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1 IS THE FUTURE BLUE-GREEN? A REVIEW OF THE CURRENT MODEL

2 PREDICTIONS OF HOW CLIMATE CHANGE COULD AFFECT PELAGIC

3 FRESHWATER CYANOBACTERIA

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9 ABSTRACT

10 There is increasing evidence that recent changes in climate have had an effect on lake 11 phytoplankton communities and it has been suggested that it is likely that Cyanobacteria will 12 increase in relative abundance under the predicted future climate. However, testing such a 13 qualitative prediction is challenging and usually requires some form of numerical computer 14 model. Therefore, the lake modelling literature was reviewed for studies that examined the 15 impact of climate change upon Cyanobacteria. These studies, taken collectively, generally 16 show an increase in relative Cyanobacteria abundance with increasing water temperature, 17 decreased flushing rate and increased nutrient loads. Furthermore, they suggest that whilst 18 the direct effects of climate change on the lakes can change the timing of bloom events and 19 Cyanobacteria abundance, the amount of phytoplankton biomass produced over a year is not 20 enhanced directly by these changes. Also, warmer waters in the spring increased nutrient 21 consumption by the phytoplankton community which in some lakes caused nitrogen 22 limitation later in the year to the advantage of some nitrogen-fixing Cyanobacteria. Finally, it 23 is also possible that an increase in Cyanobacteria dominance of the phytoplankton biomass 24 will lead to poorer energy flow to higher trophic levels due to their relatively poor edibility 25 for zooplankton.

26

27 KEYWORDS:

28 lake modelling, nitrogen limitation, phenology, water quality, eutrophication

30 1. INTRODUCTION

31 In recent years, there has been increased concern in the field of limnology about how climate 32 change may affect phytoplankton populations. This is a logical area of interest, given the 33 way that climate affects the temperature and physical structure of a lake, as well as numerous 34 in-lake chemical (e.g. dissolved oxygen concentrations) and biological processes (e.g. 35 through water temperature) (Kalff 2002). However, out of all the phytoplankton species that 36 make up the lake communities of the world, it is perhaps those species that fall under the 37 phylum Cyanobacteria that have caused the greatest amount of concern and speculation about 38 how climate change may affect them (Paerl and Huisman, 2008).

39 Cyanobacteria are photosynthetic prokaryotes that used to be referred to as blue-green 40 algae. In lakes, they generally form large colonies or filaments and many species possess the 41 ability to be buoyant through intracellular gas vesicles (Reynolds, 2006). Although this 42 property can in itself lead to unsightly blooms forming near the lake surface, the so-called 43 algal scums, it is their ability to produce toxins that concerns humans the most. There are 44 several types of toxins produced including hepatoxins, neurotoxins and cytotoxins (Codd, 45 Morrison and Metcalf, 2005). Hepatoxic microcystins damage the digestive tract and liver, 46 and in humans can cause pneumonia-like symptoms, whereas neurotoxins affect the nervous 47 system. Cytotoxins cause widespread necrotic injury in mammals (e.g. liver, kidneys, lungs, 48 spleen, intestine) and are also genotoxic, causing chromosome loss and DNA strand breakage 49 (Codd, Morrison and Metcalf, 2005) (for more information, see Chapter 3 in Chorus and 50 Bartram, 1999). Such has been the recognition in recent decades of the threat posed by these 51 toxins, the World Health Organisation (WHO) has produced a specific report on the topic 52 (Chorus and Bartram, 1999).

53 The general view held for Cyanobacteria is that they grow better at higher 54 temperatures (>25 °C), although there are exceptions at lower temperatures (see Reynolds and Walsby, 1975) and in lakes that experience low winter flushing (Hendry et al. 2006). Of 55 56 course, in the field such high temperatures usually occur in lakes at the same time as 57 increasing stratification which allows Cyanobacteria with buoyancy regulating properties to 58 appear in near-surface waters. Therefore discerning whether temperature or stratification (or both) are the key driver to the formation of a large Cyanobacteria bloom can be difficult 59 60 (Reynolds and Walsby, 1975). Regardless, the positive connection between higher 61 temperatures and increased Cyanobacteria success (e.g. biomass and/or dominance of the 62 phytoplankton community) would seem to mean that the predicted warmer world of the late 63 21st century (IPCC, 2007) will be more suitable for these phytoplankton. However, in order 64 to test such a prediction we need to subject lakes to future conditions and one of the best 65 ways to do that is through using computer models.

66 Given their importance in affecting water quality, it is unsurprising the many lake 67 models include a Cyanobacteria component. However, given the interest in climate change 68 in recent years, it is surprising how few studies have used models to examine the potential 69 effect climate change could have on Cyanobacteria; perhaps this reflects the complexity of 70 modelling phytoplankton sub-groups and the confidence of modellers. Nevertheless, this 71 review collects together the published modelling evidence so far (Table 1) in order to gain a 72 collective synthesis of how climate change could affect Cyanobacteria, moving beyond 73 speculation based on present day observations and trying to predict the future responses of 74 these phytoplankton. The studies included had to meet the strict criteria of having used a 75 computer lake model, which included a Cyanobacteria component, and directly tested climate 76 change scenarios or the sensitivity of climate drivers (e.g. changing water temperature). The

review is structured by the approach used in the studies which fall into two broad categoriesdetailed below.

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- 80

81 2. PREDICTING CLIMATE CHANGE IMPACTS

82 2.1 Using Regional Climate Models (RCMs)

83 This method involves taking the future predictions of a climate model and using them to 84 drive a lake model that includes a Cyanobacteria element (e.g. species, taxonomic group). 85 However, usually the daily weather prediction covers an area much bigger than the lake 86 system being modelled (e.g. > 50-100km grids) and therefore some kind of downscaling is 87 required. Also, any predictions are limited to the particular climate scenario model used, 88 even when groups of different models are applied, giving only limited scope for examining where key thresholds of change might occur or how changes in other stressors unrelated to 89 90 the climate scenario may affect the response.

91 One of the earliest applications of this method for Cyanobacteria response predictions 92 was conducted by Howard and Easthope (2002) using CLAMM (Cyanobacteria Lake Mixing 93 Model). In this study, Microcystis growth in Farmoor reservoir (UK) was simulated using 90 94 years of future predicted output from the HADCM2 (see Jones et al., 1997) climate model. 95 Curiously, the key drivers used were wind speed, incoming solar radiation and cloud cover; 96 air temperature was not used. Consequently, as the main trend of change in the climatic 97 variables tested was only a slight decline in solar radiation due to an increase in cloud cover, 98 there was little forecasted change in Microcystis growth.

A more comprehensive study was conducted by Elliott et al. (2005), where the
outputs of HADCM2 were used to drive a smaller scale RCM and, after suitable

101 downscaling, provide weather drivers for the PROTECH model. PROTECH (Phytoplankton 102 RespOnses To Environment CHange) is a process-based lake phytoplankton community 103 model that can simulate 8-10 taxa (genus or species) and can include numerous types of 104 Cyanobacteria (see Reynolds et al., 2001 and Elliott et al., 2010 for details). In this study of 105 the phytoplankton community of Bassenthwaite Lake (UK), Anabaena, Aphanizomenon and 106 Planktothrix made up the Cyanobacteria element. The simulations first validated that using 107 20 years of the downscaled weather from a present (1970-1990) day climate scenario 108 produced the observed phytoplankton community and then tested the effect that 20 years of 109 future (2080-2100) climate had on the phytoplankton. Surface water temperature increased 110 on average 2.7 °C but the mixed depth was relatively unaffected. The Cyanobacteria 111 response was to grow earlier in the year (spring time) but there was a decline in their mean 112 biomass later in the year when they had previously been more abundant (Fig. 1). This effect 113 was due to nutrient limitation caused by an increased uptake of nutrients when growth was 114 enhanced in the spring; thus, as the nutrient-defined carrying capacity of the lake had not 115 been changed by the scenarios, the overall annual Cyanobacteria biomass produced remained 116 fairly constant and only the timing of its production was altered.

117 Of course, climate change is likely to affect the catchment that any given lake resides 118 in and two Swedish studies have sought to link climate, catchment and lake models. The first 119 (Arhiemer et al., 2005) examined the impact of several downscaled climate scenarios on the 120 Rönneå catchment and the eutrophic Lake Ringsjön (Sweden). The catchment part of the 121 study mainly focussed on nitrogen export to the lake which increased under all of the future 122 scenarios. The impact of this upon the lake was modelled using PROBE (PROgram for 123 Boundary layers in the Environment; Svensson, 1998) to simulate the lake physics coupled to 124 BIOLA (BIOgeochemical LAke model; Pers 2002) which includes Cyanobacteria as a whole 125 group rather than individual species. As in Elliott et al. (2005), the authors validated the

simulated phytoplankton driven by the present day climate against observed data, which produced a reasonable fit for the main summer bloom but simulated a spring bloom when none was observed. Despite thus, the relative differences between the present climate and future climates suggested a huge increase in Cyanobacteria biomass produced (>80% increase). The cause behind this response was mainly raised water temperatures (by 1-5 °C) stimulating an increase in nutrient mineralization and Cyanobacteria growth rates coupled to a higher nutrient load to the lake.

133 The second Swedish study (Markensten et al., 2010) coupled the catchment model, 134 GWLF (Generalised Watershed Loading Functions; Haith and Shoemaker, 1987) to PROBE 135 (Svensson, 1998) and PROTBAS (PROTech Based Algal Simulations; Markensten and 136 Pierson, 2007). The Galten basin of western Lake Mälaren (Sweden) was the study site and, 137 after validating the lake models against present day observations, a 21 year A2 climate 138 change scenario (assumes doubling of present CO₂ concentrations; IPCC 2001) was used to 139 test the potential climate change impacts. The effect of this scenario was to increase the 140 period of stratification (by >25%), reduce ice-cover and increase surface water temperatures. 141 The impact of this on the phytoplankton was to slightly increase the total biomass (+9%) and 142 Cyanobacteria dominance. The drivers identified for this change were the altered timing of 143 nutrient delivery to the lake rather than changes in water temperature and stratification. The 144 former, coupled to an extended growing season, increased the likelihood of nitrogen 145 limitation later in the year, to the advantage of the nitrogen-fixing Cyanobacteria.

A study of three lakes in New Zealand of different trophic status using the lake model
DYRESM-CAEDYM (DYnamic REservoir Simulation Model – Computational Aquatic
Ecosystem DYnamics Model; Hamilton and Schladow, 1997) also used this A2 scenario but
only the air temperature element (Trolle et al., 2011). After initial calibration and validation
against recent observations, only the eutrophic Lake Rotoehu was run with a Cyanobacteria

151 state variable. Under the future scenario, the Cyanobacteria showed an increase of >15% in 152 dominance due to an increase in water temperature and/or nutrient load to the lake. What 153 was especially interesting about this study, however, was that the future scenario was tested 154 under a range of nutrient loads which showed that, at least in terms of total chlorophyll *a*, the 155 tested climate scenario caused effects equivalent to increasing the nitrogen and phosphorus 156 load to the lake by 25-50%.

157

158 2.2 Using the sensitivity approach

Studies that use a sensitivity procedure take a present day simulation of a lake system and then run it again altering, for example, temperature and nutrient loading in a factorial design. This produces a range of "what if...?" scenarios and allows the exploration of two key drivers simultaneously. The outputs from the model runs can then be plotted on an X-Y-Z plot to reproduce a response surface for the variable concerned. The method also allows the identification of non-linear changes and thresholds.

The first modelling study to use this method in relation to climate change and 165 166 Cyanobacteria was Elliott et al. (2006). They examined the impact of changing nutrient 167 (phosphorus and nitrate) loads and water temperature upon the phytoplankton community of 168 Bassenthwaite Lake (UK). Focussing on just the Cyanobacteria part of the simulated 169 community, the impact of increased water temperature was clear. It caused the bloom to 170 become earlier (by 2 days per 1 °C increase) and increasing the maximum percentage 171 dominance of Cyanobacteria (by 7.6% per 1 °C increase) from a present day level of 17.3% 172 to 56.3% at +5 °C (Fig. 2). Importantly, the factorial nature of the study also showed that 173 these responses to temperature were enhanced by higher nutrient loads to the lake and, 174 conversely, suppressed by the lower nutrient scenarios.

175 Mooij et al. (2007) also used this factorial approach to test the effect of a wide range 176 of nutrient loadings and water temperature patterns upon a conceptual shallow lake using the 177 lake ecosystem model PCLake (e.g. Janse and Van Liere, 1995). The study found that the 178 Cyanobacteria part of PCLake responded favourably (e.g. % Cyanobacteria abundance rising 179 from 21 to 79%) to increasing temperature (particularly in the winter) but only if the nutrient 180 supply to the lake was above a critical threshold. More importantly, they concluded that this 181 threshold was lower under the warmer water scenarios compared to the control run under 182 present day temperatures. Furthermore, the model was run in two different states: 183 macrophyte-dominated clear state and phytoplankton dominated turbid state. Unsurprisingly, 184 Cyanobacteria dominated the latter state even under present day conditions and their 185 dominance was enhanced with the warming scenario. However, in the clear state this 186 response by the Cyanobacteria was greatly reduced, with little change in biomass and a 3-4 187 week shift in their bloom formation to earlier in the year. In general, though, the 188 consequence of this increased dominance by Cyanobacteria to the modelled food web was 189 that, because of their poor edibility, the flow of energy to higher trophic levels was reduced. 190 In another study, Loch Leven (UK) was examined using the PROTECH model to test 191 the response of its phytoplankton community to changes in water temperature and nutrient 192 supply (Elliott and May, 2008). The effect of increased water temperature upon annual mean 193 Cyanobacteria percentage abundance was very small (+1-2% per 1 °C increase) and generally 194 enhanced at the lower nutrient scenarios (which tested changing only phosphorus and 195 phosphorus and nitrogen together). The complex nature of this response was caused by the 196 lake experiencing low nitrate levels during the prime growing period for Cyanobacteria (July-197 September). As the dominant Cyanobacteria was the nitrogen-fixing taxon Anabaena, this 198 meant that they actually performed better under the lower nitrate/SRP scenarios because they 199 were the only phytoplankton in the simulations that could utilise the phosphorus from the

spring bloom that carried over to later in the year. However, the warmer scenarios also
caused more of the nutrients to be used earlier in the year by non-Cyanobacteria taxa, leading
to less phosphorus being available and thus a decline in annual mean *Anabaena* abundance
(despite their percentage abundance actually increasing). This study again emphasises the
complex coupling of climate-change driven responses to nutrient availability.

205 The above studies focussed on the interaction of nutrient load and water temperature, 206 but a study by Elliott (2010) used the PROTECH model to test the sensitivity of 207 Cyanobacteria to changing flushing rate and water temperature. Esthwaite Water (UK) was 208 the lake studied and a new response metric was used that recorded the number of days that 209 Cyanobacteria chlorophyll a concentrations exceeded thresholds defined by the World Health 210 Organisation (WHO; Chorus and Bartram, 1999). Annual mean percentage Cyanobacteria 211 abundance increased with higher temperatures and lower flushing rates (Fig. 3a), although the 212 present day level of dominance was very high (annual mean: 41%, annual max: 93%) 213 meaning the actual change was relatively small. However, the seasonal responses were 214 different: in the spring, mean percentage Cyanobacteria increased with temperature but 215 showed little response to changing flushing rate (Fig. 3b) whereas in the summer, the pattern 216 was similar to that seen in the annual means i.e. high percentage abundance with increased 217 temperatures and decreased flushing (Fig. 3c). However, in terms of absolute concentration, 218 as indicated by the number of days exceeding the WHO thresholds, the response was quite 219 different (Fig. 3d); low flushing rates increased the number of days above the threshold 220 whereas higher temperatures generally reduced the number. The mechanisms behind all 221 these responses were that the blooms were less prolonged and collapsed earlier due to the 222 increase in the community growth rate caused by the raised temperatures throughout the year. 223 Furthermore, under decreased flushing, nutrient load (i.e. of phosphorus, nitrogen and silica) 224 via the inflowing rivers was reduced leading to increasing reliance of internally released

phosphorus to support the summer and autumn growth, which, again, gave the nitrogen-fixing Cyanobacteria an advantage.

227	The final study in this review concerns PROTECH simulations of England's largest
228	lake, Windermere (Elliott, 2012). The lake consists of two interconnected basins (North and
229	South) and, using a present day simulation of both, the effect of changing air temperature and
230	nutrient load was examined. In both basins, the annual mean Cyanobacteria biomass
231	increased with temperature but the effect from nutrient load changes was more pronounced
232	and enhanced the temperature effect. This response was also echoed in the number of days
233	on which the WHO Cyanobacteria chlorophyll a threshold of 10 mg m ⁻³ was exceeded,
234	although there was a striking dependence on nutrients. F or example under the baseline
235	nutrient load, the increase in days averaged 2 days per 1°C increase, whereas under the +50%
236	phosphorus load scenarios the increase was 7 days per 1°C.

240 3. DISCUSSION

In the studies covered in this review, a range of scenarios were tested which allowed the
importance of different drivers to assessed. The key factors were changing water
temperature, stratification and nutrient loading. Therefore, the influence of these factors is
discussed below separately, drawing together the results of the different models and studies. *3.1 Water temperature*

246 Across most of the studies there was a general trend of enhanced Cyanobacteria biomass 247 and/or dominance with increasing water temperature, although, interestingly both of the 248 Swedish studies reviewed showed the least effect (Arhiemer et al., 2005; Markensten et al., 249 2010). This overall result fits the common speculation, advanced by studies of current observations (e.g. Paerl and Huisman, 2008), whereby it is assumed that Cyanobacteria 250 251 biomass will increase with a future warmer climate. However, just as has been observed in 252 studies of current climate change impacts on phytoplankton (e.g. Staehr & Sand-Jensen, 253 2006; Huber et al., 2008; Tadonléké, 2010), the strength of this response to a changing 254 climate appears to be greatly influenced by the nutrient resource base of the system i.e. the trophic status of the lake. 255

Despite the obvious close relationship between stratification and temperature, some studies had either controlled for the effect of stratification (e.g. Elliott et al., 2006 where the present day pattern of stratification was forced for the warmer scenarios), stratification did not change greatly (Elliott et al., 2005) or the model used assumed a continuously mixed water column (e.g. Mooij et al., 2007). These studies allowed the direct effects caused by the elevated water temperature to be tested and seemed to cause an alteration in the timing of Cyanobacteria growth (usually an advancement e.g. Elliott et al., 2005; Mooij et al., 2007)

and an increase in their dominance of the phytoplankton biomass (Elliott et al., 2006; Mooij
et al., 2007). The latter is of concern, because it shows that a lake under a future climate may
not necessarily be more productive but a greater proportion of the phytoplankton produced
could be Cyanobacteria, thus reducing water quality with little or no change in trophic status.

Interestingly, whilst the study using PCLake (Mooij et al., 2007) parameterized the 267 268 Cyanobacteria group in the model to have a stronger temperature dependency than the other 269 two simulated groups (diatoms and green algae), no such method was used for the 270 Cyanobacteria taxa modelled in the PROTECH simulations (Elliott et al., 2005; Elliott et al., 271 2006) where the growth rate of the taxa is dependent on its morphology. Subsequent testing 272 of PROTECH has shown that it is the movement characteristics and other abilities (nitrogen 273 fixation) of the Cyanobacteria taxa in the model that seems to give them their advantage 274 during the typical period of Cyanobacteria seasonal dominance (i.e. late summer) (Elliott et 275 al., 2010). This would suggest that the stratification pattern of the lake could be influential.

276 *3.2 Stratification*

277 Some of the modelling studies reviewed simulated lake stratification and examined the effect 278 the scenarios had on it. Stratification was not always affected by increased air temperature 279 (Elliott et al., 2005) but where it was, it generally led to an increase in the number of days 280 stratified and/or a stronger stratification (Markensten et al., 2010; Elliott, 2012). Markensten 281 et al. (2010) concluded that despite an increase in stratification duration, its impact on the 282 Cyanobacteria was small compared to catchment influences (e.g. nutrient load). In Elliott 283 (2012), the effect of changing stratification period in the autumn was to disrupt the general 284 relationship of increasing Cyanobacteria biomass with warmer surface temperatures, and was 285 related to reduced nutrient availability at the end of the phytoplankton growing season. Such 286 a strong relationship between stratification, nutrient availability and Cyanobacteria

abundance has been seen in other studies (Wagner and Adrian, 2009) and warrants greater
consideration in future modelling studies, especially given that there is evidence that
phytoplankton biomass in surface waters can enhance stratification (e.g. Jones et al., 2005;
Rinke et al., 2010).

291 3.3 Nutrient load

292 Most of the modelling studies that included a change in nutrients showed an enhancement 293 under the higher nutrient scenarios of the Cyanobacteria response to the climate drivers (e.g. 294 Fig. 2). This draws out the interesting point that in most lake systems, even eutrophic ones, 295 nutrients ultimately restrain the annual biomass of phytoplankton produced and that direct 296 effects of climate change on the lake are unlikely to change the annual carrying capacity. 297 However, the studies in this review (Arhiemer et al., 2005; Markensten et al., 2010) that 298 included catchment models, highlighted that climate change could affect the nutrient load to 299 the lake via the catchment, complicating the response of the phytoplankton. Therefore, the 300 importance of nutrient availability also shows that it is possible to try and alleviate climate-301 driven effects through reducing the nutrient load to the lake. Therefore, whilst demanding, 302 local solutions via nutrient load reduction to the lake are available to solve the added 303 complications that climate change could cause regarding Cyanobacteria.

304 3.4 Nitrogen fixation

This relationship between the climate-driven response and nutrients is further complicated by the influence of nitrogen-fixing Cyanobacteria, a property simulated in some of the models in this review (e.g. PROTECH, PROTBAS). This ability allows these Cyanobacteria to effectively circumvent nitrogen limitation, making the nutrient that is limiting growth important. The effects of this were particularly evident in the Loch Leven (Elliott and May, 2008) and Esthwaite Water (Elliott, 2010) studies. In the former, the warmer scenarios

311 produced less biomass due to increased nutrient consumption earlier in the year, but increased 312 the Cyanobacteria dominance of the phytoplankton because of the modelled ability of 313 Anabaena to utilise the phosphorus in the lake despite nitrogen concentrations being limiting. 314 The same mechanism was evident in the Esthwaite Water simulations, where the reduced 315 flow scenarios restricted nutrient supply to the lake and caused less nitrogen to be available 316 later in the year, leading to increased Cyanobacteria dominance. Therefore, both of these 317 examples show how increased water temperature can cause Cyanobacteria to experience an 318 indirect advantage though a general raising of growth rates earlier in the year, leading to 319 greater nutrient uptake and therefore an increased likelihood of nitrogen limitation later in the 320 year.

321 *3.5 Other consequences*

322 If climate change does increase the dominance of Cyanobacteria amongst the phytoplankton 323 of lakes, there is another potential impact to the whole food-web that was highlighted by 324 Mooij et al. (2007). As PCLAKE modelled the whole lake system, it showed that the 325 presence of large quantities of essentially inedible Cyanobacteria could reduce the amount of 326 energy that can flow up to the higher trophic levels. This would see negative and disruptive 327 impacts upon the zooplankton and fish populations within the lake community. Of course, as 328 Mooij et al. (2007) suggest themselves, this is an area of impact that warrants further 329 consideration by other studies and models before it is known how universal an effect it could 330 be, nevertheless, it is another result from these modelling studies that is a cause of concern 331 for lake ecosystem function.

332 3.6 The future for Cyanobacteria lake modelling

In writing this review, it was surprising how few published studies there were that lookedspecifically at the potential impact of climate change on lake Cyanobacteria populations.

335 One possible answer could be that that many modellers have a low level of confidence in the 336 ability of their lake model to capture the dynamics of these important phytoplankton. Most of 337 the models included in this review treated Cyanobacteria as a generic group whereas only 338 PROTECH and PROTBAS tried to model individual taxa of Cyanobacteria at a scale 339 analogous to the species level which would allow for successional changes within the group 340 to be explored. Furthermore, even these models did not try and model the detailed life cycle 341 of the Cyanobacteria that some models have attempted to capture (e.g. Hense and Beckmann, 342 2006). Given these issues, what would be the best approach to take the modelling of lake 343 Cyanobacteria forward?

344 Perhaps the first step would be to try and apply the models we already have, despite 345 our confidence in them. Obviously, models can be developed and further complicated almost 346 indefinitely in the search of perfection (or at least something close to it) but there should 347 come a time when they are used to investigate science questions and contribute to our 348 understanding of lake ecology. For example, PROTECH is a far from perfect model and 349 carries many simplifications (e.g. no Cyanobacteria life-cycle mechanics, assumes that 350 nitrogen-fixing taxa growth rates can never be limited by nitrogen availability) and yet it has 351 been used in five of the eleven studies presented here. Furthermore, despite these 352 simplifications, the general results from those studies are supported by the results produced 353 by the other models reviewed as well as the speculations derived from analysis of observed 354 data (e.g. Paerl and Huisman, 2008). This shows how models, regardless of their complexity, 355 can, and should be, used to help the lake phytoplankton community understand and predict 356 how climate change may impact upon these systems and particularly Cyanobacteria.

357

358 4. CONCLUSION

Despite the importance of knowing how Cyanobacteria may be influenced by climate change,
surprisingly few lake modelling studies have tackled the issue. However, from the few
studies that have, it seems clear that a number of important deductions can be drawn which,
whilst not totally conclusive, do have some merit worthy of further consideration.

Firstly, the direct effect of climate change via water temperature appears to affect the
 timing and proportional dominance of the Cyanobacteria, but not the amount of annual
 biomass of the phytoplankton community. Furthermore, the more nutrient rich the lake and
 greater the response of the Cyanobacteria populations modelled. There is also some evidence
 that climate change could increase this loading to lakes.

Secondly, due to the ability of some Cyanobacteria to utilise nitrogen-fixing, these
 phytoplankters can gain an advantage later in the growing season through nitrogen limitation
 caused by warmer waters in the spring increasing growth rates and nutrient consumption.

Finally, it is possible that an increase in Cyanobacteria dominance of the

372 phytoplankton biomass will lead to poorer energy flow to higher trophic levels due to their

373 relatively poor edibility for zooplankton.

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Table 1 Summary of the main climate drivers and their affect on Cyanobacteria in the studies reviewed. Note: RCM? Y = Driven by Regional 460

Climate Model, N = sensitivity method (see text for details). 461

Lake (country)	Trophic	Depth (m)	Volume	RCM?	Driver	Response
Model(s) used	status	(mean/max)	(10^6 m^3)			_
Farmoor Reservoir (UK) ¹	Eutrophic	9.2 / 11	4.5	Y	Reduced short-wave	None
CLAMM					radiation	
Bassenthwaite Lake $(UK)^2$	Eutrophic	5.3 / 19	27.9	Y	Higher temperature	No change in overall biomass, earlier
PROTECH						growth
Ringsjön (Sweden) ³	Eutrophic	5 / 17.5	184.2	Y	Higher temperature	Increase in overall biomass (via
PROBE & BIOLA						nutrients)
Galten basin of Lake	Eutrophic	3.4 / 19	210	Y	Higher temperature	Increase in dominance (via nutrients)
Mälaren (Sweden) ⁴						
PROTBAS						
Lake Rotoehu (New	Eutrophic	8.2 / 13.5	60	Y	Higher	Increase in dominance
Zealand) ⁵					temperature/nutrients	
DYRESM-CAEDYM						
Bassenthwaite Lake (UK) ⁶	Eutrophic	5.3 / 19	27.9	Ν	Higher temperature	Increase in dominance
PROTECH						
Generic shallow lake'	Varies	N/A	N/A	Ν	Higher temperature	Increase in dominance if nutrients
PCLake						high and/or lake turbid
Loch Leven $(UK)^8$	Eutrophic	3.9 / 25.5	52.4	Ν	Higher temperature	None
PROTECH						
Esthwaite Water (UK) ⁹	Eutrophic	6.4 / 15.5	6.4	Ν	Higher temperature	Increase in dominance
PROTECH					Lower flushing	
						Increase in dominance
Windermere (UK) ¹⁰	Mesotrophic	21.3 / 64	314.5	Ν	Higher temperature	Increase in dominance
PROTECH						

462

¹Howard and Easthope (2002), ²Elliott et al. (2005), ³Arheimer et al. (2005), ⁴Markensten et al. (2010), ⁵Trolle et al. (2011), ⁶Elliott et al. (2006), ⁷Mooij et al. (2007), ⁸Elliott

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465 Figure legends

466 Fig. 1 - Comparison of modelled Cyanobacteria chlorophyll *a* (fortnightly means) based on
467 present climate (solid line) and future climate (dotted line) in Bassenthwaite Lake (After
468 Elliott et al., 2005).

- 469 Fig. 2 The maximum annual percentage abundance of Cyanobacteria in the simulated
- 470 phytoplankton communities of Bassenthwaite Lake (After Elliott et al., 2006).
- 471 Fig. 3 Response of annual maximum percentage Cyanobacteria abundance in Esthwaite
- 472 Water to changing water temperature (°C) and flushing rate for (a) the whole year, (b) spring,
- 473 (c) summer and (d) number of days exceeding the lower WHO (World Health Organisation)
- 474 Cyanobacteria concentration threshold of > 10 chlorophyll *a* mg m⁻³ (After Elliott, 2010).









