1	Title: Expression	of Microcystis	biosynthetic gene	clusters in natural	populations suggests
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- 2 temporally dynamic synthesis of novel and known secondary metabolites in western Lake Erie
- 3 **Running Title:** Novel & known Microcystis biosynthetic genes
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24 Abstract

25 Microcystis spp. produces diverse secondary metabolites within freshwater cyanoHABs 26 around the world. In addition to the biosynthetic gene clusters (BGCs) encoding known 27 compounds, Microcystis genomes harbor numerous BGCs of unknown function, indicating a poorly understood chemical repertoire. While recent studies show that Microcystis produces 28 29 several metabolites in the lab and field, little work has focused on analyzing the abundance and 30 expression of its broader suite of BGCs during cyanoHAB events. Here, we use metagenomic 31 and metatranscriptomic approaches to track the relative abundance of Microcystis BGCs and 32 their transcripts throughout the 2014 western Lake Erie cyanoHAB. Results indicate the presence of several transcriptionally active BGCs that are predicted to synthesize both known and novel 33 secondary metabolites. The abundance and expression of these BGCs shifted throughout the 34 bloom, with transcript abundance levels correlating with temperature, nitrate, and phosphorus 35 concentrations, and the abundance of co-occurring predatory and competitive eukaryotic 36 37 microorganisms, suggesting the importance of both abiotic and biotic controls in regulating expression. This work highlights the need for understanding the chemical ecology and potential 38 risks to human and environmental health posed by secondary metabolites that are produced but 39 40 often unmonitored. It also indicates the prospects for identifying pharmaceutical-like molecules from cyanoHAB-derived BGCs. 41

42 Importance

Microcystis spp. dominate cyanobacterial harmful algal blooms (cyanoHABs) worldwide
 and pose significant threats to water quality through the production of secondary metabolites,
 many of which are toxic. While the toxicity and biochemistry of microcystins and several other
 compounds have been studied, the broader suite of secondary metabolites produced by

Microcystis remains poorly understood, leaving gaps in our understanding of their impacts on 47 human and ecosystem health. We use community DNA and RNA sequences to track the 48 diversity of genes encoding synthesis of secondary metabolites in natural Microcystis 49 populations and assess patterns of transcription in western Lake Erie cyanoHABs. Our results 50 reveal the presence of both known gene clusters that encode toxic secondary metabolites as well 51 52 as novel ones that may encode cryptic compounds. This research highlights the need for targeted studies of the secondary metabolite diversity in western Lake Erie, a vital freshwater source to 53 the United States and Canada. 54

55 Introduction

CyanoHABs, which are dense proliferations of cyanobacteria that can discolor water and produce toxins, occur annually and globally, and are expected to increase in severity with climate change (1–3). These blooms, which can persist through the early summer into late fall in temperate environments, produce a range of secondary metabolites that can be deleterious to ecosystem function and human health (4–6). CyanoHABs result largely from anthropogenic eutrophication via nutrient runoff (2, 7, 8) and are expected to become more toxic with increasing nutrient loading and continued, intensifying, climate change (2, 9, 10).

Microcystis is a non-N-fixing, potentially toxic cyanobacterium that often dominates freshwater cyanoHABs on every continent except Antarctica (5). Blooms made up of largely *Microcystis* are expected to expand and increase in severity over the next several years (2, 6). This expansion is a concern because *Microcystis* produces diverse bioactive secondary metabolites with an array of ecological and physiological functionalities (11–13). Many *Microcystis* secondary metabolites possess toxic or inhibitory properties toward diverse cell types (13–15) and/or antibiotic, antifungal, or cytotoxic properties that may be useful in

pharmaceutical discovery (16–18). Secondary metabolites likely provide *Microcystis* with
ecological advantages such as grazer defense, allelopathy, quorum sensing, and/or protection
against reactive oxygen species (ROS) and may influence cyanoHAB community composition
(17–20). Despite their significance to ecology, human, and ecosystem health, the physiological
and environmental controls on biosynthesis of *Microcystis* secondary metabolites remain poorly
understood.

76 Microcystins, the most extensively studied family of secondary metabolites produced by *Microcystis* (21–23), are structurally related hepatotoxins that are responsible for several 77 78 drinking crises (4, 24, 25) and livestock poisoning events (26) around the world. However, Microcystis genomes also contain additional BGCs (18, 20, 27) encoding secondary metabolites 79 such as aeruginosins, anabaenopeptins, cyanobactins, cyanopeptolins, microginins, and 80 microviridins (11, 13), which can be toxic and may occur in the environment and drinking water 81 treatment plants at frequencies and concentrations that equal or exceed those of microcystins 82 83 (28-30). In addition, many BGCs identified from *Microcystis* genomes do not have associated products, and thus have been designated "orphan" clusters (12, 31), highlighting the need to 84 85 identify and determine the structural characteristics and biological properties of these unknown 86 compounds. While *Microcystis* genomes harbor a diverse and variable suite of BGCs (19, 20) and some metabolites have been shown to be expressed and produced in culture (18), little is 87 88 known about the occurrence or expression of these *Microcystis* BGCs in natural environments. 89 Understanding their distribution and expression in relation to the biotic and abiotic environment 90 may shed light on the ecological role of these molecules and inform risks to public health.

In this study we used metagenomics and metatranscriptomics to assess the abundance,
diversity, and expression of BGCs in *Microcystis* populations in a time series at three stations

during the 2014 western Lake Erie bloom, which show a succession of Microcystis strains 93 alongside changing nutrient availability, microcystin concentrations, and microbial communities 94 (32, 33). Western Lake Erie is subject to annual cyanoHABs that have intensified in recent 95 decades (6, 34), and the 2014 bloom caused a drinking water crisis in which Toledo residents lost 96 access to potable water due to dangerously high levels of microcystins (4). While microcystins 97 98 are heavily monitored and studied in these waters, monitoring of "other" secondary metabolites has been limited to saxitoxin, anatoxin, and cylindrospermopsin (35). Motivated by the known 99 diversity of Microcystis BGCs and the influence of nutrient stoichiometry on biosynthesis of 100 101 secondary metabolites (36, 37), we hypothesized that the relative abundance and transcription of BGCs would vary across the bloom season as nutrient availability changes. 102

103 **Results**

104 The 2014 Western Lake Erie cyanoHAB

The 2014 western Lake Erie cyanoHAB was notable due to high levels of microcystins at 105 station WE12, the Toledo Drinking water crib, which led to a drinking water crisis event in 106 Toledo, OH, USA. This study used samples and data collected and described previously (32, 33, 107 38, 39). Briefly, samples were collected from three core stations as part of greater sampling 108 efforts performed by the NOAA Great Lakes Environmental Research Laboratory (GLERL) to 109 monitor cyanoHAB development in the western basin of Lake Erie. WE2 and WE12 are close to 110 the Ohio coast and Maumee River inlet and are considered nearshore stations. WE4 is located 111 more centrally in the basin and is considered offshore (Fig 1A). 112 The 2014 bloom consisted of high levels of nitrate, particulate microcystin, and 113

114 cyanobacteria biomass in August at the nearshore stations WE2 and WE12. A secondary peak in

115 cyanobacteria biomass, where microcystin concentrations were low, was observed in late September at WE12 and WE2. The trends at the nearshore stations suggest that high nitrate 116 concentrations may support initial bloom development, proliferation, and/or production of 117 microcystin, but that biomass and microcystin production decreases as nitrate is being depleted. 118 At the offshore station, WE4, microcystin concentrations were considerably lower (Fig 1B). 119 Soluble reactive phosphorus (SRP) and nitrate concentrations (NO₃⁻) were also variable across 120 stations and sampling times, with marked decreases in nitrate as the bloom persisted at stations 121 WE2 and WE12 from 4 August, onwards. Notably, at all three sampling stations, the lowest 122 123 measurement of nitrate occurred one sampling point before the lowest measurement of particulate microcystin (Fig 1B). 124

125 Microcystis MAG Generation

Ten Microcystis metagenome-assembled genomes (MAGs) were generated from fifteen 126 metagenomic samples from the 2014 western Lake Erie cyanoHAB (Table 1). Samples that 127 failed to generate a Microcystis MAG were likely due to low abundance of the organism as seen 128 in early phases of the bloom (early July), or poor assembly of *Microcystis* contiguous sequences 129 due to high strain heterogeneity (40). Most MAGs had high completion (above 90%) and 130 variable redundancy given the high strain heterogeneity observed (41), except for the MAG 131 generated from August 4th, at Station 12, at 40% completion (Table 1). This MAG was kept for 132 further analysis as this sample was taken directly during the time of the Toledo Drinking Water 133 Crisis. Five fragmented BGCs (<5kb) identified in this MAG were removed from further 134 analysis due to incompletion, while one complete BGC (T1PKS) was identified and used for 135 136 subsequent analysis. Fragmented BGCs were determined to be incomplete due to the fragmentation or contiguous sequence breaks present within gene open reading frames. 137

138 BGC Diversity and Abundance

139	Nineteen distinct BGCs, representing multiple classes of putative secondary metabolite
140	synthesis, were identified (Table 2, Fig. 2). Of the nineteen BGCs, eleven clusters are thought to
141	putatively encode known synthesis products, while eight are considered cryptic (Table 2). Within
142	the cryptic clusters, PKS-M 1 and 2 appear to be fragmented or incomplete microginin encoding
143	BGCs, four have been identified in previous work as "orphan clusters" (31), and two are novel,
144	and to our knowledge, have not been previously discussed in the literature (Table 2).
145	The relative abundance of these BGCs, which was estimated by quantifying metagenomic
146	reads mapped to assembled BGCs and normalized to reads mapped to Microcystis 16S rRNA
147	genes, showed differences across stations and sampling dates (Fig 2). The relative abundance of
148	the complete mcy operon, which was previously analyzed (33) is included for reference as well.
149	Relative abundances of identified BGCs, which putatively produce aeruginosins,
150	anabaenopeptin, cyclophane-like 1, cyanopeptolin 3, and microviridin B, and a cyclophane-like
151	molecule (PKSmod-NRSlike-T3PKS), increased at station WE12 in late September through
152	October, during later phases of the bloom. BGCs predicted to putatively encode aeruginosins,
153	cyanobactins (58-75% similarity to piricyclamides encoding clusters), microviridins, and
154	cyanopeptolins each yielded multiple distinct clusters with conserved gene content, but gene
155	rearrangements, insertions, and duplications were observed within these clusters (Fig. S1). Gene
156	order and orientation of BGCs encoding aeruginosin and microviridin were more conserved (Fig
157	S1A & S1C), while rearrangements, insertions, and deletions were more common in the
158	cyanobactin and cyanopeptolin BGCs (Fig S1B & S1D). Deeper annotations for clusters thought
159	to encode anabaenopeptin, microviridin, aeruginosin, and a piricyclamide-like cyanobactin as
160	well as chemical structures of potential congeners produced are shown in Fig. S2. Some clusters,

such as the PKSmod-NRPSlike-T3PKS 1, which may encode a cyclophane-like metabolite, were
rare at most sampling times and stations, except on 20 October when it was one of the most
abundant clusters. The cyanobactin clusters, which encode ribosomally synthesized
cyanobacterial derived macrocyclic metabolites (42), shifted in relative abundance during the
bloom with 2 and 3 being more abundant early, and 1 being the most abundant during late phases
(Fig. 2).

Two gene clusters, PKS-M 1 and 2, are believed to be part of the microginin synthesis 167 pathway, which was recently confirmed to be present in *Microcystis* (43); however, both clusters 168 contain a fragmented hybrid NRPS-PKS encoding gene and lack the additional two NRPS 169 encoding genes required for microginin synthesis (Fig 3A). When compared to the microginin 170 pathway in Microcystis aeruginosa LEGE 91341 (43), PKS-M 1 is shown to contain additional 171 synthesis genes thought to encode a fatty acyl-AMP ligase and acyl carrier protein, as well as a 172 hypothetical protein. PKS-M 2, however, lacks these putatively identified genes, and contains a 173 174 putative endonuclease (Fig. 3A). Both core biosynthesis genes present in both PKS-M clusters contain domains for ketosynthase, acyltransferase, PP-binding, and aminotransferase activity, but 175 lack the condensation, AMP binding, and peptidyl-carrier protein domains observed in the 176 177 complete clusters (Fig. 3A). To determine if these incomplete clusters were a result of natural pathway truncations or poor metagenomic assembly, mapping of western Lake Erie (WLE) 178 metagenomic reads against the reference microginin BGC from LEGE 91341 was completed. 179 180 Mapping revealed read coverage across the entire length of the BGC, suggesting the presence of this complete BGC in natural populations, although there are some notable variations in 181 sequence similarity and regions of non-uniform coverage (Fig 3B). These gene clusters were 182 most abundant at nearshore stations in later phases of the bloom (Fig. 2). 183

184 Three cryptic polyketide synthase (PKS) or hybrid nonribosomal peptide synthetase (NRPS)-PKS cluster classes were also identified and varied greatly in abundance (Fig. 2,4). 185 These systems putatively encode unknown metabolites and have either very low percent 186 similarities ($\leq 40\%$; Table S1) or no similarity at all to previously described clusters in the 187 MiBIG database, suggesting high potential for biosynthesis of unique structures. The cryptic type 188 189 III polyketide synthase (T3PKS) identified reached peak relative abundance at the nearshore stations on 4 August (Fig. 2) and contains a putative naringenin-chalone synthase (Fig 4A). The 190 cluster encoding a predicted iterative PKS was greatest in relative abundance on 20 October at 191 192 nearshore stations and contains a PKS encoding a putative enediyne biosynthesis system (Fig. 2, 4A). MIC 1 was previously described in *Microcystis aeruginosa* NIES-843 (31), and putatively 193 encodes two modular T1PKSs, an NRPS like enzyme, an additional PKS, and a halogenase (Fig. 194 4C). Deeper annotations for genes identified in the MIC 1 cluster are described in Table S2. 195

196 Protein Phylogenetics for Biosynthesis Genes

Phylogenetic trees of various predicted proteins from PKS-containing clusters were analyzed 197 for potential insights into the corresponding biosynthetic pathways. The enediyne biosynthesis 198 protein sequence (PKSE) from the PKS iterative cluster, which was initially identified in 199 Microcystis aeruginosa strain PCC 7806 (31), was investigated more deeply as enediyne 200 synthesis is a critical step in the formation of anticancer compounds such as calicheamicin (44). 201 While there were no closely related homologous sequences from known biosynthesis pathways, 202 the enediyne protein sequence had over 97% identity and 100% alignment to a putatively 203 identified PKSE in Microcystis aeruginosa strain NIES-843 (45), with a bootstrap of 1 (Fig 5). 204 205 Other putative PKSE enzymes identified from a variety of cyanobacteria (45), formed a subclade with the PKSE identified in W. Lake Erie MAGs (Fig 5). These protein sequences form a larger 206

207	clade with known PKSE enzyme sequences that generate a variety of enediyne-containing
208	metabolites with pharmaceutical relevance (bootstrap=0.81, Fig. 5).

209 Protein sequences from the T3PKS gene cluster (Fig. 3B), also provided clues into potential 210 functions of biosynthesis enzymes. An exporter protein sequence shared high sequence similarity to proteins involved in siderophore-like export from other *Microcystis* strains (~95-97% identity, 211 212 bootstrap = 1, Fig S3A). Distantly related clades contained protein sequences with annotations 213 for export of siderophores and/or cyclic peptides from a variety of cyanobacteria including benthic *Leptolyngbya* sp. (bootstrap = 0.894, ~63% identity, Fig S3A). The core synthesis 214 215 enzyme within this cluster, annotated by antiSMASH as naringenin-chalcone synthase, was most closely related to other Microcystis naringenin-chalcone synthase and T3PKS enzymes (~98% 216 217 identity, bootstrap = 0.877, Fig. S3B) as well as naringenin-chalcone synthases from *Chamaesiphon* sp. from the cyanobacteria Order *Synechococcales* (bootstrap = 0.688, 57-62% 218 identity, Fig. S3B), none of which are linked to a known synthesized compound. Its protein 219 220 sequence was also similar to an experimentally confirmed α -pyrone synthesis polyketide 221 synthase (PKS18) (Fig. S3B) from the actinobacterium Nocardia seriolae (50.6% identity, 97% alignment length). 222

223 Transcriptional activity of BGCs

Quantification of relative abundances of metatranscriptomic sequences, which were available for seven of the 15 samples analyzed, revealed that BGC clusters were differentially expressed across stations and sampling times (Fig. 6). During early phases of the bloom, there was little expression of BGCs as observed on 21 July at WE2 and 29 July at WE4, likely due to low biomass of *Microcystis*. At WE12 during the peak bloom phase (4 August), clusters T3PKS, PKS-M 1, and *mcn* 2 increased in relative transcript abundance compared to other clusters (Fig.

230	6). While the relative abundance of transcripts for <i>mcy</i> genes encoding microcystin (33) was
231	highest at WE12 during August, other BGCs had similar, and in some cases greater, relative
232	transcript abundances than the microcystin encoding gene cluster during other phases of the
233	bloom (Fig. 6). The relative abundance of BGC transcripts shifted and greatly increased,
234	especially in comparison to the mcy operon, at all three stations during middle and late phases of
235	the bloom, with high relative abundance of transcripts for T3PKS and MIC1 at WE12 on 25
236	August, WE2 on 6 October, and 8 September at WE4. On 23 September at WE12, the PKSmod-
237	NRPS-T3PKS 1 gene cluster had the highest relative transcript abundance compared to other
238	gene clusters. There was also a marked increase in relative transcript abundance of mcn 2 during
239	later phases of the bloom at both WE2 and WE4. Greater transcriptional abundance at WE2 was
240	also observed for clusters PKSmod-NRPSlike-T3PKS 1 and aer 2 on 6 October as well. (Fig. 6).
241	Correlation Analysis between transcript abundance and both abiotic and biotic variables
242	To explore potential controls on the expression of BGCs, we performed correlational
243	analysis of BGC transcript abundance with a variety of available abiotic conditions and relative
244	abundances of known Microcystis predators and competitors in metagenomic data. Relative
245	abundances of BGC transcripts were significantly correlated with both abiotic conditions (Table
246	3) and relative abundance of predatory and competitive organisms (Table 4). The transcript
247	abundance of all BGC classes examined was significantly and negatively correlated with nitrate
248	concentration while the transcript abundances of genes for cyanobactins (Pearson's R=-0.531,
249	p=0.029), PKSmod-NRPSlike-T3PKS (Pearson's R=0.598, p=0.040), and mcn (cyanopeptolin)
250	(Pearson's R=-0.412, p=0.089) BGCs were negatively correlated to temperature. Transcript
251	abundances for aer, mcn, mdn, and PKSmod-NRPSlike-T3PKS clusters had positive and

significant or near significant correlations with both soluble reactive phosphorus (SRP) and total
phosphorus (TP) concentrations (Table 3).

254	The relative metagenomic abundance of ciliates and diatoms exhibited some of the
255	strongest positive correlations with relative transcript abundance of BGCs, including those
256	predicted for <i>aer</i> , <i>mdn</i> , and PKSmod-NRPSlike-T3PKS clusters (Pearson's R=0.653, p=0.041 to
257	Pearson's R=0.517, p=0.003, Table 4). Cyanobactin BGC expression was significantly and
258	positively correlated with the relative abundance of the diatom Skeletonema (Pearson's R=
259	0.617, p=0.033). Cyanopeptolin encoding BGC expression was also significantly and positively
260	correlated with the abundance of <i>Skeletonema</i> (Pearson's R= 0.741, p=0.006) as well as the
261	betaproteobacterium, <i>Paraburkholderi</i> a (Pearson's $R= 0.488$, $p= 0.025$). All other biotic
262	correlations are summarized in Table S4. Taken together, these results suggest that both abiotic
263	and biotic variables may be important in stimulating biosynthesis in natural cyanoHAB
264	Microcystis populations.

265 **Discussion**

Microcystis blooms are renowned for their production of microcystins, toxins that 266 threaten drinking water supplies (46-48). However, *Microcystis* genomes contain many other 267 268 BGCs that encode known and unknown products, and their genetic diversity, biosynthesis, and ecological functions in the environment are poorly understood. In this study, we tracked spatial 269 and temporal shifts in BGC abundance and transcriptional activity through seasonal changes in 270 the biotic and abiotic environment of the 2014 western Lake Erie bloom, providing insights into 271 the biosynthetic potential and expression of diverse Microcystis populations in their natural 272 habitat. Our results demonstrate that Microcystis contains a diverse and highly dynamic suite of 273 BGCs that putatively encode both known cyanotoxins and cryptic compounds and are 274

transcriptionally active in natural blooms. The transcriptional activity of these putative BGCs is
related to both abiotic conditions (nutrients, temperature) as well as the composition of the
biological community. While previous studies have identified BGCs in *Microcystis* isolates (20,
27, 31) or described metabolomic profiles in both field and culture studies (12, 30), our work
focused on the relative abundance, diversity, and drivers of transcriptional activity of BGCs
within natural populations as they occur *in situ*.

281 *Microcystis* MAGs displayed extensive genetic diversity at the sub-species level, which is 282 increasingly recognized to be prevalent in natural blooms (32, 33, 38), and encodes important 283 biosynthetic and physiological diversity (4, 20, 33). Although these MAGs contained some redundancy and short contiguous sequences due to high strain heterogeneity present in natural 284 populations (19, 33), they were sufficient to track spatial and temporal shifts in *Microcystis* gene 285 content. Our approach to estimating relative abundance of BGCs also had limitations. It was 286 based on normalization of BGC abundance to Microcystis 16S rRNA abundance. In some 287 instances, the estimated relative abundance of some BGCs was greater than 1, implying multiple 288 copies of a BGC per cell. We discussed potential explanations for this observation previously 289 (33), including uncertainty of 16S rRNA gene and BGC gene copies per Microcystis genome 290 291 (49, 50) in natural populations, presence of multiple copies of BGCs per cell due to replication of genes near the origin of replication during rapid growth (51–53), or non-specific mapping of 16S 292 rRNA and/or BGC genes. In addition, BGCs may be present on plasmids (54), highlighting the 293 294 potential for multiple copies of these genes per genome. Our previous analysis showed that nonspecific mapping of metagenomic sequence reads to *Microcystis* 16S rRNA and BGC genes was 295 not a substantial issue under the stringent thresholds that we used here (33), but the current data 296 is not sufficient to assess other potential issues. Regardless of these caveats, the results of this 297

study do not depend on absolute calculations but rather relative differences, highlightingdynamics in time and space.

300 The diversity of BGCs we observed, and their spatiotemporal shift in abundance, reflects the 301 presence and shifting abundance of diverse Microcystis strains with varying BGC content in their genomes (19, 20, 55). Whereas we identified and deeply analyzed nineteen BGCs, Microcystis 302 303 typically has only 5-7 NRPS, PKS, or hybrid NRPS/PKS BGCs per genome (20, 56), and BGC 304 content varies across clades (20). Western Lake Erie Microcystis populations contain multiple strains in varying abundance (33, 38, 57), often with microcystin producing strains dominating in 305 306 early and peak bloom phases and non-microcystin producing strains dominating later phases of blooms (32, 33). Our results show that the "non-toxic" phases of the bloom, so-called because of 307 low concentrations of microcystins, can be enriched in other BGCs (Fig. 2, Table 2) that are 308 309 transcriptionally active, including those that putatively encode anabaenopeptin, aeruginosin, 310 aeruginoguanidine, cyanopeptolin, and cyanobactins (Fig. 6). Even more striking is the pervasive 311 abundance and transcriptional activity of cryptic, putative, BGCs containing PKS modules with 312 unknown products (Figs. 2,4,6). BGCs with the highest transcript abundance included those encoding unknown compounds such as T3PKS, PKSmod-NRPSlike-T3PKS 1 (thought to 313 314 putatively synthesize a cyclophane-like product), and MIC 1 (Fig. 6), indicating that they may be functionally and ecologically important. The lack of functional knowledge of these abundant and 315 316 transcriptionally active PKS containing clusters highlights the need to investigate further their 317 potential for generating unique metabolites.

The apparent "replacement" of *mcy* genes by other BGCs that we observed from early to late blooms in the field is consistent with studies of cultured *Microcystis* strains showing that nonmicrocystin producing strains often contain other BGCs in their place (20). Our finding of a

321 strong temporal shift in *Microcystis* BGC content and expression suggests environmental drivers of BGC content and expression. Strains containing other BGCs instead of those encoding 322 nitrogen-rich microcystins may be more fit during nitrogen deplete phases of the bloom that 323 occur later in the season (Fig. 1B). For example, cyclophane-like molecules, which may be 324 encoded by cryptic PKSmod-NRPSlike-T3PKS clusters, are nitrogen poor, in many cases 325 326 lacking N atoms altogether (Table 3, (58)). Strains with BGCs that putatively encode such molecules may have a competitive advantage over microcystin producers with higher N demands 327 (36, 37, 59). However, inconsistent with this conclusion is the increase in abundance of BGCs 328 329 putatively encoding other N-rich molecules, such as aeruginosin (Table 3), while nitrate concentrations decrease. These results suggest that the different secondary metabolites may have 330 different fitness tradeoffs and/or environmental/physiological controls on gene expression. The 331 high potential for biosynthesis of other secondary metabolites when microcystins are at low 332 concentrations suggests functionally redundancy or niche differentiation among secondary 333 metabolites. 334

BGCs that were more highly expressed in later bloom phases included *apn* (anabaenopeptin), 335 *aer* (aeruginosin), *mcn* (cyanopeptolin), and cyanobactin classes, which have inhibitory and toxic 336 337 effects towards eukaryotes (13, 60–62). The T1PKS-M BGCs likely represent poorly assembled microginin synthesis pathways, which, when synthesized, have inhibitory properties against 338 angiotensin-converting enzymes (63, 64). Other transcriptionally active BGCs encode unknown 339 340 compounds with unknown toxicology. Given the continued emergence and discovery of novel cyanotoxins and their underpinning BGCs (65, 66), and the potential for toxicological synergism 341 between multiple compounds (67, 68), these results underscore the need to assess potential 342 threats to human and ecosystem health. 343

In addition to potential risks of unmonitored, toxic, cyanobacteria secondary metabolites, 344 these molecules may be an untapped source for drug discovery (69). For example, the T3PKS 345 BGC was one of the most abundant and transcriptionally active PKSs during early phases of the 346 bloom (Fig. 2,6) and contained a putative naringenin-chalcone synthase encoding gene (Fig. 4), 347 which may have the potential to generate antibiotics or other compounds with wide ranges of 348 biological activity and pharmaceutical relevance (70, 71). The cluster bearing an iterative PKS 349 (Fig 4A, 5) is predicted to encode biosynthesis of an enediyne-containing molecule, which are 350 excellent candidate antibiotics and anticancer compounds with rare structural characteristics (72, 351 352 73). Previous work has shown that enediyne gene clusters usually contain a set of five conserved biosynthetic genes included the enediyne PKS, a thioesterase, and three other genes that encode 353 proteins of unknown function (73). The iterative PKS cluster contains a putative enediyne PKS, 354 thioesterase, and two of the three genes of unknown function that are part of the enediyne 355 biosynthesis cassette (Fig 4A). This, taken together with the presence of a putative enediyne 356 cluster identified in Microcystis strain NIES-843 (45), which shares 97% identity with our 357 cluster, supports the conclusion that Microcystis are capable of synthesizing uncharacterized 358 enediyne metabolites. Finally, the presence of a halogenase encoding gene within the MIC1 359 360 cluster (Fig. 3D) is noteworthy due to the broad pharmaceutical application of halogenated compounds, which have largely been studied in marine organisms (74). 361

Correlation of transcript abundance with abiotic and biotic factors provided insights into environmental conditions that may influence expression of BGCs in natural environments. Negative correlations between relative abundance of transcripts and nitrate and temperature suggest a functional need for synthesized compounds in conditions of lower nitrate concentrations and temperature (Table 3). Positive correlations of BGC transcript abundance

367 with abundances of known eukaryotic predators (Table 4) are consistent with the hypothesis that BGC products may deter grazers (ciliates) that commonly feed on cyanobacteria in aquatic 368 systems (75, 76). Of the cyanobacteria commonly found in western Lake Erie cyanoHABs, 369 *Microcystis* are the most resistant to grazing by both daphnid and protozoan microorganisms (77) 370 consistent with findings that *Microcystis* is resistant to grazing due to production of secondary 371 metabolites (78-80). Likewise, significant correlations between expression of BGCs and 372 abundance of photosynthetic competitors (diatoms) is consistent with a role of secondary 373 metabolites in allelopathic interactions (81, 82). Overall, these results are also consistent with 374 375 recent findings that community composition and interspecies interactions influence expression of BGCs and secondary metabolites affect interspecies interaction networks (83). Microcystis 376 secondary metabolites are likely multi-functional and may serve as nitrogen storage compounds 377 (84, 85), aid in grazer defense (14, 80, 86), protection from reactive oxygen species (87, 88), 378 and/or cell communication/phycosphere recruitment (89, 90). While correlations uncovered here 379 cannot determine true drivers of BGC abundance or expression, they provide hypotheses that can 380 be tested with targeted experiments on the ecological and cellular role of the diversity of 381 secondary metabolites encoded in *Microcystis* genomes. 382

This study shows that natural populations of *Microcystis* within blooms contain and express a diverse and dynamic suite of genes encoding secondary metabolites. Several of these BGCs have no known products and are expressed most highly during the late phases of the bloom, which is dominated by "nontoxigenic" *Microcystis* strains that are incapable of producing microcystins. These results suggest the potential for diverse metabolite production beyond microcystins in western Lake Erie, with implications for understanding *Microcystis* physiology and diversity, community ecology, and water quality. Seasonal shifts in the abundance and

expression of BGCs along with correlations with both abiotic and biotic variables suggest that 390 the BGCs underpin adaptations to changing environmental conditions and may encode important 391 niche differentiation among Microcystis strains. Several clusters identified may encode toxic 392 compounds while others putatively encode cryptic compounds that may be toxic, highlighting 393 the need to identify their products, determine their bioactivities, and assess potential threats to 394 395 human and ecosystem health. Others show the potential to produce metabolites with interesting chemical features that suggest avenues for exploring biotechnological and medicinal 396 applications. Finally, the abundant environmental expression of BGCs encoding secondary 397 398 metabolites that are of unknown function, and their correlation with biotic and abiotic conditions, suggest that the BGCs may play important roles in supporting the dominance of *Microcystis* over 399 a wide range of conditions, underscoring the need to understand the functioning of these putative 400 secondary metabolites in the ecophysiology and community interactions of *Microcystis* blooms. 401

402 Materials and Methods

403 Study Site and Sample Collection

Samples were collected weekly from various NOAA Great Lakes Environmental
Research Laboratory (GLERL) sampling stations throughout the western Basin of Lake Erie
from mid-June through late October 2014 (91). The long-term NOAA GLERL stations selected
for sampling were WE2, WE4, and WE12. WE2 is close to the inlet for the Maumee River inlet
(41° 45.743'N, 83° 19.874' W), WE4 is considered an offshore site closer to the center of the
basin (41° 49.595'N, 83° 11.698'W), and WE12 is proximal to the Toledo drinking water inlet
(41° 42.535'N, 83° 14.989'W) (Fig. 1A).

411	Twenty liters of depth integrated samples were collected from each station for biological
412	and chemical analysis. Integrated depths were defined as the surface of the water to 1m above
413	the lake floor. While samples were collected, measurements of pH and water temperature were
414	performed as well. To capture Microcystis aggregates, 2L of integrated depth collected water
415	were filtered through a 100 μ m polycarbonate mesh filter. The biomass on the filter was then
416	collected and filtered through a 0.22 μm filter. The filter-bound biomass was preserved in 1 mL
417	of RNALater TM (Invitrogen TM , Ambion TM) and placed on ice. Upon arrival to the lab, samples
418	were stored in the -80°C freezer until DNA and RNA extractions could be completed.
419	Sample Processing and Sequencing
420	Processing and sequencing of these samples was described previously (33). Briefly, DNA
421	was extracted with the Qiagen DNeasy Blood and Tissue Extraction Kit (Qiagen, Hilden,
422	Germany) and quantified with the Quant-iT [™] PicoGreen [™] dsDNA Assay Kit (Eugene, OR,
423	USA). The Qiagen RNeasy Kit (Qiagen, Hilden, Germany) was used to complete RNA
424	extraction. Shotgun DNA and RNA sequencing was conducted at the University of Michigan
425	DNA Sequencing Core on the Illumina® HiSeq [™] platform (2000 PE 100, Illumina, Inc., San
426	Diego, CA, USA).

427 Bioinformatic Analysis

Shotgun reads were assembled *de novo* to recover *Microcystis* Metagenome Assembled
Genomes (MAGs). Metagenomic raw read sequences obtained from the University of Michigan
Omics Core were analyzed and processed through the following metagenomic workflow. Raw
reads for metagenomic analysis can be found at NCBI BioProject Accession: PRJNA464361.

432 The IMG-JGI supported software, BBTools was used to complete trimming, adaptor removal, and quality check of raw reads. Quality checked (QC) reads were normalized to 100x 433 coverage using bbnorm from the BBTools package (92). 100x coverage was chosen based on the 434 assembly quality of the following housekeeping genes: a global nitrogen regulator *ntcA*, the beta 435 subunit phycocyanin gene *cpcB*, the photosystem I P700 chlorophyll apoprotein *psaB*, the 436 437 photosystem II P680 reaction center D1 protein *psbA2*, as well as the assembly quality of the mcy operon. Assembly quality was assessed based on contiguous sequence length, and De Bruijn 438 Graph assembly visualized in Bandage (93). 439

De novo assemblies were generated for each sample using MEGAHIT (94). To obtain 440 differential coverage for binning software, mapping of QC reads was completed using Bowtie2 441 (95). A multiple binning approach was used in which bins generated from Concoct, Metabat, 442 Tetra ESOM, and Vizbin were assessed and chosen with DASTool (96–100). To improve 443 accuracy of binning software, differential coverage of all 15 samples was used by mapping reads 444 from each sample to each assembly using Bowtie2. The minimum cut off for contiguous 445 sequence length was 1 kb for Concoct, 1.5 kb for Metabat, and 3.5 kb for Tetra ESOM and 446 Vizbin. The cut off for Metabat was selected based on minimum cut off length the program 447 448 required and that for ESOM and Vizbin was selected based on clarity of maps. *Microcystis* MAGs were assessed and manually refined using Anvi'o v. 5, "Margaret" (101). Completion and 449 strain heterogeneity of each MAG was assessed using CheckM with default parameters (102) as 450 451 well as metrics generated through Anvi'o outputs. MAGs were mined for BGCs with antiSMASH 6.0 using a full Hidden Markov Modelled search, orthologous gene prediction, gene 452 border prediction, PFAM analysis, and COG identification (42). For further BGC annotation, 453

454 identified BGC sequences from all *Microcystis* MAGs found by antiSMASH were run through
455 PRISM (103) software as well.

456 For downstream analysis, BGC sequences identified by antiSMASH from all Microcystis 457 MAGs were pooled together and deduplicated using the dedupe tool from the BBTools package (92). Dedupe was set to deduplicate sequences that were at least 95% identical and had no more 458 459 than 10 substitutions or deletions. This deduplicated set of BGC clusters was used for subsequent analysis with special attention to specific clusters discussed in greater detail. BGC sequences 460 461 longer than 10,000 bp were saved for future gene and expression quantification calculations. In addition to these BGCs we also closely examined three identified cyanobactin BGC clusters that 462 were \sim 7 kb in length. To compare the similarities between clusters of the same class or 463 identification, the clinker software (104) was used to generate gene maps to visually inspect for 464 similarities and differences in gene cluster structure. In addition to using annotations provided by 465 466 antiSMASH, we aligned our BGCs to a previously described gene clusters found in *Microcystis* 467 (31, 43, 105) to assess the presence of BGCs known to be present in *Microcystis* genomes but are not deposited in the MiBIG database. BGC identified from w. Lake Erie that shared $\geq 95\%$ 468 identity to previously described clusters they were annotated as such. 469

Two of the clusters identified from w. Lake Erie MAGs, PKS-M 1 &2, were shown to have 57% similarity to the recently characterized microginin biosynthesis pathway from *Microcystis aeruginosa* LEGE 91341 (43), but were incomplete as they lacked two core NRPS encoding genes, and the hybrid PKS-NRPS encoding gene was shown to be fragmented. To assess if these patterns reflect truncated microginin pathways observed in natural populations, or are a result of poor assembly, metagenomic reads were mapped to the microginin biosynthesis pathway found in LEGE 91341 to assess the completion and presence of this cluster. Read mapping and

477 coverage plots were then used to visualize and assess the completeness of this pathway and478 microginin producing potential.

479 BGC Cluster Quantification

To estimate the relative abundance of BGC clusters the following steps were completed. 480 QC metagenomic reads were aligned to all identified BGC clusters found in any cyanobacterial 481 MAG (including *Microcystis*, *Anabaena*, *Pseudanabaena*, and *Cyanobium*) generated from this 482 sample set using Basic Local Alignment Search Tool (BLAST) v. 2.8.1 (106). Using all 483 484 identified BGC clusters expanded our database of biosynthetic genes to allow for competitive mapping. Reads were kept for quantification if they were 95% identical to the query sequence, 485 and at least 80% in alignment length. If reads identically mapped to more than one sequence, i.e., 486 487 they had identical bit scores, the read count was divided among the total number of ambiguously mapped regions using the blast hit counts.sh script found at https://github.com/Geo-488 omics/scripts/tree/master/scripts. Reads mapped to a BGC were then normalized by BGC length. 489 To compare across samples, reads were also normalized by the 16S rRNA gene V4 490 variable region. This is based on the common assumption that all Microcystis cells contain one 491 16S rRNA gene, but do not contain all BGCs; therefore, we used this ratio to estimate the 492 relative abundance of BGCs in *Microcystis* populations. All V4 regions for the entire Phylum of 493 Cyanobacteria were used in the mapping databases to ensure competitive, sensitive, and accurate 494 495 mapping. These sequences were pulled from the SILVA v.1381.1 database (107) accessed in February 2021. This database can be found at: https://github.com/ceyancey/mcyGenotypes-496 databases. QC metagenomic reads were aligned to this database using BLAST and kept for 497 498 downstream analysis if they met the same parameters listed above. These alignment parameters, databases, and mapping tool quality were tested in full to ensure specificity and sensitivity. 499

Reads were then quantified in the same fashion described above and normalized by 16S rRNA
V4 region length. The equation below summarizes how relative BGC cluster abundance was
calculated:

Coverage ratio (*Microcystis BGC*: 16*s*):

 $=\frac{Reads mapped to BGC \div length of BGC (bp)}{Reads mapped to V4 region \div length of V4 region (bp)} = \frac{BGC coverage}{16s V4 coverage}$

To assess the completeness of cryptic PKS-containing clusters across sampling points, which have not been extensively characterized previously, we additionally calculated the coverage and percentage of nucleotides with reads mapped within a cluster. If at least 80% of the nucleotides within a BGC were covered, it was considered complete for that sampling point. If the percentage of nucleotides within gene clusters were less than 80%, they were considered absent or incomplete, and relative abundance was not calculated. A summary of coverage and completeness for gene clusters MIC 1, iterative PKS, and T3PKS is observed in Table S5.

510 Biosynthesis Gene Protein Phylogenies

To explore the novelty and relatedness of coarsely annotated PKS clusters from our 511 *Microcystis* MAGs to previously identified biosynthesis pathways, protein phylogenies were 512 513 generated. For the iterative PKS gene cluster, we assessed phylogenetic relationships for the PKSE gene, believed to be critical in enediyne biosynthesis. To generate a phylogenetic tree, we 514 used PKSE amino acid sequences from biosynthesis pathways known to encode enediyne-515 516 containing metabolites in addition to sequences that have been putatively identified as PKSEs in various cyanobacterial taxa (45, 73). The AufC amino acid sequence from aurafuron biosynthesis 517 was used as an outgroup. For the T3PKS cluster, the core T3PKS and a transport related gene 518 were selected for analysis. To find related protein sequences, these sequences were searched 519

against the nonredundant protein database on NCBI (accessed July 2022) using BLASTP (108).
Protein sequences were used to build phylogenetic trees if they had at least a 70% alignment
length and at least 30% identity to the target sequences. Protein sequences were aligned using
MUSCLE (109) and FastTree (110) was used to construct phylogenetic trees. Bootstrap values
equal to or greater than 0.5 were reported. PaperBLAST (https://papers.genomics.lbl.gov/cgibin/litSearch.cgi) was used to query published manuscripts to investigate whether these protein
sequences had been investigated previously.

527 BGC Cluster Transcript Abundance

528 The relative transcript abundance of BGCs *in situ* was also determined.

529 Metatranscriptomic raw reads can be accessed at NCBI under BioProject Accession:

530 PRJNA370007. Alignments were completed using BLAST with identical parameters mentioned

531 previously. However, to normalize gene expression, reads were mapped to reference genome

532 Microcystis aeruginosa PCC 7806SL (Accession: CP020771.1, GI: 1181755937) instead of the

533 V4 region. Competitive genome mapping was achieved by creating a database that contains

reference genomes for other common cyanobacteria found in western Lake Erie cyanoHABs:

- 535 Anabaena sp. 90 (GCA_000312705.1), Cyanobium sp. NIES-981 (GCA_900088535.1), and
- 536 *Pseudanabaena sp.* PCC 7367 (GCA_000317065.1). The following equation summarizes
- 537 calculations completed to relatively quantify BGC expression:

 $Relative Transcript Abundance = \frac{reads mapped to BGC \div length of BGC (bp)}{reads mapped to PCC 7806SL \div length of PCC 7806 SL (bp)}$

538 BGC Expression Correlation Analysis

To assess the relationships between expression and 1) abiotic environmental conditions

and 2) abundance of known competitors and predators of *Microcystis*, simple linear regressions

541	were generated. Correlations were generated for all abiotic variables, and for all organisms
542	identified in each metagenome that had at least 200 genes assigned to their taxonomic
543	identification. Organism abundance and analysis was completed at the genus level. Pearson's
544	correlation coefficient, R, as well as p-values were determined to assess the strength of
545	correlation and the significance of individual variables, respectively. Abiotic environmental
546	condition data was provided by CIGLR and can be found in Table S6. The relative abundance of
547	organisms was determined by aligning QC reads to the UniRef100 database using DIAMOND
548	(111). Relative abundance of each organism was then determine by calculating RPKM of all
549	genes assigned to the taxa using the AnnotateContigs.pl script found at
550	https://github.com/TealFurnholm/Strain-Level_Metagenome_Analysis/wiki/Step-2Gene-
551	Annotation-and-Community-Analysis which summarizes the complete community analysis
552	workflow and outputs.

553 Data Visualization

All figures were generated using R and R Studio v. 3.5.1 (112). Packages used for data 554 visualization included ggplot2 (113), ggthemes (114), and ggpubr (115). To generate the map in 555 Figure 1, QGIS was used using the Quick Map Services Plugin (QGIS Development Team, 556 557 2020). The base map used is an open-source resource, the Open Street Map (OSM, https://wiki.osmfoundation.org/wiki/Main Page). Clinker (104) software was used to generate 558 BGC comparison figures observed in Figure S1. Metagenomic and metatranscriptomic raw reads 559 can be found on NCBI under the BioProject Accessions PRJNA464361 and PRJNA370007, 560 respectively. 561

562 **Conflict of Interest**

563 The authors state no conflict of interest.

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574 Data availability

- All data pertinent to this manuscript is accessible for review within the manuscript. Metagenomic
- and metatranscriptomic raw reads can be found on NCBI under the BioProject Accessions
- 577 PRJNA464361 and PRJNA370007, respectively.

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879 Tables and Figure Captions

880 Table 1: Summary and Statistics of *Microcystis* MAGs

Station	Date	Completion (%)	Redundancy (%)	N50	Size (Mbp)	Number of contigs	GC Content (%)
	4-Aug	40.29	2.16	2907	0.816	335	39.67
	25-Aug	89.21	35.25	2353	5.99	2767	43.17
WE12	23-Sep	100	4.32	8601	5.31	903	42.79
	29-Sep	93.53	10.23	2548	4.27	1888	43.4
	20-Oct	96.4	12.95	4155	4.92	1461	42.83
	21-Jul	97.84	4.32	5000	4.27	1102	42.9
WE2	6-Oct	97.12	10.07	5061	5.39	1404	43.28
	20-Oct	99.28	5.76	5472	4.69	1108	42.8
WF4	29-Jul	92.81	20.86	1786	3.93	2234	43.53
	8-Sep	94.96	9.67	2668	4.29	1797	43.49

881

882

Table 2: Summary of Identified BGCs and putative synthesis products from WLE MAGs.

BGC name	Putative synthesis product	Described in
aer 1	aeruginosin	Ishida et al., 1999; Ishida et al., 2009;
aer 2	aeruginosin	al., 2013 (31, 117–119)
agd	aeruginoguanidine	Pancrace et al., 2019 (105)
apn	anabaenopeptin	Welker and Dohren 2006; Humbert
		et al., 2013 (13, 31)
cyanobactin 1	piricyclamide	

cyanobactin 2	piricyclamide	Leikoski et al., 2012 (62)
cyanobactin 3	piricyclamide	
<i>mcn</i> 1	cyanopeptolin	Tooming-Klunderud et al., 2007;
mcn 2	cyanopeptolin	al., 2013 (31, 120, 121)
mdn 1	microviridin	Ziemert et al., 2008; Humbert et al.,
mdn 2	microviridin	2013 (31, 122)
MIC1 (T1PKS-NRPS-hgIEKS)	unknown	Humbert et. al., 2013 (31)
Iterative PKS	unknown, possible	Humbert et. al., 2013 (31)
	enediyne	
PKS-M 1	microginin	Humbert et al., 2013; Eusébio et al.,
PKS-M 2	microginin	2022 (31, 43)
PKSmod-NRPSlike-T3PKS 1	cyclophane-like	Humbert et. al., 2013 (31)
PKSmod-NRPSlike-T3PKS 2	cyclophane-like	
T1PKS 1	unknown	Novel
T3PKS	unknown	Novel

884

Table 3: Abiotic and BGC expression Correlations.

	T1PKS NRPSlike (cyclopha	mod- T3PKS ne-like)	cyanob	actin	<i>mc</i> (cyanope	n ptolin)	<i>md</i> (microv	<i>n</i> 'irdin)	ae (aerugi	er inosin)
				N-		N-		N-		
C:N ratio	~0	C-rich	3-6	rich	4.7	rich	5.25	rich	~6	N-rich
	R-	p-	R-	p-	R-	p-	R-	p-	R-	p-
	squared	value	squared	value	squared	value	squared	value	squared	value
NO ₃	-0.551	0.041	-0.418	0.06	-0.434	0.049	-0.749	0.002	-0.48	0.028
Temperature	-0.598	0.04	-0.531	0.029	-0.412	0.089	-0.423	0.17	-0.375	0.1249
SRP	0.769	0.0013	0.251	0.27	0.216	0.35	0.358	0.21	0.364	0.1
ТР	0.601	0.023	0.114	0.62	0.227	0.32	0.537	0.048	0.449	0.04

886

Table 4: Pearson's correlations for BGC expression and selected organism abundance

		<i>aer</i> (aerug)	inosin	cyanob	actin	<i>mcn</i> (cyanoj n)	peptoli	<i>mdn</i> (micro)	virdin	T1PK NRPS T3PK (cycloj -like)	Smod- like S phane
	Genus	R	p- valu e	R	p- value	R	p- value	R	p- value	R	p- valu e
Ciliates	ICHTHYOPHTHIRIUS	0.553	0.03	0.245	0.378	0.336	0.221	0.653	0.041	0.79	0.00

			3							9	6
	PARAMECIUM	0.562	0.02	0.256	0.379	0.340	0.215	0.657	0.039	0.81	0.00
			9							0	5
	PSEUDOCOHNILEMBU	0.549	0.03	0.230	0.410	0.326	0.236	0.641	0.046	0.80	0.00
	S		4							3	5
	TETRAHYMENA	0.563	0.02	0.238	0.393	0.335	0.222	0.659	0.038	0.81	0.00
			9							7	4
Diatoms	CRASPEDOSTAUROS	0.567	0.05	0.298	0.347	0.436	0.156	0.414	0.308	0.85	0.00
			4							3	7
	SKELETONEMA	0.630	0.02	0.617	0.033	0.741	0.006	0.832	0.010	0.03	0.94
			8							1	1
Beta-	PARABURKHOLDERIA	0.296	0.28	0.081	0.773	0.117	0.678	0.380	0.279	0.64	0.04
proteobacteri			5							0	6
а	RHODOFERAX	0.409	0.06	0.419	0.059	0.488	0.025	0.669	0.009	0.17	0.55
			6							1	9
Proteobacteri	OLIGOFLEXUS	0.505	0.09	0.057	0.860	0.191	0.553	0.530	0.177	0.82	0.01
а			4							8	1
Cytophagia	CHRYSEOTALEA	0.405	0.09	0.149	0.555	0.276	0.030	0.360	0.250	0.44	0.14
			6							8	4

888

Figure 1: Overview of the 2014 W. Lake Erie cyanoHAB. A) Map of western Lake Erie and the 889 890 sampling sites used for this study. WE2 and WE12 are close to the Ohio coast and Maumee 891 River and are considered nearshore stations, while WE4 is located more centrally in the basin and considered offshore. B) Various bloom metrics for the 2014 bloom including cyanobacterial 892 893 biomass as measured by both chlorophyll α (blue) and phycocyanin (green) concentrations, as well as measured concentrations of particulate microcystins, soluble reactive phosphorus, and 894 nitrate. Figure 1A was generated via QGIS using the Open Street Map (OSM) 895 (https://wiki.osmfoundation.org/wiki/Main Page). 896 Figure 2: Relative abundance of biosynthetic gene clusters (BGCs), including the complete mcy 897 operon described previously (33). Relative abundance is presented as estimated relative 898 proportion of the *Microcystis* population containing the BGC (see methods). Predicted secondary 899 metabolites are indicated on the right vertical axis) an in the legend. On 4 August at WE2, the 900

901 *mcy* operon and T3PKS gene clusters had a relative proportion greater than 1, indicating more

902 copies of BGC genes than 16S rRNA genes.

903 Figure 3: Comparison of PKS-M BGCs identified in WLE MAGs to the known microginins encoding BGC from Microcystis aeruginosa LEGE 91341. A) Both PKS-M gene clusters are 904 incomplete/fragmented compared to the complete cluster where PKS-M contains conserved 905 additional synthesis and hypothetical protein encoding genes while PKS-M 2 lacks these genes 906 and contains a putative nuclease. Both PKS-M clusters contain a fragmented hybrid NRPS-PKS 907 908 encoding gene found in the microginin encoding pathway. B) Read mapping from a sample collected on 4-August at WE2 against the complete microginin biosynthesis pathway in LEGE 909 91341. Coverage is observed throughout with notable variation in sequence identity and 910 911 evenness.

Figure 4: Gene schematics for select cryptically annotated PKS-containing gene clusters. A) The
iterative PKS gene cluster contains transport genes, a putative enediyne biosynthesis gene, a
thioesterase, 2/3 genes of unknown function that are part of the enediyne biosynthesis cassette,
and a putative nucleases B) The T3PKS cluster contains several transport genes and a
cytochrome B5 binding gene. C) The MIC cluster contains several core biosynthesis and
transport genes, and a tryptophan halogenase.

Figure 5: Phylogenetic tree consisting of protein sequences from the putatively identified PKSE
from the WLE MAGs, putatively identified PKSEs from previous studies (45, 73), and PKSEs
from known biosynthesis pathways that synthesize enediyne containing molecules. The PKSE
from the WLE MAGs is colored red, the putative PKSEs identified in other cyanobacteria are
colored in green, and a black circle indicates a known synthesized product associated with the
PKSE enzyme.

Figure 6: Relative abundance of transcripts from BGCs throughout the 2014 Bloom separated
by sampling station. Relative transcript abundance was calculated by determining the number of

- 926 reads mapped to a BGC via specific cut off parameters, normalized to the number of reads
- 927 mapped to an entire reference *Microcystis* genome to determine "expression effort" (see
- 928 Experimental Procedure).

A)





Date



Microginin BGC from *Microcystis aeruginosa* LEGE 91341 (BGC0002623, MiBIG)



A)



Metagenomic Read Mapping to *Microcystis aeruginosa* LEGE 91341, a sample collected on 4-Aug at WE2 B)



A) iterative PKS



B) T3PKS



C) MIC 1



T1PKS - enediyne biosynthesis





SUPPLEMENTALS



Supplemental Figure 1: Comparison of gene order, orientation, and sequence similarity between distinct BGCs from different *Microcystis* MAGs. Clinker was used to visually compare similarities and differences of the A) aeruginosin, B) cyanobactin, C) microviridin B, and D) micropeptin BGC classes. Homologous genes are the same color and percent similarity between genes is shown via a bar connecting homologous genes.



Supplemental Figure 2: Gene Schematics for identified BGC clusters including A) anabaenopeptin, B) microviridin B, C) aeruginosin, and D) a cyanobactin with 58% similarity to piricyclamide. Gene schematics were rendered through AntiSMASH v.6. Examples of chemical congeners that may be synthesized by these BGCs are shown. Percent similarity represents the percentage of genes within the closest known compound on the MiBIG database that have a significant BLAST hit



Pks18 is involved in the biosynthesis of tri- and

tetraketide alpha-pyrones

Alpha-pyrone synthesis polyketide synthase-like Pks18 [Nocardia seriolae] 52 yea III polyketide synthase [Nocardia seriolae] polyketide synthase [Nocardia seriolae] a stype III polyketide synthase [Nocardia seriolae] type III polyketide synthase [Nocardia seriolae] etype III polyketide synthase [Nocardia seriolae] etype III polyketide synthase [Nocardia seriolae] etype III polyketide synthase [Nocardia seriolae] **Supplemental Figure 3:** Phylogenies for A) a transport protein and B) core biosynthesis naringenin-chalcone synthase from the T3PKS gene cluster. The transport protein has high homology to proteins involved in siderophore transport while the naringenin chalcone synthase is similar to cyanobacterial naringenin chalcone synthases. Bootstrap values are included at nodes where values are greater than 0.5. Proteins that share the same functional annotation of interest are colored. Dark red and bolded protein sequences are from the W. Lake Erie cyanoHAB as identified through antiSMASH.

Cluster ID	Known Compound Cluster Hit	Similarity (%)
T1Iterative PKS	Calicheamicin	4
T3PKS	NA	NA
MIC 1	Phenalamide	40

Table S1: PKS Cluster Similarities to known compounds as identified via antiSMASH.

Table S2: Annotations for genes identified in the MIC BGC. NRPS/PKS domains are provided when applicable as well as proposed functions. DCL = Condensation linking L-amino acids to a peptide ending with a D-amino acid. LCL = Condensation linking L-amino acid to a peptide ending with an L-amino acid as described via antiSMASH output.

Gene	Length (bp)	Function	Domains	Proposed Function
1	3099	Transport-related		Permease, efflux RND
2	1545	Transport-related		Periplasmic adapter, efflux RND
3	975	Other/Hypothetical		Unknown/Hypothetical
4	903	Regulatory		Transcriptional Regulator
5	519	Other/Hypothetical		Unknown/Hypothetical
6	261	Additional Biosynthesis	Acyl Carrier	Acyl Carrier Protein
7	1797	Additional Biosynthesis		Halogenase
8	4734	Core - T1PKS	Hybrid Ketosynthase, Acyltransferase, Ketoreductase, PP-binding	T1PKS
9	4824	Core - T1PKS	Hybrid Ketosynthase, Acyltransferase, PP- binding, Condensation DCL, PP-binding	T1PKS
10	2925	Core - NRPS	AMP-binding, Condensation LCL	NRPS
11	3822	Core - PKS	Hybrid Ketosynthase, Acyltransferase, PP- binding	PKS
12	720	Other/Hypothetical		Unknown/Hypothetical
13	1758	Additional Biosynthesis	Ketoreductase	Short chain Dehydrogenase/Reductase

14	5004	Additional Biosynthesis	Acyltransferase	Short chain Dehydrogenase/Reductase
15	1575	Additional Biosynthesis		Enoyl reductase
16	1173	Transport-related		MFS transporter
17	1278	Other		FAD-dependent oxidoreductase
18	765	Other/Hypothetical		Unknown/Hypothetical
19	876	Other/Hypothetical		Unknown/Hypothetical
20	894	Transport-related		ABC transporter

Table S3: Correlations of BGC relative transcript abundance and available abiotic variables

		p-
Variable	R-squared	value
NO3	-0.418	0.59
NH4	0.188	0.41
Temperature	-0.514	0.029
pН	-0.129	0.58
SRP	0.251	0.27
TP	0.114	0.62
PC	-0.0536	0.82
chlA	-0.0603	0.8

Aeruginosins (aer)						
		p-				
Variable	R-squared	value				
NO3	-0.48	0.028				
NH4	0.0988	0.67				
Temperature	-0.375	0.1249				
pН	0.0413	0.859				
SRP	0.364	0.1				
ТР	0.449	0.04				
PC	0.252	0.27				
chlA	0.162	0.48				

PKSmod-NRPSlike-T3PKS

		p-
Variable	R-squared	value
NO3	-0.551	0.041
NH4	0.0802	0.79
Temperature	-0.598	0.04
pН	-0.238	0.41
SRP	0.769	0.0013
TP	0.601	0.023
PC	0.129	0.66
chlA	-0.0753	0.8

Cyanopeptolin <i>(mcn)</i>		
		p-
Variable	R-squared	value
NO3	-0.434	0.049
NH4	0.178	0.44
Temperature	-0.412	0.089
pН	-0.00607	0.98
SRP	0.216	0.35
TP	0.227	0.32
PC	0.0991	0.669
chlA	0.0839	0.72

Microviridin (<i>mdn</i>)					
		p-			
Variable	R-squared	value			
NO3	-0.749	0.002			
NH4	0.00647	0.98			
Temperature	-0.423	0.17			
pН	0.144	0.62			
SRP	0.358	0.21			
ТР	0.537	0.048			
PC	0.312	0.28			
chlA	0.218	0.45			

Table S4: Correlations of BGC relative transcript abundance and relative abundance of select organisms

		R-	
BGC Class	Organism Genus	squared	p-value
Cyanobactin	UNCLASSIFIED_GEMMATIMONADACEAE_GENUS	0.532002	0.023055
Cyanobactin	HERBASPIRILLUM	-0.64425	0.023743
Cyanobactin	WOLBACHIA	-0.52319	0.025879
Cyanobactin	UNCLASSIFIED_CYTOPHAGALES_GENUS	0.444163	0.04368

Cyanobactin	RHODOFERAX	0.418508	0.059008
Cyanobactin	GEMMATIMONAS	-0.41654	0.060332
Cyanobactin	ROSEOMONAS	-0.4043	0.069096
Cyanobactin	UNCLASSIFIED_VERRUCOMICROBIACEAE_GENUS	-0.45766	0.086274
Cyanobactin	UNCLASSIFIED_ARMATIMONADETES_GENUS	-0.378	0.09112
Cyanobactin	ACINETOBACTER	-0.50547	0.093656
Cyanobactin	NOVIHERBASPIRILLUM	-0.48984	0.105981
Cyanobactin	UNCLASSIFIED_GEMMATIMONADETES_GENUS	0.352644	0.116898
Cyanobactin	UNCLASSIFIED_SYNECHOCOCCALES_GENUS	-0.41704	0.121975
Cyanobactin	BREVUNDIMONAS	-0.34195	0.129211
Cyanobactin	PIRELLULA	-0.35852	0.144021
Cyanobactin	BDELLOVIBRIO	-0.35723	0.145583
Cyanobactin	UNCLASSIFIED_PHYCISPHAERAE_GENUS	-0.35563	0.147523
Cyanobactin	SKELETONEMA	0.617171	0.03252
Cyanobactin	APOSTICHOPUS	-0.51817	0.08438
Cyanobactin	PHYTOPHTHORA	0.372823	0.12757
Cyanobactin	EURYTEMORA	0.32275	0.153585
aer (aeruginosin)	UNCLASSIFIED BATHYARCHAEOTA GENUS	-0.54321	0.010932
aer (aeruginosin)	UNCLASSIFIED EURYARCHAEOTA GENUS	-0.5217	0.081919
aer (aeruginosin)	WOLBACHIA	-0.68302	0.001783
aer (aeruginosin)	UNCLASSIFIED_VERRUCOMICROBIACEAE_GENUS	-0.71213	0.002897
aer (aeruginosin)	GEMMATIMONAS	-0.56891	0.007115
aer (aeruginosin)	ACINETOBACTER	-0.72704	0.007383
aer (aeruginosin)	PAUCIBACTER	-0.71693	0.008691
aer (aeruginosin)	UNCLASSIFIED_ARMATIMONADETES_GENUS	-0.54466	0.010681
aer (aeruginosin)	UNCLASSIFIED_SYNECHOCOCCALES_GENUS	-0.62251	0.013194
aer (aeruginosin)	UNCLASSIFIED_CHLOROFLEXI_GENUS	-0.50549	0.019408
aer (aeruginosin)	BRYOBACTER	-0.58735	0.021321
aer (aeruginosin)	UNCLASSIFIED VERRUCOMICROBIALES GENUS	-0.49482	0.02258
aer (aeruginosin)	UNCLASSIFIED_VERRUCOMICROBIA_GENUS	-0.49006	0.024122
aer (aeruginosin)	ROSEOMONAS	-0.48363	0.026336
aer (aeruginosin)	POLYNUCLEOBACTER	-0.56913	0.02681
aer (aeruginosin)	METHYLOBACTERIUM	-0.47444	0.029776
aer (aeruginosin)	UNCLASSIFIED_ACTINOMYCETIA_GENUS	-0.47312	0.0303
aer (aeruginosin)	RUNELLA	-0.62123	0.031077
aer (aeruginosin)	UNCLASSIFIED_GEMMATIMONADACEAE_GENUS	0.497769	0.035545
aer (aeruginosin)	UNCLASSIFIED_RHODOBACTERACEAE_GENUS	-0.49136	0.03837
aer (aeruginosin)	SYNECHOCOCCUS	-0.53619	0.03936
aer (aeruginosin)	LIMNOHABITANS	-0.43737	0.047397
aer (aeruginosin)	UNCLASSIFIED_OPITUTAE_GENUS	-0.51859	0.047632
aer (aeruginosin)	GEMMOBACTER	-0.43359	0.049571
aer (aeruginosin)	BREVUNDIMONAS	-0.43254	0.050189

aer (aeruginosin)	UNCLASSIFIED_FLAVOBACTERIACEAE_GENUS	-0.50918	0.052554
aer (aeruginosin)	HERBASPIRILLUM	-0.5664	0.054859
aer (aeruginosin)	PLANKTOPHILA	-0.55568	0.060676
aer (aeruginosin)	UNCLASSIFIED_PELAGIBACTERALES_GENUS	-0.54773	0.065257
aer (aeruginosin)	RHODOFERAX	0.408822	0.065753
aer (aeruginosin)	BELNAPIA	-0.48663	0.065843
aer (aeruginosin)	UNCLASSIFIED_BRYOBACTERALES_GENUS	-0.46307	0.082158
aer (aeruginosin)	TABRIZICOLA	-0.37622	0.092781
aer (aeruginosin)	OLIGOFLEXUS	0.504783	0.094177
aer (aeruginosin)	CHRYSEOTALEA	0.404708	0.095728
aer (aeruginosin)	UNCLASSIFIED_PHYCISPHAERAE_GENUS	-0.39564	0.104123
aer (aeruginosin)	SICCIRUBRICOCCUS	-0.43209	0.10774
aer (aeruginosin)	AQUINCOLA	-0.36081	0.108089
aer (aeruginosin)	UNCLASSIFIED_CYTOPHAGALES_GENUS	0.360119	0.108811
aer (aeruginosin)	ROSEOCOCCUS	-0.42753	0.111925
aer (aeruginosin)	BOSEA	0.381182	0.118589
aer (aeruginosin)	ALGORIPHAGUS	-0.47269	0.120699
aer (aeruginosin)	PARACRAUROCOCCUS	-0.41354	0.125466
aer (aeruginosin)	DERXIA	-0.40095	0.138566
aer (aeruginosin)	NOVIHERBASPIRILLUM	-0.44759	0.144545
aer (aeruginosin)	CHITINOPHAGA	-0.32747	0.147314
aer (aeruginosin)	PIRELLULA	-0.35561	0.147549
aer (aeruginosin)	APOSTICHOPUS	-0.70705	0.010128
aer (aeruginosin)	SKELETONEMA	0.630377	0.027992
aer (aeruginosin)	TETRAHYMENA	0.562994	0.028878
aer (aeruginosin)	PARAMECIUM	0.561877	0.029267
aer (aeruginosin)	TIGRIOPUS	-0.46945	0.031789
aer (aeruginosin)	PARAMURICEA	-0.46756	0.032577
aer (aeruginosin)	ICHTHYOPHTHIRIUS	0.552709	0.032614
aer (aeruginosin)	PSEUDOCOHNILEMBUS	0.549228	0.033957
aer (aeruginosin)	NYMPHAEA	0.544111	0.036006
aer (aeruginosin)	LEPEOPHTHEIRUS	-0.46573	0.051427
aer (aeruginosin)	CRASPEDOSTAUROS	0.567259	0.054411
aer (aeruginosin)	ATTHEYA	-0.38152	0.087909
aer (aeruginosin)	STYLONYCHIA	-0.50671	0.092719
aer (aeruginosin)	CHAETOCEROS	-0.50033	0.097601
aer (aeruginosin)	NOCCAEA	0.440112	0.100646
aer (aeruginosin)	DITYLUM	-0.36555	0.103194
aer (aeruginosin)	UNCLASSIFIED_SIPHOVIRIDAE_GENUS	-0.58451	0.022115
aer (aeruginosin)	CYANOPHAGE	-0.44199	0.044845
aer (aeruginosin)	UNCLASSIFIED_CAUDOVIRALES_GENUS	-0.42915	0.052221
aer (aeruginosin)	UNCLASSIFIED PODOVIRIDAE GENUS	-0.39876	0.101175

	mcn (cyanopeptolin)	UNCLASSIFIED BATHYARCHAEOTA GENUS	-0.38504	0.084774
	mcn (cyanopeptolin)	UNCLASSIFIED_GEMMATIMONADACEAE_GENUS	0.625287	0.00552
	mcn (cyanopeptolin)	WOLBACHIA	-0.62255	0.005792
	mcn (cyanopeptolin)	HERBASPIRILLUM	-0.69002	0.013013
	mcn (cyanopeptolin)	GEMMATIMONAS	-0.49778	0.021662
	mcn (cyanopeptolin)	ACINETOBACTER	-0.64436	0.023713
	mcn (cyanopeptolin)	RHODOFERAX	0.488081	0.024786
	mcn (cyanopeptolin)	PAUCIBACTER	-0.61405	0.033661
	mcn (cyanopeptolin)	UNCLASSIFIED_VERRUCOMICROBIACEAE_GENUS	-0.54704	0.03482
	mcn (cyanopeptolin)	UNCLASSIFIED_ARMATIMONADETES_GENUS	-0.45981	0.035975
	mcn (cyanopeptolin)	ROSEOMONAS	-0.44897	0.04119
	mcn (cyanopeptolin)	POLYNUCLEOBACTER	-0.51391	0.050033
	mcn (cyanopeptolin)	UNCLASSIFIED CYTOPHAGALES GENUS	0.427287	0.053361
	mcn (cyanopeptolin)	PLANKTOPHILA	-0.54341	0.067844
	mcn (cyanopeptolin)	UNCLASSIFIED CHLOROFLEXI GENUS	-0.40392	0.06938
	mcn (cyanopeptolin)	BREVUNDIMONAS	-0.4015	0.071227
	mcn (cyanopeptolin)	UNCLASSIFIED PHYCISPHAERAE GENUS	-0.43177	0.073576
	mcn (cyanopeptolin)	UNCLASSIFIED PELAGIBACTERALES GENUS	-0.52941	0.076714
	mcn (cyanopeptolin)	NOVIHERBASPIRILLUM	-0.52646	0.078678
	mcn (cyanopeptolin)	BELNAPIA	-0.45468	0.088609
	mcn (cyanopeptolin)	UNCLASSIFIED SYNECHOCOCCALES GENUS	-0.45449	0.088754
	mcn (cyanopeptolin)	ROSEOCOCCUS	-0.44649	0.095243
	mcn (cyanopeptolin)	UNCLASSIFIED ACTINOMYCETIA GENUS	-0.36592	0.102817
	mcn (cyanopeptolin)	SICCIRUBRICOCCUS	-0.43354	0.106437
	mcn (cyanopeptolin)	UNCLASSIFIED_OPITUTAE_GENUS	-0.42291	0.116282
	mcn (cyanopeptolin)	PARACRAUROCOCCUS	-0.41768	0.121342
	mcn (cyanopeptolin)	METHYLOBACTERIUM	-0.33637	0.135985
	mcn (cyanopeptolin)	SKELETONEMA	0.741023	0.005825
	mcn (cyanopeptolin)	APOSTICHOPUS	-0.68998	0.01302
	mcn (cyanopeptolin)	TIGRIOPUS	-0.37636	0.092651
	mcn (cyanopeptolin)	PARAMURICEA	-0.36919	0.09955
	mcn (cyanopeptolin)	PSEUDO	0.346343	0.124046
	mcn (cyanopeptolin)	CYANOPHAGE	-0.34948	0.120445
	PKSmod-NRPSlike-			
l	T3PKS	UNCLASSIFIED_BATHYARCHAEOTA_GENUS	-0.56172	0.036579
	T3PKS	WOLBACHIA	-0.76316	0.003881
	PKSmod-NRPSlike-			
	T3PKS	OLIGOFLEXUS	0.827902	0.011155
	T3PKS	UNCLASSIFIED SYNECHOCOCCALES GENUS	-0.73805	0.014805
	PKSmod-NRPSlike-			
	T3PKS	PAUCIBACTER	-0.80483	0.015971
	T3PKS	UNCLASSIFIED BDELLOVIBRIONALES GENUS	0.710863	0.021193

PKSmod-NRPSlike-	UNCLASSIFIED VERBUCOMICROBIALES GENUS	-0.6066	0 021449
PKSmod-NRPSlike-	ONCLASSIFIED_VERROCOMICROBIALES_GENOS	-0.0000	0.021449
T3PKS	UNCLASSIFIED_VERRUCOMICROBIA_GENUS	-0.60468	0.021977
PKSmod-NRPSlike-	METHVLODACTEDIUM	0 50006	0.026226
PKSmod-NRPSlike-	MEIHILOBACIERIUM	-0.39000	0.020330
T3PKS	BOSEA	0.628276	0.02868
PKSmod-NRPSlike-			
T3PKS	UNCLASSIFIED_XANTHOMONADALES_GENUS	0.754947	0.030359
PKSmod-NRPSlike-	UNCLASSIFIED VERRUCOMICROBIACEAE GENUS	-0 6773	0.031427
PKSmod-NRPSlike-		0.0775	0.051427
T3PKS	UNCLASSIFIED_ARMATIMONADETES_GENUS	-0.56951	0.033514
PKSmod-NRPSlike-			
T3PKS	GEMMATIMONAS	-0.55262	0.040423
T3PKS	PARABURKHOLDERIA	0.639625	0 046424
PKSmod-NRPSlike-		0.0000020	0.0.0.2
T3PKS	OHTAEKWANGIA	0.629832	0.050985
PKSmod-NRPSlike-		0.50000	0.05104
T3PKS	UNCLASSIFIED_HYPHOMICROBIALES_GENUS	-0.53039	0.05104
T3PKS	ROSEOMONAS	-0.50999	0.062449
PKSmod-NRPSlike-		0.0000000	0.0021.0
T3PKS	UNCLASSIFIED_PLANCTOMYCETACEAE_GENUS	0.50495	0.065528
PKSmod-NRPSlike-		0.57255	0.002004
I JPKS PKSmod_NPPSlike	UNCLASSIFIED_OPTIUTAE_GENUS	-0.5/355	0.083004
T3PKS	UNCLASSIFIED CHLOROFLEXI GENUS	-0.46759	0.091806
PKSmod-NRPSlike-			
T3PKS	SYNECHOCOCCUS	-0.55893	0.093031
PKSmod-NRPSlike-	UNCLASSIELED ACTINOMYCETLA CENUS	0 16576	0.002257
PKSmod-NRPSlike-	UNCLASSIFIED_ACTINOMICETIA_GENUS	-0.40370	0.093237
T3PKS	ACINETOBACTER	-0.62657	0.096445
PKSmod-NRPSlike-			
T3PKS	RHODOPIRELLULA	0.624078	0.098181
T3PKS	POLYNIJCI FOR ACTER	-0 54784	0 10113
PKSmod-NRPSlike-	TOEINOCLEOBACIER	-0.34704	0.10115
T3PKS	VAMPIROVIBRIO	0.61026	0.108112
PKSmod-NRPSlike-			
T3PKS	UNCLASSIFIED_FLAVOBACTERIACEAE_GENUS	-0.53047	0.114691
T3PKS	CYANOBIUM	-0 51312	0 129322
PKSmod-NRPSlike-		0.01012	0.12/022
T3PKS	LIMNOHABITANS	-0.42267	0.132157
PKSmod-NRPSlike-		0 4 4 7 7 4 7	0.1.4.420
T3PKS	CHRYSEOTALEA	0.447/47	0.14439
T3PKS	TETRAHYMENA	0.817487	0.00387
PKSmod-NRPSlike-			
T3PKS	PARAMECIUM	0.809702	0.004528
PKSmod-NRPSlike-		0.0000005	0.005104
ISPKS	<i>L'SEODOCOHNILEMBOS</i>	0.803325	0.005124

DKSmod NDDSlike			
T3PKS	ICHTHYOPHTHIRIUS	0.799409	0.005517
PKSmod-NRPSlike-			
T3PKS	NYMPHAEA	0.787085	0.006891
T3PKS	CRASPEDOSTAUROS	0.853161	0.007069
PKSmod-NRPSlike-			
T3PKS	CHAETOCEROS	-0.78724	0.020399
T3PKS	ATTHEYA	-0.58158	0.029142
PKSmod-NRPSlike-			
T3PKS	DITYLUM	-0.57856	0.030194
PKSmod-NRPSlike-	NOCCAEA	0 673847	0.032641
PKSmod-NRPSlike-	NOCCAEA	0.073047	0.032041
T3PKS	PHAEODACTYLUM	-0.55405	0.0398
PKSmod-NRPSlike-		0	0.045021
T3PKS PKSmod NDPSlike	ODONTELLA	-0.5466	0.065931
T3PKS	GRAMMATOPHORA	-0.45155	0.140597
PKSmod-NRPSlike-		0	
T3PKS	PARAMURICEA	-0.41046	0.144905
PKSmod-NRPSlike-		0 4076	0.149004
I 3PKS PK Smod-NRPSlike-	FISTULIFERA	-0.40/6	0.148004
T3PKS	TIGRIOPUS	-0.40737	0.148253
PKSmod-NRPSlike-			
T3PKS	UNCLASSIFIED_SIPHOVIRIDAE_GENUS	-0.5798	0.078941
PKSmod-NRPSlike-	UNCLASSIFIED PODOVIRIDAE GENUS	-0 525	0 079661
PKSmod-NRPSlike-	UNCLASSIFIED_I ODOVINIDAE_OENOS	-0.525	0.079001
T3PKS	UNCLASSIFIED_CAUDOVIRALES_GENUS	-0.43758	0.117642
mdn (microviridin)	UNCLASSIFIED_BATHYARCHAEOTA_GENUS	-0.54366	0.044485
mdn (microviridin)	UNCLASSIFIED EURYARCHAEOTA GENUS	<u>-0.581</u> 47	0.130566
<i>mdn</i> (microviridin)	WOLBACHIA	-0.80994	0.001406
mdn (microviridin)	UNCLASSIFIED VERRUCOMICROBIACEAE GENUS	-0.82399	0.003375
mdn (microviridin)	PAUCIBACTER	-0.87028	0.00494
mdn (microviridin)	GEMMATIMONAS	-0.68815	0.006514
mdn (microviridin)	ACINETOBACTER	-0.84757	0.007873
mdn (microviridin)	RHODOFERAX	0.66903	0.008884
mdn (microviridin)	UNCLASSIFIED ARMATIMONADETES GENUS	-0 64982	0.011887
mdn (microviridin)	UNCLASSIFIED CHIOROFIEYI GENUS	-0 63376	0.014948
<i>mdn</i> (microviridin)	POI VNIICI FOR ACTER	-0.03370	0.014940
<i>mdn</i> (microviridin)	I OLINOCLEODACIEN	0.70006	0.020003
<i>mdn</i> (microviridin)	UNCLASSIFIED_SINECHUCUCCALES_GENUS	-0./0900	0.0210/4
mdn (microviridin)		-0.39088	0.0260/6
man (microvinidin)	HERBASPIRILLUM	-0.75803	0.029302
man (microviriain)	UNCLASSIFIED_ACTINOMYCETIA_GENUS	-0.56311	0.036019
man (microviridin)	UNCLASSIFIED_GEMMATIMONADACEAE_GENUS	0.601612	0.038501
<i>man</i> (microviridin)	LIMNOHABITANS	-0.54686	0.043
mdn (microviridin)	BRYOBACTER	-0.64031	0.046116

<i>mdn</i> (microviridin)	BREVUNDIMONAS	-0.51833	0.057584
mdn (microviridin)	PLANKTOPHILA	-0.67286	0.067456
mdn (microviridin)	UNCLASSIFIED_OPITUTAE_GENUS	-0.59786	0.067924
mdn (microviridin)	SYNECHOCOCCUS	-0.59707	0.06838
mdn (microviridin)	METHYLOBACTERIUM	-0.48917	0.075866
mdn (microviridin)	UNCLASSIFIED_VERRUCOMICROBIALES_GENUS	-0.48596	0.078103
mdn (microviridin)	UNCLASSIFIED_PELAGIBACTERALES_GENUS	-0.64992	0.081073
mdn (microviridin)	UNCLASSIFIED_VERRUCOMICROBIA_GENUS	-0.47108	0.089076
mdn (microviridin)	UNCLASSIFIED_FLAVOBACTERIACEAE_GENUS	-0.55257	0.097624
mdn (microviridin)	RUNELLA	-0.6133	0.105882
mdn (microviridin)	UNCLASSIFIED_CYTOPHAGALES_GENUS	0.440241	0.11517
mdn (microviridin)	BELNAPIA	-0.52926	0.115681
mdn (microviridin)	NOVIHERBASPIRILLUM	-0.58682	0.126215
mdn (microviridin)	VAMPIROVIBRIO	-0.57153	0.13887
mdn (microviridin)	UNCLASSIFIED_RHODOBACTERACEAE_GENUS	-0.4489	0.143238
mdn (microviridin)	APOSTICHOPUS	-0.97738	2.84E-05
mdn (microviridin)	SKELETONEMA	0.831539	0.010493
mdn (microviridin)	PARAMECIUM	0.659467	0.038036
mdn (microviridin)	TETRAHYMENA	0.656704	0.039138
mdn (microviridin)	TIGRIOPUS	-0.55241	0.040515
mdn (microviridin)	ICHTHYOPHTHIRIUS	0.653081	0.040614
mdn (microviridin)	NYMPHAEA	0.651423	0.041302
mdn (microviridin)	PSEUDOCOHNILEMBUS	0.641096	0.045763
mdn (microviridin)	NOCCAEA	0.588531	0.073478
mdn (microviridin)	PARAMURICEA	-0.49062	0.07487
mdn (microviridin)	LEPEOPHTHEIRUS	-0.49739	0.099902
mdn (microviridin)	EURYTEMORA	0.442545	0.113059
mdn (microviridin)	CYANOPHAGE	-0.53873	0.046846
mdn (microviridin)	UNCLASSIFIED CAUDOVIRALES GENUS	-0.48634	0.077838
mdn (microviridin)	UNCLASSIFIED_SIPHOVIRIDAE_GENUS	-0.54786	0.101118
mdn (microviridin)	UNCLASSIFIED PODOVIRIDAE GENUS	-0.45023	0.141909

Table S5: Coverage and Completeness metrics for cryptic PKS containing cluster T3PKS, MIC 1 and Iterative PKS.

Station	Date	BGC Name	Bases with	Average	Status
Station	7/8/2014	T3PKS	83 798	6	Complete
	7/8/2014	MIC 1	81.36	5.96	Complete
	7/8/2014	Iterative PKS	28.455	1.647	Absent
	8/4/2.014	T3PKS	100	2431	Complete
	8/4/2014	MIC 1	100	28.85	Complete
	8/4/2014	Iterative PKS	100	923.7	Complete
	8/25/2014	T3PKS	100	110.9	Complete
	8/25/2014	MIC 1	100	239.2	Complete
WE12	8/25/2014	Iterative PKS	100	83.51	Complete
WE12	9/23/2014	T3PKS	87.26	64.67	Complete
	9/23/2014	MIC 1	100	339.4	Complete
	9/23/2014	Iterative PKS	57.053	62.49	Absent
	9/29/2014	T3PKS	100	188.5	Complete
	9/29/2014	MIC 1	100	827.3	Complete
	9/29/2014	Iterative PKS	100	170.6	Complete
	10/20/2014	T3PKS	79.603	21.62	Complete
	10/20/2014	MIC 1	98.53	28.11	Complete
	10/20/2014	Iterative PKS	100	57.804	Complete
	7/21/2014	T3PKS	47.75	9.126	Absent
	7/21/2014	MIC 1	100	11.089	Complete
	7/21/2014	Iterative PKS	91.82	16.79	Complete
	8/4/2014	T3PKS	100	1217	Complete
	8/4/2014	MIC 1	100	1275	Complete
WE2	8/4/2014	Iterative PKS	100	303.4	Complete
	9/29/2014	T3PKS	100	521.4	Complete
	9/29/2014	MIC 1	100	1599	Complete
	9/29/2014	Iterative PKS	100	312.9	Complete
	10/6/2014	T3PKS	85.303	167.1	Complete
	10/6/2014	MIC 1	100	49.09	Complete

	10/6/2014	Iterative PKS	100	214.4	Complete
	10/20/2014	T3PKS	93.98	81.19	Complete
	10/20/2014	MIC 1	100	53.84	Complete
	10/20/2014	Iterative PKS	100	230.2	Complete
	7/29/2014	T3PKS	98.85	31.6	Complete
	7/29/2014	MIC 1	100	52.59	Complete
WEA	7/29/2014	Iterative PKS	98.75	33.4	Complete
	8/4/2014	T3PKS	100	29.42	Complete
	8/4/2014	MIC 1	100	70.94	Complete
	8/4/2014	Iterative PKS	100	21.82	Complete
VV 124	9/8/2014	T3PKS	97.49	28.78	Complete
	9/8/2014	MIC 1	100	71.88	Complete
	9/8/2014	Iterative PKS	94.43	23.79	Complete
	9/29/2014	T3PKS	99.6	24.26	Complete
	9/29/2014	MIC 1	100	89.52	Complete
	9/29/2014	Iterative PKS	100	19.56	Complete

Table S6: Paired environmental variables for the 2014 cyanoHAB.

Station	Date	NO3 (mg/L)	NH4 (ug/L)	Temperat ure (°C)	Latitude	Longitidue	Distance to Maumee River (km)	Shore	рН	Particulat e_Microcy stins (ug/L)	Phycocy anin (ug/L)	CHLa (ug/L)	SRP (ug/L)	TP (ug/L)	DOC (uM)
WE12	8-Jul-14	0.34	16.16	23.5	NA	NA	17.44	Near	8.33	NA	0.12	6.98	0.2	34	214
WE12	4-Aug-14	0.62	3.25	24.6	41 42.157	83 15.781	16.35	Near	9.29	9.28	45.43	54.46	0.1	44.4	384
WE12	25-Aug-14	0.03	2.58	25	41 42.298	83 15.377	16.88	Near	9.29	3.1	45.72	38.28	3.68	98.1	415
WE12	23-Sep-14	0.15	3.14	17.9	41 42.196	83 15.499	16.72	Near	8.33	0.9	17.55	16.97	17.39	60.3	495
WE12	29-Sep-14	0.14	0.44	19.5	41 42.185	83 15.546	16.7	Near	8.74	0.6	21.87	22.25	7.98	69.2	398
WE12	20-Oct-14	0.16	5.18	13	41 42.269	83 15.663	16.54	Near	7.98	NA	0.54	5	13.27	45.2	419
WE2	21-Jul-14	0.52	2.93	23.5	41 45.873	83 19.849	12.23	Near	8.38	4.94	5.13	16.26	0.1	32.1	311
WE2	4-Aug-14	0.56	3.03	23.1	41 45.912	83 19.835	12.3	Near	9.2	4.33	39.01	29.44	1.45	32	462
WE2	29-Sep-14	0.15	0.1	19.4	41 45.884	83 20.107	11.95	Near	8.84	4.2	205.5	66.63	21.22	65.3	601
WE2	6-Oct-14	0.15	11.18	15.3	41 45.795	83 19.942	12.08	Near	8.04	NA	6.24	8.06	12.65	53	502
WE2	20-Oct-14	0.16	6.57	12.5	41 45.827	83 19.855	12.2	Near	8.04	NA	2.12	7.23	9.16	44.3	287
WE4	29-Jul-14	0.31	9.39	22.6	41 49.637	83 11.713	25.53	Off	8.59	1.01	4.77	6.12	1.01	16.5	328
WE4	4-Aug-14	0.14	1.64	24.6	41 49.714	83 11.654	25.64	Off	9.02	1.54	15.38	18.11	0.55	22.1	332
WE4	8-Sep-14	0.07	5.28	NA	41 49.585	83 11.629	25.55	Off	8.86	0.8	15.81	26.24	0.27	30.6	328
WE4	29-Sep-14	0.19	0.33	19.9	41 49.608	83 11.684	25.44	Off	8.58	0.2	3.65	9.21	0.1	17.7	464