperature conditions (Zlotnik and Dubinsky 1989, Parker and Armbrust 2005). 110

111 Our objective was to characterize the combined effects of phytoplankton, light, and 112 temperature on bacterial community composition from two humic lakes in Northern Wisconsin where interactions with phytoplankton are partially responsible for orchestrating changes in 113 114 bacterial composition through time (Kent et al. 2007, Paver et al. 2013). Change in light and temperature with depth is especially pronounced in darkly stained humic lakes (Huovinen et al. 115 2003). We specifically aimed to determine 1) whether the influence of light and temperature on 116 bacterial communities is mediated by phytoplankton and 2) whether the effects of interactions 117 with phytoplankton depend on light and temperature conditions (Fig. S1). If light and 118 temperature affect bacteria through interactions with phytoplankton, then it is expected that the 119 120 variation in bacterial community composition explained by light and temperature treatment will be greater in microcosms where phytoplankton are present relative to those where phytoplankton 121 122 are absent. If phytoplankton effects depend on the light and temperature context, the variation in bacterial community composition explained by phytoplankton will change under different light 123 and temperature conditions. 124

125

126 Materials and Methods

127 Study sites

South Sparkling Bog (SSB; 46°00'13.6"N, 89°42'19.9"W) and Trout Bog (TB;

46°02'27.5"N, 89°41'09.6"W) are two north temperate humic lakes in Vilas County, Wisconsin

that have been studied as part of the North Temperate Lakes Microbial Observatory. SSB and TB

- 131 were selected for their similarity in maximum depth (~8m) and differences in phytoplankton
- 132 community composition (Paver et al. 2013). These lakes are characterized by acidic pH and high

levels of dissolved organic carbon (Paver et al. 2013).

134 Experimental design

135	We conducted a multi-factorial microcosm experiment to determine the direct and
136	interactive effects of phytoplankton presence and composition, temperature, and light on
137	bacterial community composition. On 6 July 2011, microorganisms were collected from SSB and
138	TB integrated epilimnion samples (0-1m). Filtration through a 1µm Polycap AS cartridge filter
139	(Whatman, Piscataway, NJ, USA) was used to separate bacteria from larger organisms.
140	Phytoplankton assemblages were collected by filtering lake water through a $100\mu m$ nylon mesh
141	(Spectrum Laboratories, Rancho Dominguez, CA, USA) to remove zooplankton and collecting,
142	then rinsing phytoplankton cells captured on a $20\mu m$ nylon mesh with SSB water filter-sterilized
143	through a 0.2µm Polycap AS cartridge filter (Spectrum Laboratories), which allowed smaller
144	organisms such as heterotrophic nanoflagellates and bacteria to pass through. Phytoplankton
145	collected on 20µm mesh were resuspended in 0.2µm filter-sterilized water from SSB,
146	concentrating phytoplankton from 40L of lake water to 2.5L of sterilized water. All
147	combinations of bacteria from each lake (5L of $1\mu m$ filtered water) were combined with 0.25L
148	of concentrated phytoplankton from one of the two lakes lake or a no-phytoplankton control
149	(0.25L of 0.2µm filter-sterilized SSB water) in triplicate 10L LDPE cubitainers (I-Chem,
150	Rockwood, TN, USA). Combined bacteria and phytoplankton were gently inverted to mix and
151	then partitioned into 500ml clear glass bottles (Wheaton, Millville, NJ, USA) in a predetermined,
152	randomized order (33 bottles/ treatment).
153	For each bacteria-phytoplankton combination, three bottles were used to characterize the
154	initial community composition and three bottles were incubated for five days under each of five
155	temperatures and two light levels (Fig. 1). Temperature treatments were established and
156	maintained by continuously pumping defined proportions of high temperature (~25°C) surface

 S2). High light and low light treatments were established by incubating both bottom (~25% of surface irradiance) of each floating container incubator. I conditions were monitored throughout the incubation using HOBO light an pendant data loggers (Onset, Pocasset, MA, USA) with three loggers placed two at opposite corners at the surface and one on the bottom. <i>Microbial community analysis</i> Microorganisms from initial samples and from each bottle microcost incubation were concentrated onto 0.22µm filters (Supor-200; Pall Gelman frozen at -20°C. DNA was extracted using FastDNA purification kits (MP) OH, USA). Bacterial community composition was characterized using auto intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stree 	157	water and low temperature (~5°C) subsurface water (1:0, 3:1, 1:1, 1:3, 0:1) into floating plastic
 bottom (~25% of surface irradiance) of each floating container incubator. I conditions were monitored throughout the incubation using HOBO light an pendant data loggers (Onset, Pocasset, MA, USA) with three loggers place two at opposite corners at the surface and one on the bottom. <i>Microbial community analysis</i> Microorganisms from initial samples and from each bottle microcos incubation were concentrated onto 0.22µm filters (Supor-200; Pall Gelman frozen at -20°C. DNA was extracted using FastDNA purification kits (MP OH, USA). Bacterial community composition was characterized using auto intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	158	container incubators (0.73m x 0.53m x 0.46m, The Container Store, Coppell, TX, USA) (Fig.
 conditions were monitored throughout the incubation using HOBO light an pendant data loggers (Onset, Pocasset, MA, USA) with three loggers placed two at opposite corners at the surface and one on the bottom. <i>Microbial community analysis</i> Microorganisms from initial samples and from each bottle microcos incubation were concentrated onto 0.22µm filters (Supor-200; Pall Gelman frozen at -20°C. DNA was extracted using FastDNA purification kits (MP OH, USA). Bacterial community composition was characterized using auto intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	159	S2). High light and low light treatments were established by incubating bottles at the surface and
 pendant data loggers (Onset, Pocasset, MA, USA) with three loggers placed two at opposite corners at the surface and one on the bottom. <i>Microbial community analysis</i> Microorganisms from initial samples and from each bottle microcos incubation were concentrated onto 0.22µm filters (Supor-200; Pall Gelman frozen at -20°C. DNA was extracted using FastDNA purification kits (MP OH, USA). Bacterial community composition was characterized using auto intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	160	bottom (~25% of surface irradiance) of each floating container incubator. Light and temperature
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Microorganisms from initial samples and from each bottle microcost incubation were concentrated onto 0.22µm filters (Supor-200; Pall Gelman frozen at -20°C. DNA was extracted using FastDNA purification kits (MP OH, USA). Bacterial community composition was characterized using auto intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid	163	two at opposite corners at the surface and one on the bottom.
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OH, USA). Bacterial community composition was characterized using auto intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid	166	incubation were concentrated onto $0.22 \mu m$ filters (Supor-200; Pall Gelman, East Hills, NY) and
 intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	167	frozen at -20°C. DNA was extracted using FastDNA purification kits (MP Biomedicals, Solon,
 (2013). Fluorescently labeled ARISA PCR amplicons were combined with bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	168	OH, USA). Bacterial community composition was characterized using automated ribosomal
bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesbord by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid	169	intergenic spacer analysis (ARISA) (Fisher and Triplett 1999) as described by Paver et al.
by the Keck Center for Functional Genomics at the University of Illinois vi electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid	170	(2013). Fluorescently labeled ARISA PCR amplicons were combined with a custom $100 - 1250$
 electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosyste California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonom GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	171	bp Rhodamine X-labeled internal size standard (Bioventures, Murfreesboro, TN) and analyzed
California, USA). Electropherograms from each sample were aligned and p fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid	172	by the Keck Center for Functional Genomics at the University of Illinois via denaturing capillary
fluorescence units were sized and grouped into bins of operational taxonon GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid	173	electrophoresis using an ABI 3730XL Genetic Analyzer (Applied Biosystems Inc., Carlsbad,
 GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	174	California, USA). Electropherograms from each sample were aligned and peaks greater than 500
 correspond to chloroplasts were removed from the analysis. The signal stre normalized to account for run-to-run variations in signal detection by divid 	175	fluorescence units were sized and grouped into bins of operational taxonomic units using
178 normalized to account for run-to-run variations in signal detection by divid	176	GeneMarker version 1.95 (SoftGenetics, State College, PA, USA). ARISA fragments known to
	177	correspond to chloroplasts were removed from the analysis. The signal strength of each peak was
individual peaks by the total fluorescence (area) detected in each profile.	178	normalized to account for run-to-run variations in signal detection by dividing the area of
	179	individual peaks by the total fluorescence (area) detected in each profile.

180 Statistical approach

181	Pairwise Bray-Curtis similarities were calculated for every combination of samples using
182	Hellinger-transformed ARISA data and visualized using non-metric multidimensional scaling in
183	PRIMER version 6 (PRIMER-E Ltd, Plymouth Marine Laboratory, UK) (Clarke and Warwick,
184	2001). Permutational multivariate analysis of variance (PERMANOVA) was used to test: 1) the
185	effects of light and temperature on bacterial community composition following incubation for
186	each combination of bacteria and phytoplankton and 2) the effect of each phytoplankton
187	treatment (compared to the no phytoplankton control) on bacterial community composition at
188	each light and temperature level, stratified by the bacterial community source lake.
189	PERMANOVA is a non-parametric multivariate analysis of variance that generates p-values
190	using permutations (McArdle and Anderson 2001, Anderson 2001). PERMANOVA tests were
191	run using the adonis function from the vegan package (Oksanen et al. 2011) in the R statistical
192	environment (R Core Development Team, 2010).
193	
194	Results

195 Direct and phytoplankton-mediated light and temperature effects

Over the five-day incubation, phytoplankton presence and composition, light availability, and temperature affected bacterial community composition (Fig. 2). We assessed direct effects of light and temperature on bacterial communities by analyzing changes in bacterial community composition in no-phytoplankton control treatments. Light had a significant direct effect on the composition of bacterial communities from both lakes (Fig. 3). In contrast to light, temperature had a significant direct effect on the community composition of bacteria from SSB, but not bacteria from TB (Fig. 3). 203 The effects of light and temperature on bacteria incubated with phytoplankton depended on the specific combination of phytoplankton and bacteria (Fig. 3). When TB bacteria were 204 incubated with their "home" phytoplankton, there was a significant interaction between light and 205 206 temperature. Light explained small but significant variation in bacterial community composition when SSB bacteria were combined with their "home" phytoplankton. Significant light effects 207 were not detected when bacteria were combined with phytoplankton from the "away" lake. In 208 contrast, temperature had a consistently significant effect on the composition of bacterial 209 communities when phytoplankton were present. Notably, the variation in bacterial community 210 composition explained by temperature was higher for bacteria incubated with phytoplankton 211 from the "away" lake than for bacteria incubated with phytoplankton from their "home" lake. 212

213 *Context-dependence of phytoplankton interactions*

214 We used percent variation in bacterial community composition explained by phytoplankton treatment to evaluate the strength of phytoplankton effects (Fig. 4). Prior to 215 216 incubation, 22% of the variation in bacterial community composition in SSB phytoplankton and corresponding control treatments was explained by phytoplankton treatment. Following 217 incubation, variation explained by SSB phytoplankton was only greater than the initial explained 218 variation in microcosms incubated at the highest temperature (45% variation explained). In 219 contrast, variation in bacterial community composition in TB phytoplankton and corresponding 220 control treatments due to phytoplankton treatment was not significant prior to incubation 221 (p>0.05). At the coldest two temperatures, 13% and 19% more variation in bacteria community 222 composition was explained by TB phytoplankton in low light compared to high light treatments. 223 As temperature increased, the variation explained by TB phytoplankton generally increased until 224 225 peaking at the second warmest temperature.

226 Discussion

It is well established that species interactions and environmental conditions act in concert 227 to affect community composition, but how these factors combine to determine community 228 229 composition is largely undefined. In this study we selected light, temperature, and phytoplankton-bacterial interactions to investigate the interplay among biotic and environmental 230 factors at the community level. In general, observed treatment effects were highly similar across 231 replicates, suggesting that deterministic processes controlled the development of bacterial 232 communities over the course of the five-day experimental incubation. We observed direct light 233 effects, phytoplankton-mediated temperature effects, and temperature- and light-dependent 234 phytoplankton effects, each of which is explored below. 235

236 Direct light effects

237 Light had a consistent, direct effect on bacterial community composition in experimental microcosms. One potential explanation for the direct effect of light on bacterial composition is 238 selection for phototrophic bacteria generally, or specific types of phototrophic bacteria. 239 240 Organisms that harvest light energy are adapted to use specific wavelengths of light and, as a result, phototroph distribution within the water column reflects the available light spectrum (Vila 241 and Abella 2001, Haverkamp et al. 2008). As our study lakes have high humic content, light 242 spectra are enriched in long wavelength photons (>600nm) and light attenuates rapidly with 243 increased depth, creating specific spectral niches across modest changes in depth (e.g., 0.5m) 244 245 (Vila et al. 1998). Photoheterotrophic bacterial cells can comprise a sizable portion of lake bacterial communities, and change in abundance with depth (Mašín et al. 2008, Martínez-García 246 et al. 2011, Lew et al. 2016). Alternatively, or perhaps additionally, direct effects of light may 247 248 have been caused by ultraviolet radiation. Based on the transparency properties of borosilicate

249 glass, plankton were exposed to a fraction of long wave ultraviolet A radiation, but none of the 250 ultraviolet B and shorter wavelengths of light, penetrating to their respective incubation depths (Döhring et al. 1996). Effects of ultraviolet radiation can decrease the growth efficiency of 251 252 certain bacterial strains while having no effect or increasing the growth efficiency of other strains (Hörtnagl et al. 2010). It is additionally possible that products of dissolved organic matter 253 photolysis, including low molecular weight dissolved organic matter and reactive oxygen species 254 induced changes bacterial community composition (Glaeser et al. 2010, Paul et al. 2011). 255 *Phytoplankton-mediated temperature effects* 256

In contrast to the effects of light, observed effects of temperature were primarily 257 mediated through interactions with phytoplankton. Temperature had significant direct effects on 258 bacterial community composition for SSB bacteria, but only marginally significant direct effects 259 260 for TB bacteria. Lack of significant direct effects of temperatures spanning approximately 15°C on TB bacteria was surprising and may have been due to resource limitation. Alternatively, 261 differences in the chemistry of the lake water added along with the bacterial treatment or the 262 distribution of bacterial traits (e.g., ability to use light energy or breakdown available dissolved 263 organic matter) may explain observed lake-specific differences in temperature response. 264 Concentrations of dissolved organic carbon, total phosphorus, and total nitrogen tend to be 265 higher in TB compared to SSB (Paver et al. 2013). Temperature effects were consistently 266 significant in treatments with added phytoplankton, and enhanced in "away" phytoplankton 267 268 treatments, potentially due to production of organic matter novel to the bacterial community 269 (Fogg 1983, Sarmento and Gasol 2012). Bacterial community composition is continuously shaped by interactions with phytoplankton from their "home" lake (Paver et al. 2013), so initial 270 271 bacterial assemblages were acclimated to "home" phytoplankton resources. Our results provide

further evidence that bacteria rely on phytoplankton-derived dissolved organic carbon in darkly
stained humic lakes with high background concentrations of dissolved organic carbon (Kritzberg
et al. 2006, Kent et al. 2007).

275 Observed phytoplankton-dependent temperature effects along with previously published findings support a signal transduction hypothesis where changes in the environment are relayed 276 to bacterial populations through their interactions with phytoplankton. In a suite of north 277 temperate humic lakes (including SSB and TB), changes in the composition of phytoplankton 278 assemblages were largely explained by environmental (e.g., water temperature, nutrients) and 279 meteorological (e.g., photosynthetically active radiation, precipitation) factors (Kent et al. 2007). 280 In contrast, changes in bacterial community composition were primarily explained by changes in 281 phytoplankton population abundances and covariation between phytoplankton populations and 282 283 the environment (Kent et al. 2007). Results from the current study provide experimental evidence that the signal of increased temperature in these lakes is largely relayed to bacteria by 284 their interactions with phytoplankton. Experimental observations of Baltic Sea bacterial 285 communities yielded a complementary result that phytoplankton bloom stage was a more 286 important factor structuring bacterial communities than a 6°C change in temperature (Scheibner 287 et al. 2013). Previous work on bacterial growth and activity provide additional support for 288 environmental signal transduction via phytoplankton. Multiple regression and hierarchical 289 partitioning analysis of data from 300 field studies indicated that temperature has a positive 290 291 relationship with phytoplankton primary production, but not bacterial production (Faithfull et al. 2011). Despite positive correlations between temperature and bacterial production, bacterial 292 production was primarily explained by a combination of total phosphorus and primary 293 294 productivity (Faithfull et al. 2011). At low temperatures, bacterial growth and activity are

frequently temperature limited (Simon and Wünsch 1998, Vrede 2005, Adams et al. 2010).

296 When temperature is not limiting, bacterial growth in temperate lakes is commonly limited by

297 phosphorus, dissolved organic carbon, or the two in combination (Vrede 2005). Phytoplankton

clearly support bacterial growth and activity and have the potential to relay environmental

signals to bacteria with which they interact.

300 Temperature- and light- dependent of phytoplankton effects

Effects of phytoplankton on bacterioplankton were temperature, and to a lesser extent, 301 light dependent. Observed interdependence of temperature and phytoplankton effects is 302 consistent with the framework that phytoplankton provide organic matter resources to bacteria 303 (Cole 1982, Sarmento and Gasol 2012) and temperature regulates the flow of carbon from 304 phytoplankton to bacteria (Overmann 2013, Scheibner et al. 2013). Notably, when bacteria from 305 306 TB were incubated with "home" phytoplankton, there was a significant interaction between light and temperature (Fig. 3). In contrast to high light assemblages that became increasingly different 307 from their initial composition along a somewhat linear trajectory in ordination space as 308 temperature increased, low light assemblages exhibited a curved response (Fig. 2). Bacterial 309 composition in the coldest low light treatment was uncharacteristically different from the initial 310 assemblage relative to other assemblages incubated at that temperature. These observations may 311 be explained by changes in the concentration and composition of phytoplankton exudates under 312 different temperature and light conditions. Combined effects of light and temperature have been 313 314 shown to affect the dominant metabolic pathways used to process carbon, thereby controlling exudate release (Parker and Armbrust 2005). Alternatively, the dominant mechanism of bacteria-315 phytoplankton interactions may change depending on light and temperature. For example, low 316

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317	light conditions	can increase	the rate c	of bacterial	consumption b	v mixotrop	hic phytoplankton
		••••••••••••••••			•••••••••••••••••••	,	

under certain conditions (e.g., low nutrient availability; [McKie-Krisberg and Sanders 2014]).

319 *Consequences for planktonic microbial communities*

Observations that light has primarily direct effects on bacterioplankton, while 320 temperature effects are mediated through interactions with phytoplankton, have implications for 321 interpreting seasonal changes in planktonic communities and forecasting future changes. The 322 pronounced, direct effect of light on bacterial community composition emphasizes the 323 importance of collecting high-resolution samples over depth and incorporating mechanisms 324 structuring bacterial communities into a depth-specific framework. Temperature is frequently 325 correlated with the succession of aquatic bacterial communities (Crump and Hobbie 2005, 326 Fuhrman et al. 2006, Shade et al. 2007). Our results demonstrate that, in some systems, much of 327 328 the bacterial community response to temperature is fueled by their interactions with phytoplankton. Thus, if the objective is to forecast how bacterial community structure and 329 function will change in response to environmental changes, it is necessary to incorporate the 330 predicted response of phytoplankton and context-dependence of interactions linking 331 phytoplankton and bacterial assemblages. For example, elevated temperature in mesocosms 332 during the spring phytoplankton bloom in Kiel Bight accelerated the onset of the phytoplankton 333 bloom, decreased the intensity of maximum chlorophyll a and particulate organic carbon by 334 approximately 20%, and caused dissolved organic carbon concentrations to increase more 335 336 rapidly than under ambient temperature conditions (Biermann et al. 2014).

Environmental context is critical for understanding how planktonic communities will change over time and in response to environmental change. Many of our observations were highly dependent on the biotic or environmental context – responses seen in one lake were not

340 replicated in the other and, for TB bacteria combined with TB phytoplankton, response to temperature depended on light availability. Context-dependence has also been described for 341 bacterial production in high mountain lakes where bacterial response to solar radiation treatments 342 343 depended on the presence of phytoplankton and whether bacteria were phosphorus-limited (Medina Sánchez et al. 2002). The prevalence of non-additive interaction effects among 344 environmental factors emphasizes the importance of multi-factorial experimental investigations 345 into drivers of microbial community composition and activity. It additionally underscores a need 346 to identify mechanisms underlying context-dependent microbial responses and build these 347 mechanisms into frameworks describing aquatic microbial community dynamics. 348 Effects of interacting factors on community composition 349 Phytoplankton, light availability, and temperature act in concert to bring about changes in 350 351 bacterial community composition over time. Light availability directly affects bacterial

community composition while interactions with phytoplankton amplify or relay the signal of 352 increasing temperature to bacteria. The strength of phytoplankton interactions with bacteria, 353 inferred through comparisons of bacterial community composition across temperature and light 354 355 treatments relative to no-phytoplankton controls, depends on temperature. For certain combinations of phytoplankton and bacteria, the outcome of phytoplankton interactions appears 356 to additionally be light dependent. The enhanced effect of "away" phytoplankton relative to 357 "home" phytoplankton and the lack of consistent temperature effects in treatments without 358 359 phytoplankton provide strong support for phytoplankton resources shaping bacterial communities and their response to environmental conditions. These findings emphasize the 360 importance of observing population and community responses to multiple ecological drivers 361 362 simultaneously and under a range of environmentally relevant conditions. Studies aimed at

363 understanding the effects of climate change frequently compare ambient temperature conditions to an elevated temperature treatment. Our results suggest that there are inflection points in 364 community responses to temperature that would be overlooked in ambient vs. elevated 365 366 temperature comparisons. The idea that under-sampling the range of potential temperatures constrains our ability to make general inferences about temperature effects is reinforced by 367 studies that have investigated the effect of temperature on bacterial production and observed two 368 temperature optima (Simon and Wünsch 1998, Adams et al. 2010). The problem of under-369 sampling treatment levels is potentially problematic for other factors as well, including light 370 371 availability (Gu and Wyatt 2016). Movement away from context-specific observations towards generalizable, theoretical advances and identifying parameters that can be incorporated into 372 predictive frameworks will depend on investigations such as this into the mechanisms driving 373 374 observed changes in community composition.

375

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549 Figures

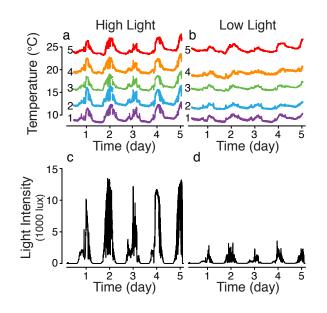
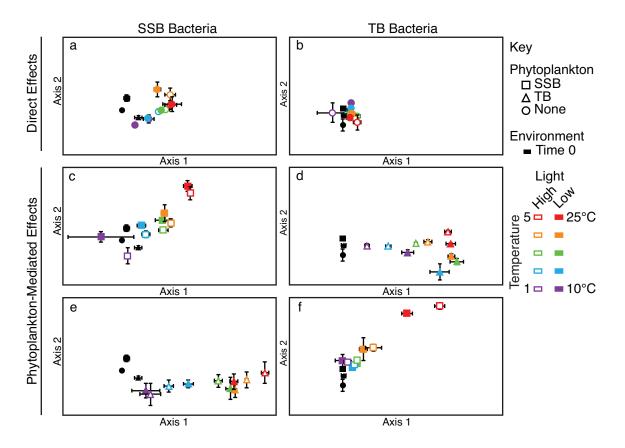


Figure 1. Temperature (a, b) and average light intensity (c, d) at the top (high light; a, c) and bottom (low

- light; b, d) of each temperature incubator over the five-day incubation. Temperature treatments are
- 553 labeled 1 (coldest) to 5 (warmest).



554

555 Figure 2. Non-metric multidimensional scaling ordination of bacterial community composition in -

556 microcosms (average ± standard error) with SSB bacteria (a,c,e; stress value=0.08) and TB bacteria (b,d,f;

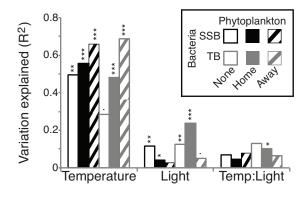
stress value=0.14) before and after incubation. To simplify depiction of overlapping treatments,

community composition in no-phytoplankton control microcosms following incubation is shown in plots

a and b, community composition in "home" phytoplankton treatments is shown in plots c and d, and

community composition in "away" phytoplankton treatments is shown in plots e and f. Community

composition before incubation (Time 0) is included in all plots for reference.

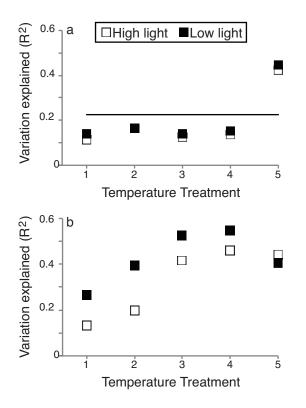


562

563 Figure 3. Variation in bacterial community composition explained by temperature, light, and the

interaction between temperature and light for each combination of phytoplankton and bacteria. *P*

values below 0.01 are indicated by symbols (*** <0.001, ** <0.01, * < 0.05, . <0.01).



566

Figure 4. Variation in bacterial community composition explained by SSB phytoplankton
treatment (a) and TB phytoplankton treatment (b) across all temperature and light conditions. A
horizontal line indicates variation in bacterial community composition explained by SSB
phytoplankton prior to incubation. TB phytoplankton did not explain significant variation in
bacterial community composition before incubation.