



Phytoplankton community succession and biogeochemistry in a bloom simulation experiment at an estuary-ocean interface

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Abstract. Phytoplankton blooms, especially diatom blooms, account for a large fraction of marine carbon fixation. Species
10 succession and biogeochemical parameters change rapidly over a bloom, and determine the resulting biological productivity. This study implemented daily sampling of a 24-L microcosm bloom simulation experiment to assess changes in assemblage and biogeochemical processes while excluding changes due to advection. ¹⁵NO₃⁻ and H¹³CO₃⁻ tracer incubations were performed alongside pigment and DNA sampling to compare temporal trends in community composition and primary productivity (nitrogen (N) and carbon (C) transport rates). Rapid drawdown of nutrients and maximum C and N transport rates
15 corresponded with peak chlorophyll-a and fucoxanthin pigment concentrations. Fucoxanthin, typically associated with diatoms, was the dominant diagnostic pigment, with very low peridinin (dinoflagellate) and zeaxanthin (cyanobacteria) concentrations, indicating a diatom bloom. 18S rRNA gene analysis showed clear community succession throughout the duration of the bloom and multiple species of diatoms co-occurred, including during the bloom peak. The presence of metazoan 18S, high carbon-to-chlorophyll ratios, and a model analysis provide evidence of grazing in the latter half of the
20 bloom. A traditional bloom framework suggests that species succession occurs as the bloom progresses and that phytoplankton diversity reaches a minimum of just one or two dominant species when phytoplankton productivity is at its maximum. However, this study produced a negatively monotonic productivity-diversity relationship with relatively high minimum diversity values. This 18S-based analysis therefore presents a more complex relationship between bloom progression and phytoplankton diversity.

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Short summary

Concurrent sampling of environmental parameters, productivity rates, photopigments, and DNA were used to analyze a 24-L
estuarine diatom bloom microcosm. Biogeochemical data and an ecological model indicated that the bloom was terminated by
grazing. Comparisons to previous studies revealed (1) additional community and diversity complexity using 18S amplicon vs.
30 traditional pigment-based analyses, and (2) a potential global productivity-diversity relationship using 18S and carbon transport rates.



1 Introduction

Marine primary productivity is dominated by phytoplankton and accounts for nearly half of global net carbon fixation (Field et al., 1998). Phytoplankton blooms are of particular importance because their communities are often dominated by large
35 phytoplankton (e.g., diatoms) which can grow rapidly. Diatoms are unique phytoplankton not only because of their large size and high maximum growth rates, but also because they have silica shells which may reduce grazing pressure (Grønning and Kjørboe, 2020) and increase their sinking rate (Legendre and Le Fèvre, 1995), leading them to contribute significantly to marine carbon export from the surface ocean (Jin et al., 2006). Diatoms that do not escape grazing become important food
40 large zooplankton (Sommer et al., 2002) or directly to fish (Ryther, 1969; Van Der Lingen, 2002; Costa and Garrido, 2004). Much of the research on diatom blooms is focused on upwelling systems, but the shallow, high nutrient environment of estuaries also supports annual diatom blooms and large fisheries (Marshall et al., 2005; Harding et al., 2015; Bilkovic et al., 2019). Estuarine systems serve as a buffer between land and the open ocean, modulating the type and amount of carbon that is exported to continental shelves, regions which are responsible for high carbon burial and export to the deep ocean (Bauer et
45 al., 2013). Thus, understanding phytoplankton communities and productivity in estuarine systems is key to understanding global nutrient cycling.

The productivity–diversity relationship (PDR) is used to investigate complex mechanisms such as species interactions, coexistence, and overall drivers of community composition. However, the PDR of marine microbial communities varies greatly
50 between studies and is heavily dependent on study region and methods used for determining diversity and productivity (Smith, 2007; Graham and Duda, 2011). The canonical view of bloom diversity and succession is that initial high nutrient conditions allow fast–growing phytoplankton to outcompete other species, leading to a minimum community diversity when phytoplankton biomass or rates of productivity are at their peak (Irigoiien et al., 2004). Blooms are then terminated by a combination of bottom–up factors like nutrient depletion, and top–down factors such as grazing and viral and parasitic
55 infection, transitioning into a more diverse microbial community.

Chesapeake Bay is the largest estuary in the United States (Bilkovic et al., 2019) and is often used as a model system for a broader understanding of coastal environments. Many studies have documented the seasonal succession of the phytoplankton community in the bay, primarily using pigment– and microscopy–based analyses (Marshall et al., 2005; Adolf et al., 2006;
60 Harding et al., 2015). Such studies generally agree on the year–round importance of diatoms; diatoms dominate the phytoplankton community in the spring and fall while a more diverse phytoplankton community composed of diatoms, dinoflagellates, and cyanobacteria is present in the summer. Though there have been 18S–based studies in the tributaries of Chesapeake Bay, especially of harmful dinoflagellate bloom communities, there are very few reported metabarcoding studies of the eukaryotic community in the main stem of Chesapeake Bay (Cram et al., 2024; Wang et al., 2024). Molecular techniques



65 like metabarcoding can detect a much greater diversity of marine eukaryotes than microscopy or pigment analyses, including
species that are not possible to detect or distinguish visually (López-García et al., 2001; Massana and Pedrós-Alió, 2008; Xu
et al., 2023).

Here we compare the composition and PDR of an estuarine diatom bloom with both historical trends in Chesapeake Bay and
70 with patterns observed globally in the open ocean. We utilized daily sampling of 18S amplicons, pigment concentrations,
carbon (C) and nitrogen (N) transport rates, and particulate organic matter (POM) accumulation in carboy incubations to follow
a single community and its associated biological processes. Observations were complemented by a biogeochemical model to
address unmeasured nutrient pools and potential grazing-related processes. This study provides a deeper understanding of the
relationship between eukaryotic community and productivity in a bloom by (1) coupling direct measurements of primary
75 productivity rates with high resolution metabarcoding techniques for community analysis and (2) comparing metabarcoding
to historical methods of community analysis.

2 Methods

2.1 Bloom simulation and sampling

24-L microcosm bloom simulation experiments were carried out in August 2021 in the Chesapeake Bay aboard the R/V Hugh
80 Sharp. Surface water (2–5 m) was collected from the study site (37.27° N, 76.09° W), located near the mouth of the bay. All
materials used to set up and sample the microcosms were cleaned with 1 M HCl and rinsed with MilliQ water. Incubation
medium was prepared by pumping surface water (~5 m) directly from the sample site through a series of nylon mesh and glass
fiber filters, ending with a 0.3 μm filter, using a double diaphragm pump into three 24-L translucent polycarbonate (PC)
carboys. Surface water inoculum was collected using a rosette system with 12-L Niskin bottles and a CTD profiler from 2–4
85 m depth and pre-filtered through 210 μm nylon mesh before being added to the carboys. Filtered surface medium (21.6 L) and
pre-filtered surface water inoculum (2.4 L) were added to each carboy to produce a 10 % inoculation. Additional 0.3 μm
filtered medium and 210 μm pre-filtered surface water inoculum were collected for DNA analysis. Lastly, NaNO_3 , NaH_2PO_4 ,
and Na_2SiO_3 solutions were added to each carboy to achieve final concentrations of approximately 40 μM , 5 μM , and 50 μM ,
respectively. A nitrogen:silica:phosphorus ratio of ~10:10:1 is compatible with ambient nutrient ratios observed in lower
90 Chesapeake Bay (Fisher et al., 1992) and with diatom nutrient quotas (Brzezinski, 1985; Lomas et al., 2019). Carboys were
incubated for eight days in an on-deck water bath, using a seawater flow-through system drawn from surface water and a
plastic screen shade covering to keep incubation temperature and light similar to in situ conditions. Continuous light and
temperature measurements were recorded using two Onset HOBO Pendant Temperature/Light data loggers suspended ~10 cm
below the surface of the on-deck water bath.

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Samples for nutrient concentrations were collected from each carboy three times per day, at approximately 06:00, 12:00, and 18:00 local time. Samples were filtered through a 0.22 μm syringe filter into 50-mL plastic conical tubes and stored at -20°C until analysis.

100 Samples for pigment analysis were collected twice daily at 12:00 and 18:00 starting on day 2. Duplicate samples (100–400 mL) were filtered onto pre-combusted (500°C for ~ 5 h) 0.3 μm 25 mm GF-75 filters using pigment-dedicated filter-holders. Filters were stored individually at -80°C .

DNA samples were taken concurrently with the 12:00 pigment and nutrient samples. Duplicates were collected for select days from carboy C and two additional DNA samples were collected from the surface water inoculum. Sample water was collected from each carboy and filtered onto 0.22 μm Sterivex filters (320–1290 mL of sample water per filter) using a peristaltic pump. 105 Filters were flash-frozen in liquid nitrogen and stored at -80°C until DNA extraction. For more detail on sampling protocols, see **Supporting Information, Methodology**.

110 2.2 Nutrient and pigment concentrations

Reactive nitrite (NO_2^-) and silicate (silicic acid, H_4SiO_4) concentrations were measured on a ThermoScience Genesys150 UV-Vis spectrophotometer using colorimetric methods (sulfanilamide + *N*-(1-naphthyl)-ethylenediamine and metol sulfite, respectively) modified from Strickland and Parsons (1972). $\text{NO}_2^- + \text{NO}_3^-$ concentration was measured via chemiluminescent detection using a Teledyne NOx analyzer (NOxBox) according to Braman & Hendrix (1989). NO_3^- concentrations were 115 calculated by subtracting the colorimetrically derived NO_2^- concentrations from their paired $\text{NO}_2^- + \text{NO}_3^-$ NOxBox concentrations. Detection limits and measurement precision are listed in **Table S1**.

Pigment samples were analyzed by High Performance Liquid Chromatography (HPLC) (Pinckney et al., 1996, 2001). Effective detection limits are listed in **Table S2** and detailed methodology is available in Hooker et al. (2010).

120 2.3 ^{15}N and ^{13}C labeled sub-incubations and isotopic analysis for nutrient uptake measurements

Sub-incubations to measure nitrogen and carbon uptake rates using $^{15}\text{N}\text{-NO}_3^-$ and $^{13}\text{C}\text{-HCO}_3^-$ were carried out once per day. Sample water (150–200 mL) was aliquoted into PC bottles for triplicate sub-incubations for each carboy (9 sub-incubations per tracer per day). $^{15}\text{N}\text{-NaNO}_3^-$ (1.5–2.67 mL of a 0.3 mM solution) and $^{13}\text{C}\text{-NaHCO}_3^-$ (1–1.3 mL of a 30 mM solution) were added to each PC bottle to attain isotopic enrichments of $\sim 8\text{--}93\%$ and $\sim 10\%$, respectively, and gently inverted 5 times to mix. 125 Bottles were then placed into mesh bags to simulate surface water light intensities and incubated for 4 h (approximately 10:00 to 14:00 local time) in a similar water bath system as the carboys. Sub-incubations were terminated by filtration onto pre-



combusted 0.3 μm 25 mm GF-75 filters using label-specific filter-holders, followed by a 2.5 mL 0.3 μm -filtered seawater rinse (from original medium collection). Filters were stored individually at -20° C until analysis.

2.4 POM concentrations and uptake rate measurements

130 POM filters were fume acidified for 4–6 h in a desiccator with concentrated HCl, packed into tin capsules, and measured using a Sercon ANCA-SL Elemental Analyzer and a Europa 20/20 Isotope Ratio Mass Spectrometer (EA-IRMS). Standards were prepared from a stock solution of urea and measurements were calibrated using an aminocaproic acid (ACA) standard. All detection limits and measurement precisions are listed in **Table S1**. Transport rates (ρ) were calculated according to equations modified from Dugdale and Goering (1967):

$$135 \quad \text{At}\%C = \frac{\text{mol } ^{13}\text{C}}{\text{mol } ^{13}\text{C} + \text{mol } ^{12}\text{C}} \times 100 \% \quad (1)$$

$$\text{At}\%N = \frac{\text{mol } ^{15}\text{N}}{\text{mol } ^{15}\text{N} + \text{mol } ^{14}\text{N}} \times 100 \% \quad (2)$$

$$C \text{ transport } [\mu\text{M C day}^{-1}] = \rho_{\text{HCO}_3^-} = \frac{\text{At}\% \text{ POC}_{\text{exp}} - \text{At}\% \text{ POC}_{\text{baseline}}}{\text{At}\% \text{ DIC}_{\text{exp}} - \text{At}\% \text{ DIC}_{\text{natural}}} \times \frac{[\text{POC}]}{\text{hr}} \times \frac{12 \text{ light hrs}}{\text{day}} \quad (3)$$

$$N \text{ transport } [\mu\text{M N day}^{-1}] = \rho_{\text{NO}_3^-} = \frac{\text{At}\% \text{ PON}_{\text{exp}} - \text{At}\% \text{ PON}_{\text{baseline}}}{\text{At}\% \text{ DIN}_{\text{exp}} - \text{At}\% \text{ N}_{\text{air}}} \times \frac{[\text{PON}]}{\text{hr}} \times \frac{12 \text{ light hrs}}{\text{day}} \quad (4)$$

140 At% C and At% N are the atom percent for carbon and nitrogen terms and [POC] and [PON] are in units of $\mu\text{M C}$ and $\mu\text{M N}$, respectively. Because ^{13}C and ^{15}N labeled sub-incubations were done separately and concurrently each day, it was possible to compare paired labeled-unlabeled incubations, i.e., one set of sub-incubations received either ^{13}C or ^{15}N and the other did not. Therefore, $\text{At}\% \text{ POM}_{\text{exp}} - \text{At}\% \text{ POM}_{\text{baseline}}$ was used instead of $\text{At}\% \text{ POM}_{\text{final}} - \text{At}\% \text{ POM}_{\text{initial}}$ for transport rate calculations. At% POC_{exp} is the “experimental” At% POC (i.e., the At% POC of the incubations enriched with ^{13}C tracer) and At% POC_{baseline} is the At% POC of the incubations which received ^{15}N tracer instead of ^{13}C . The analogous comparison was used for At% PON_{exp} and At% PON_{baseline}. A constant DIC concentration of 1.8 mM was assumed according to the relationship between DIC concentration and salinity in Chesapeake Bay (Lee et al., 2015) and an average natural abundance ^{13}C of -25 ‰ (PDB) was used for all C-uptake calculations (Spiker, 1980; Guo et al., 1996). A natural abundance of 0.365 ‰ was used for At% N_{air} and assumed to be the same in the unlabeled DIN. Uptake per hour was converted to uptake per day assuming 12 light hours per day. Specific uptake rates (V) were calculated as:

$$V_{\text{HCO}_3^-} [\text{day}^{-1}] = \frac{\rho_{\text{HCO}_3^-}}{[\text{POC}]} \quad (5)$$

$$V_{\text{NO}_3^-} [\text{day}^{-1}] = \frac{\rho_{\text{NO}_3^-}}{[\text{PON}]} \quad (6)$$



2.5 Nutrient–Phytoplankton–Zooplankton (NPZ) model

155 A detailed description of the model is available in the **Supporting Information, Methodology**. Briefly, a simplified version of the biogeochemical model Carbon, Ocean Biogeochemistry and Lower Trophics version 2 (COBALTv2, Stock et al. 2020) was implemented to investigate potential grazing rates and grazing–associated changes in biogeochemical parameters during the bloom simulation incubations. This model includes a single nutrient (nitrate), and biological components are parameterized based on a large open ocean dataset to represent a single phytoplankton community (diatoms) and a single zooplankton community (medium–sized copepods).

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The model assumes a mass balance of five nitrogen pools: dissolved inorganic nitrogen (DIN) in the form of nitrate (N_{NO_3}), particulate organic nitrogen (PON) in the forms of phytoplankton (N_{Phyto}), zooplankton (N_{Zoo}), and detritus ($N_{Detritus}$), and dissolved organic nitrogen (DON; N_{DOM}). The model simulates the temporal evolution of the system assuming no spatial variations. The initial nitrogen pools were determined from the observed day 1 nitrate concentration ($[NO_3^-]_{avg} \approx 45 \mu mol kg^{-1}$), an estimation of day 1 phytoplankton biomass ($5 \mu mol kg^{-1}$) based on day 2 Chl–a concentrations, and an assumed average 1:100 biomass ratio of zooplankton to phytoplankton. The Redfield ratio of 106C:16N (Redfield, 1934) was used to convert PON and DON into their respective carbon (C) pools and a $C_{Phyto}:Chl\text{--}a$ ratio of 200 was used to convert between POC and Chl–a concentrations. Sensitivity testing was performed by independently and randomly varying the initial N_{Phyto} , N_{Zoo} , and N_{NO_3} by $\pm 20\%$ across 1000 model iterations.

170 2.6 DNA extraction and 18S V4 sequencing

DNA extraction and cleanup was performed using the ZymoBIOMICS DNA/RNA miniprep kit (Zymo Research, Irvine CA). Universal primer set 515F (Caporaso et al., 2011) and 951R (Mangot et al., 2013; Lepère et al., 2016) were used for PCR to amplify the V4 region of the 18S rRNA gene. (See **Supporting Information, Methodology** and **Table S3** for PCR details). The purified amplicons were indexed using an Illumina indexing kit, pooled, and sequenced for paired–end 2x250 bp sequences on an Illumina MiSeq platform at the Princeton University Genomics Core Facility.

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2.7 18S sequence clustering and quality control

Demultiplexed sequences were trimmed using prinseq v0.20.4 (Schmieder and Edwards, 2011) to remove primers and low quality read ends. Paired–end reads were merged using the Illumina utilities package v2.6 (Eren et al., 2013) with default parameters. Merged pairs were discarded if more than one third of the first half of either read had a Q–score below Q30, in accordance with Minoche et al. (2011), or if Q–score fell below Q15 for any base in the overlapped region. Merged reads were then dereplicated and potential chimeras were identified using uchime in vsearch v2.10.4 (Edgar et al., 2011; Rognes et al., 2016). Reads were clustered into operational taxonomic units (OTUs) using swarm v3 with a neighbor nucleotide difference of 1 ($d=1$) (Mahé et al., 2021). Taxonomy was assigned using usearch_global (vsearch v2.10.4) and reference database PR²

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v4.14.0 (Guillou et al., 2012). OTUs were removed if: (1) the representative sequence had less than 10 reads, (2) the OTU
185 represented < 0.01 % of total reads in all samples, (3) > 50 % of reads within the OTU were identified as potentially chimeric,
or (4) the OTU was < 90 % identical to the associated PR² reference sequence, resulting in 335,896 reads which passed quality
checks. Lastly, OTUs with a < 97 % match to a PR² reference sequence were re-classified as “Other [Class]” and combined
according to class. E.g., the reads from all OTUs with a 90 to < 97% match to a Bacillariophyta sequence in the PR² database
were combined into one OTU classified as “Other Bacillariophyta.” All other OTUs with a ≥ 97% match to a reference were
190 combined if their associated PR² reference sequence was the same, resulting in 150 total OTUs.

2.8 Relative abundance, alpha, and beta diversity analyses

OTUs remaining after quality control were further manipulated for downstream analysis. For relative abundance and alpha
diversity analyses, carboy C duplicates were combined by day and the two inoculum samples were combined and used as “Day
0” for all carboys. Lastly, all OTUs identified by PR² as metazoan (4 OTUs after combination; 42.4 % of all sequences) and
195 bacterial (7 OTUs after combination; 10.4 % of all sequences) were removed for all relative abundance and diversity analyses.

Diversity analysis was performed using the phyloseq v1.44.0 (McMurdie and Holmes, 2013) and microViz v0.12.1 (Barnett
et al., 2021) R packages. The centered log–ratio (clr) transformed Aitchison distance was used to calculate beta diversity. A
Canonical Analysis of Principal coordinates (CAP) plot was used to compare beta diversity with observations of environmental
200 parameters using the vegan v2.6-6 package in R (Oksanen et al., 2024).

2.9 Statistical analysis

Statistical analyses were performed using R v4.3.1 (R Core Team, 2023) and the MATLAB Statistic and Machine Learning
Toolbox v24.1 (MathWorks Inc., 2024). In order to assess the significance of the daily variability of rate measurements and
alpha diversity, as well as the linearity of PDR curves, the car v3.1.3 (Fox and Weisberg, 2019) package in R was used to
205 check if residuals were normally distributed using both a Shapiro–Wilk test (significant p–value < 0.01) and a visual inspection
of a quantile–quantile plot. Normally distributed data were compared using Pearson correlation or a one–way ANOVA with a
Tukey HSD family–wise comparison, and non–normally distributed data were compared using Spearman correlation or a
Kruskal–Wallis test with Bonferroni adjusted pairwise comparisons (significant p–value < 0.05). Beta diversity clustering was
assessed using homogeneity of dispersion and PERMANOVA tests (significant p–value < 0.05).

210 3 Results

3.1 Biogeochemical signatures indicate a diatom bloom

Temporal patterns of particulate organic matter (POM), ambient nutrient, and pigment concentrations in all three carboys
indicated a phytoplankton bloom that peaked on day 5 (**Fig. 1, S1**). Particulate organic carbon (POC) and nitrogen (PON)



remained low in all carboys until day 4 (avg. $[POC_{min}] = 22.8 \pm 4.5 \mu M$ and avg. $[PON_{min}] = 1.8 \pm 1.3 \mu M$) (**Fig. 1a, S1a–c**).
215 POM then rapidly increased between days 4 and 5, remaining high between days 5 and 7 and reaching average maximum
concentrations of $232.0 \pm 18.1 \mu M$ ($[POC_{max}]$) and $24.5 \pm 4.3 \mu M$ ($[PON_{max}]$).

The increases in POM through day 5 were mirrored by nutrient drawdown, with all three carboys following similar temporal
trends. Nitrate (NO_3^-) and silicate (SiO_4^{4-}) were supplemented initially to average starting concentrations of $42.6 \pm 1.0 \mu M$ and
220 $46.9 \pm 1.0 \mu M$, respectively, and remained high until day 4 in all carboys (**Fig. 1b, S1d–f**). Between noon day 4 and 18:00 day
5, $[NO_3^-]$ and $[SiO_4^{4-}]$ were rapidly drawn down to $1.1 \pm 0.2 \mu M$ and $1.0 \pm 0.2 \mu M$, respectively, and remained low for the
remainder of the incubations. Phosphate (PO_4^{3-}) concentrations followed a similar pattern (**Fig. S2**); $[PO_4^{3-}]$ was supplemented
to an initial average of $4.0 \pm 0.4 \mu M$ and gradually decreased to $0.6 \pm 0.1 \mu M$ by 18:00 day 5. Nitrite (NO_2^-) remained low
throughout the bloom, increasing slightly on day 4 to an average $[NO_2^-]_{max}$ of $1.2 \pm 0.06 \mu M$ on day 5, possibly due to
225 incomplete NO_3^- assimilation by phytoplankton (Collos, 1998; Lomas and Lipschultz, 2006). The small $[NO_2^-]$ peak was then
quickly depleted following the exhaustion of $[NO_3^-]$. A mass balance of the measured nitrogen species revealed that total
measured nitrogen ($[NO_2^-] + [NO_3^-] + PON$) was $\sim 45\text{--}50 \mu M$ leading up to the bloom peak but was $< 30 \mu M$ afterwards. This
“gap” in the total measured N may be attributable to unmeasured N pools, such as ammonium or dissolved organic nitrogen
(DON) and is addressed in the NPZ model section below (see also **Fig. S3g–i**).

230 Chlorophyll a (Chl–a) concentrations also reached their maximum at noon on day 5 for all carboys (avg. $Chl\text{--}a_{max} = 15.0 \pm 2.1$
 $\mu g L^{-1}$) (**Fig. 1c, S1g–i**). Concentrations of fucoxanthin, a diagnostic diatom pigment, followed the same pattern as Chl–a,
while peridinin (dinoflagellates) and zeaxanthin (cyanobacteria) remained low, and divinyl chlorophyll a (cyanobacteria)
remained below detection throughout the bloom. Chl–a and fucoxanthin concentrations both decreased following day 5,
235 signifying the decline of the phytoplankton population. Pigment accumulation trends along with the concurrent consumption
of SiO_4^{4-} indicate that a diatom bloom occurred. Additionally, the decoupling between Chl–a and POM concentrations in the
late bloom resulted in a large range of POC:Chl–a ratios. POC:Chl–a reached a minimum during the peak bloom with an
average value of 58.6 ± 8.2 , which is within the typical range of $\sim 10\text{--}200$ for large phytoplankton (e.g.; Laws and Bannister
1980; Schoemann et al. 2005 and references therein). Late bloom ratios were much higher, reaching an average POC:Chl–a $>$
240 1000 on day 7, with maximum values of up to 1258.6 ± 247.2 in carboy B (**Fig. S3d–f**), suggesting the accumulation of non-
phytoplankton biomass in the late bloom.

3.2 Temporal patterns of nutrient transport and specific uptake rates

Temporal patterns of absolute C and N transport rates ($\rho_{HCO_3^-}$, $\rho_{NO_3^-}$) were consistent across all carboys. Transport rates
remained low prior to day 4 (**Fig. 2**). They then increased slightly on day 4 and reached maximum rates on day 5 with average
245 rates for the three carboys of $186.0 \pm 64.7 \mu M C d^{-1}$ and $27.1 \pm 8.95 \mu M N d^{-1}$, before quickly dropping again. Days 6 and 7
had transport rates similar to those measured on day 4. The timing of maximum transport rates coincided with the day 5 peak



observed in POM and Chl-*a* concentrations and the highest rates of nutrient depletion. Despite similar patterns, day 5 $\rho_{\text{HCO}_3^-}$ varied between carboys. Peak $\rho_{\text{HCO}_3^-}$ was highest in Carboy C ($229.3 \pm 80.6 \mu\text{M C d}^{-1}$) and lowest in Carboy A ($141.6 \pm 31.8 \mu\text{M C d}^{-1}$) (**Table S4**). However, peak $\rho_{\text{HCO}_3^-}$ were not statistically different between carboys (one-way ANOVA, $p = 0.28$).
250 $\rho_{\text{NO}_3^-}$ was less variable between carboys than between triplicate measurements within a single carboy (**Table S4**).

Biomass specific uptake rates for both C and N ($V_{\text{HCO}_3^-}$, $V_{\text{NO}_3^-}$) rapidly increased one day prior to the increase in absolute transport rates and remained high until nutrients were depleted (**Fig. 2**). $V_{\text{HCO}_3^-}$ peaked on day 4 with average rate for all three carboys of $1.0 \pm 0.06 \text{ day}^{-1}$, and $V_{\text{NO}_3^-}$ peaked on day 5 with an average rate of $1.1 \pm 0.12 \text{ day}^{-1}$ (**Table S4, Fig. S4**).

255 3.3 NPZ model supports late bloom grazing

A nutrient-phytoplankton-zooplankton (NPZ) model was employed to investigate the “gap” in total measured N and the potential contribution of unmeasured biogeochemical parameters. The model was able to accurately simulate the observed magnitude and temporal evolution of $[\text{NO}_3^-]$, Chl-*a*, and POM when including grazing by zooplankton, with the best fit to $[\text{NO}_3^-]$ (**Fig. 3a-d**). The model slightly overestimated POM during the early bloom, but matched observations from the peak bloom onward. The modeled grazing rate reached an average maximum of $27.3 \mu\text{M N d}^{-1}$ and a combined non-phytoplankton PON ($N_{\text{Zoo}} + N_{\text{Detritus}}$) which represented 67.5 % of total modeled PON at the end of the bloom. Modeled grazing and respiration rates resulted in a $\sim 20 \mu\text{M N}$ accumulation of DON on day 7, which is comparable to the 20–30 $\mu\text{M N}$ “gap” in the observed N mass balance (**Fig. S3g-i**).

3.4 Community succession resulted in distinct temporally varying assemblages

265 Changes in the eukaryotic community composition were detected during the bloom using amplicon sequencing of the hypervariable V4 region of the 18S rRNA gene. Swarm clustering (Mahé et al., 2021) and quality control resulted in 143 eukaryotic OTUs. While Metazoa such as copepods were present in the 18S sequences (**Fig. S5**) and a potentially important portion of eukaryotic community, the multicellular nature of Metazoa made it difficult to interpret their relative abundances. Therefore, they were removed from community analyses, leaving 139 non-metazoan eukaryotic OTUs for further analysis.

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OTU relative abundances in all three carboys showed succession of the bloom community, with the development of three distinct, temporally driven assemblages: early- (days 2 and 3), mid- (days 4 and 5) and late-bloom (days 6 and 7) (metadata summarized in **Table S5**). Community composition in early-bloom samples was similar to inoculum samples, with the exception of parasitic Syndiniales OTUs. Syndiniales OTUs had a high relative abundance in inoculum samples but remained low in carboy samples (**Fig. 4a-c**). Most early bloom sequences were from the green algae Mamiellophyceae, diatoms (Bacillariophyta), dinoflagellates (Dinophyceae), unclassified marine Stramenopiles (MAST-1), and heterotrophic Spirotrichea OTUs. By the mid- and late-bloom, relative abundances indicated that the community was composed primarily of diatoms and dinoflagellates, though the ratio of diatom and dinoflagellate relative abundances varied between carboys. At



the class level, the mid- and late-bloom assemblages are distinguished by a late-bloom increase in the relative abundance of
280 heterotrophic organisms like Filosa-Thecofilosea and, in Carboy C, Spirotrichea OTUs. Their increase is consistent with
grazing and an accumulation of non-phytoplankton material in the late-bloom.

The early-, mid-, and late-bloom community shifts result in three distinct clusters in the beta diversity analysis
(PERMANOVA, $p = 0.001$, $R^2 = 0.60$) (**Fig. 4g**). Early-bloom samples cluster with inoculum samples and late-bloom samples
285 cluster together regardless of carboy. Mid-bloom samples show the greatest variability between both carboys and day, except
for carboy C, which had relatively similar communities on days 4 and 5. Additionally, variability between biological duplicates
(inoculum and carboy C, days 3, 5 and 6) is of the same magnitude as variability between samples from different carboys on
the same day (**Fig. 4g**). Both day and $[\text{NO}_3^-]$ vary along CAP1. Therefore, CAP1, which explains 42.2 % of community
composition, likely corresponds to a combination of time since inoculation and nutrient availability.

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The alpha diversity of the non-metazoan eukaryotic community peaked during the early-bloom, decreased during the mid-
bloom, and did not change between the mid- and late-bloom. This pattern was most prominent in the OTU richness (number
of OTUs), which separated into two statistically distinct groups: the high diversity early-bloom and the low diversity mid/late-
bloom (**Fig. S6a**). Conversely, the Shannon index (H) did not significantly vary throughout the bloom despite sharing the same
295 temporal trend (**Fig. 4d-f, S6b**). H ranged from 0.5–3.4 with an average of 2.5. While H decreased slightly in carboys A and
C during the bloom peak, it did not drop below 2. The mid-bloom decrease in H was likely driven by carboy B's ~40 % decline
in H from an early-bloom average of 3.02 to a mid/late-bloom average of 1.24. Overall, while the alpha diversity did decrease
leading up to the bloom peak, the peak was not characterized by a significant or consistent diversity minimum.

3.5 High diversity and succession of the diatom bloom assemblage

300 A more comprehensive analysis of the diatom assemblage was performed because of their prominence among the
phytoplankton community and because the biogeochemical data indicated their dominance during the peak bloom. The relative
abundance of diatom sequences generally increased through the early-bloom, reached a maximum of up to ~60 % (in carboy
C) during the mid-bloom, and then decreased during the late-bloom. The relative abundance of diatoms in carboy B peaked
slightly earlier and had lower relative abundances than the other carboys during the mid- and late-bloom, except for a day 6
305 peak of an araphid pennate diatom OTU (**Fig. 5a-c**) identified as a *Thalassionema sp.* via BLAST search. Succession was also
evident in the composition of the diatom assemblage. The highest relative abundance diatom sequences were all cosmopolitan;
Chaetoceros dominated the early-bloom, *Thalassiosira* the mid-bloom, and *Thalassionema* the late-bloom. *Ditylum*
sequences were also highly abundant, especially just preceding the peak bloom. The relative abundances of most diatom genera
were dominated by a single OTU, except for *Chaetoceros*, which had > 10 OTUs that were relatively evenly represented.

310



Diatom-specific alpha diversity was more consistent between carboys and displayed clearer temporal trends compared to overall community alpha diversity. Diatom alpha diversity increased during the early-bloom and decreased during the mid-bloom, with variable trends during the late-bloom. However, diatom richness never dropped below 17 in any carboy (**Fig. 5g–i**) and dozens of diatom OTUs, spanning multiple genera, were present at the peak of the bloom.

315 4 Discussion

4.1 Diatom-driven peak bloom patterns

Maximum nutrient transport rates, Chl-a accumulation, and nutrient drawdown all occurred on day 5 and indicate that a diatom bloom occurred (**Fig. 6**). While seasonal succession of the phytoplankton community is well documented in Chesapeake Bay and predicts a highly diverse community of diatoms, dinoflagellates, and cyanobacteria in the summer (Adolf et al., 2006; 320 Harding et al., 2015; Marshall et al., 2005), the pigment data did not indicate that dinoflagellates or cyanobacteria contributed significantly to the biomass. Light patterns could not explain the trend in POC:Chl-a observed in the carboy incubations (**Fig. S7**). Instead, high nutrient conditions (Laws and Bannister, 1980; Schoemann et al., 2005) and diatom-dominated biomass (Tada et al., 2000; Sathyendranath et al., 2009) resulted in low total community POC:Chl-a during the mid-bloom.

325 The observed maximum C and N transport rates in this study were higher than previously reported summer rates in the lower bay (e.g.; Sellner 1983; Glibert et al. 1995; Marshall and Nesiuis 1996; Bradley et al. 2010). $\rho_{\text{NO}_3^-}$ in particular was often higher than previous observations of total dissolved N transport (ρ_{TDN} , the sum of uptake of all forms of N) (e.g.; Glibert et al. 1995; Bradley et al. 2010). Similarly, maximum Chl-a-normalized C transport ($V_{\text{C_Chla}}$, **Fig. S8**) and $V_{\text{NO}_3^-}$ rates were also higher than previous reports (Adolf et al., 2006; Bradley et al., 2010) (additional information in the **Supporting Information,** 330 **Analysis**). High nutrient transport rates and a stronger NO_3^- preference are all associated with diatom blooms, especially in the lower bay (Glibert et al., 1995; Marshall and Nesiuis, 1996; Sellner, 1983). Additionally, high summer $V_{\text{C_Chla}}$ is positively correlated to the fraction of Chl-a attributable to diatoms (Adolf et al., 2006) and $V_{\text{NO}_3^-}$ tends to be higher in larger ($> 35\mu\text{m}$) size fractions than total community (Bradley et al., 2010).

335 Although specific uptake rates were high compared to previous Chesapeake Bay measurements, they were similar to those previously observed in a bloom mesocosm (Fawcett and Ward, 2011) and the open ocean (Van Oostende et al., 2017). $V_{\text{HCO}_3^-}$ and $V_{\text{NO}_3^-}$ peaked on different days, but both displayed a significant (Kruskal-Wallis, $p = 6.17 \times 10^{-4}$ and $p = 6.02 \times 10^{-3}$, respectively) increase between days 3 and 4, one day prior to the bloom peak (**Fig S4**). The observed increase in specific uptake rates prior to absolute transport rates is consistent with evidence that blooming phytoplankton (e.g., diatoms) are able to exploit 340 nutrient-rich conditions by both increasing and maintaining specific uptake rates. This ability of diatoms to shift-up their nutrient uptake in response to a sudden nutrient addition is a well-documented phenomenon (e.g.; Kudela and Dugdale 2000; Fawcett and Ward 2011; Lampe et al. 2018), including in Chesapeake Bay (Malone et al., 1996).



Both absolute transport and specific uptake rates may be overestimated after day 5, as ambient $[\text{NO}_3^-]$ became depleted.
345 However, it is unlikely that temporal patterns observed in the transport or specific uptake rates were altered by these factors
(additional information in the **Supporting Information, Analysis and Table S6**).

The depletion of nutrients by noon on day 5 and high nutrient requirements for large blooming phytoplankton suggests that
the diatom community was unable to sustain growth, leading to the bloom's demise following day 5. We observed a strong
350 coupling between NO_3^- and SiO_4^{4-} drawdown rates (**Fig. S9c**), which matched average diatom N:Si quotas, between days 2–5
(Brzezinski, 1985; Fisher et al., 1992). However, following the combined kinetics- and stoichiometry-based thresholds
outlined in Liang et al. (2019), no nutrients were consistently limiting; there were only two timepoints when nutrients were
limited in more than one carboy: SiO_4^{4-} on the evening of day 5 and NO_3^- on the evening of day 6 (**Table S7**). Therefore,
factors other than nutrient availability likely contributed to the bloom decline; the potential role of grazing is further
355 investigated below.

4.2 Late-bloom patterns were driven by grazing

In addition to or instead of nutrient limitation, the bloom decline may have been due to top-down factors such as grazing,
parasitism, and viral infection. Biogeochemical measurements, an NPZ model, and 18S community analysis provide evidence
for grazing.
360

In contrast to the low peak bloom POC:Chl-a, the extreme POC:Chl-a values observed at the end of the carboy incubations
far exceeded the average phytoplankton POC:Chl-a of ~40–90 observed in Chesapeake Bay (Cerc0 2000 and the references
therein), with average day 7 values over 10-fold greater than expected. High total community POC:Chl-a is primarily
influenced by the ratio of phytoplankton POC to zooplankton and detrital POC (Banse 1977). Even though the inoculum was
365 prefiltered using a 210 μm mesh, metazoan sequences were present in every DNA sample. Therefore, it is likely that smaller
Metazoa passed through the initial filter and grew in response to the bloom. Grazing would also contribute to unmeasured N
pools, as DON would be released during grazing or parasitic and viral lysis (Bronk 2002; Park et al. 2004). Given the timing
and preparation of the experiments, as well as the absence of Chlorophyllide a (a Chl-a degradation product) in all pigment
samples, phytoplankton detritus was unlikely to have contributed significantly to POM.

370 To further investigate potential role of copepods, we used an NPZ model to simulate the transfer of N to different pools during
the bloom, mediated by diatom and copepod processes. The model outputs for NO_3^- , Chl-a, and POM matched observations
(**Fig. 3a–d**), providing support for the application of the model as a proxy for unmeasured N pools, such as DON, and fluxes
between N pools. The NPZ model, despite being a simplification of the full community dynamics, showed that grazer–
375 phytoplankton interactions could explain the majority of observed biogeochemical patterns in the carboy bloom simulations



given the initial nutrient conditions. Transfer of PON from phytoplankton into zooplankton biomass and detritus match and explain the observed high PON at the end of the bloom despite a decrease in chlorophyll, and therefore phytoplankton biomass. Additionally, DON released as a direct result of grazing in the model can account for a significant portion of the observed N mass balance “gap” (**Fig. 3f**).

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Simplifications in the model prevented over-parameterization, but led to slight deviations between model outputs and observations (additional information in the **Supporting Information, Analysis**). Despite these minor deviations, the consensus between observations and a simplified NPZ model demonstrates that even a diverse diatom assemblage acts in accordance with globally averaged diatom behavior under blooming conditions. Furthermore, the modeled grazing contrasts with previous studies which found a decoupling of diatoms and copepods in the open ocean (Lima–Mendez et al. 2015) and suggests that the diatom community in a shallow, coastal environment can be controlled both by bottom–up and top–down factors.

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4.3 Replication of bloom timing and community between carboys

Patterns of 18S–based community composition were similar across carboys. Beta diversity analysis showed that between–carboy dissimilarity was of comparable magnitude to the dissimilarity between biological replicates for carboy C (**Fig. 4g**), except on day 4. Day 4 is a transitional period between the early–bloom and peak bloom assemblages; all three carboys appear to have been trending toward a similar peak bloom assemblage, but rapid community succession and slight variability in the exact timing of the peak may have caused the observed divergence between assemblages. Carboy C showed a similar assemblage on days 4 and 5, implying that the peak bloom community was present slightly earlier than in other carboys, concurrent with the uptick in specific uptake rates and decrease in alpha diversity (**Fig. 6**). This indicates that the community being established was responsible for the observed shift in nutrient acquisition strategy. The variability in the timing of the bloom peak may be due to minor differences in the starting community that each carboy received, as seen in the dissimilarity present between replicate inoculum samples despite being filtered from the same stock of water. Regardless of community dissimilarities between inoculum duplicates or between carboys during the mid–bloom, the early– and late–bloom samples clustered very closely by day. This may indicate that so long as inoculum volume and percentage are high enough to ensure the presence of the same potential blooming organisms, carboy experiments can be replicable and good approximations of the in situ environment.

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4.4 Disentangling the bloom community: pigment– vs. 18S–based analyses and the role of dinoflagellates

Very few studies have performed 18S–based analyses in the main stem of the Chesapeake Bay (Cram et al. 2024; Wang et al. 2024), but a comparison of microscopy– and 18S–based community analysis revealed a strong correlation between cell counts and relative abundance of major phytoplankton groups (Wang et al. 2024). However, in the current study, the high

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dinoflagellate 18S relative abundances in the mid- and late-bloom (**Fig. 4a–c**) contradict the diatom-dominated story suggested by the biogeochemical data and thus required additional inspection.

410 Diatoms often dominate the Chesapeake Bay phytoplankton community, but many studies have documented summer–fall harmful dinoflagellate blooms in the bay (e.g.; Li et al. 2015 and references therein). A large portion of the dinoflagellate sequences in this study belonged to *Karenia mikimotoi* (**Fig. S10**) (previously *Gymnodinium mikimotoi*, *Gymnodinium* cf. *nagasakiense*, and *Gyrodinium aureolum*) (Hansen et al. 2000; Li et al. 2019). *Karenia* spp. are globally distributed dinoflagellates associated with harmful algal blooms (e.g.; Glibert et al. 2014; Li et al. 2019). A key feature of *K. mikimotoi* is
415 that it contains fucoxanthin instead of the typical dinoflagellate diagnostic pigment peridinin, making it difficult to distinguish between *K. mikimotoi* and diatoms based on fucoxanthin vs. peridinin alone (Richardson and Pinckney 2004; Li et al. 2019). However, other *Karenia*-associated accessory pigments were absent during the mid- and late-bloom (**Fig. S11**) and *K. mikimotoi*'s growth strategies make it most suited to lower nutrient environments compared to this study (additional information in the **Supporting Information, Analysis**). Therefore, it is unlikely that *K. mikimotoi* was a consequential
420 component of the biomass. Instead, we suggest that the apparent high relative abundance of dinoflagellates is an issue of 18S gene copy number.

The 18S rRNA gene copy number (GCN) can vary between species and individuals, leading to bias in relative abundance data (Guo et al. 2016; Martin et al. 2022). Both diatom and dinoflagellate 18S GCNs can range over several orders of magnitude,
425 but dinoflagellates on average have almost 30 times as many copies as diatoms (Martin et al. 2022). Corrections to relative abundance data based on GCN studies can improve the accuracy and inform the interpretation of 18S amplicon data, but implementation is limited due to the low number of (and variability between) GCN studies. Even though a previous Chesapeake Bay study showed that relative abundances were correlated to cell counts at a high taxonomic level (Wang et al. 2024), relative abundances impacted by GCN are not necessarily indicative of biomass or community relevance. This study exemplifies the
430 pitfalls of relying solely on 18S relative abundance and reinforces that multiple types of data should be implemented to assess community dynamics.

4.5 18s-based analyses reveal that high diversity is maintained during the bloom

The average 18S-based H for the whole community (2.5) was lower than previous studies in Chesapeake Bay, which found H values of 4–5 (Cram et al. 2024; Wang et al. 2024), but not as low as expected of a bloom. The canonical phytoplankton productivity–diversity relationship (PDR) is characterized by a unimodal function with maximum diversity found in
435 intermediately productive communities (indicated by cell abundances, biomass, or productivity rates). Minimum diversity, as low as one dominant species or $H < 0.1$, is found in high productivity phytoplankton communities (e.g.; Irigoien et al. 2004; Vallina et al. 2014). A phytoplankton bloom is expected to follow the same patterns, with lowest diversity during the peak of the bloom and concurrent with maximum productivity.



440

While the experiments in this study did reach minimum alpha diversity during or near the peak of the bloom (**Fig. 6, S6**), neither the OTU richness nor H dropped as low as expected according to the traditional bloom framework. Higher than expected OTU richness during the bloom peak was largely due to molecular techniques' ability to detect a larger portion of the community at greater taxonomic resolution than microscopy (López-García et al. 2001; Massana and Pedrós-Alíó 2008; Xu et al. 2023). Whole community H did not vary significantly between days, partly because H and other diversity measures which account for evenness tend to weaken the unimodal PDR (Skácelová and Lepš 2014) and make it difficult to observe a traditional diversity minimum during maximum productivity. Neither H nor OTU richness were correlated with sequencing depth (Pearson, $p = 0.97$ and $p = 0.62$, respectively); however, H values can be heavily impacted by the variability of 18S rRNA GCN. H accounts for the community evenness by using the relative abundance of OTUs, which is biased due to the orders of magnitude difference in 18S rRNA GCN between taxonomic groups in a whole community analysis.

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Even when comparing multiple microcosm and mesocosm studies, both the magnitude and temporal trends of whole community H were inconsistent (**Fig. S6c**). However, clearer temporal patterns in alpha diversity, which more closely reflected the canonical unimodal PDR, were revealed when isolating an individual taxonomic group. Diatom-specific H in this study increased rapidly over the first few days of the bloom and decreased during the mid-bloom (**Fig. 5d-i, 6a**). While late-bloom diversity in varied carboy-to-carboy, the overall trend matched the decrease observed in other bloom simulation experiments (**Fig. S6d**). The classic peak bloom diversity minimum may be more easily observable in the diatom community because the unimodal PDR has typically been used to describe phytoplankton communities, rather than whole communities, when applied to microbes. Diatom-specific H values were also much closer to previous observations, regardless of whether H was determined by microscopy or 18S, indicating that alpha diversity is directly comparable between these methods so long as it is calculated for groups with similar GCNs.

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4.6 Effect of regionality and methodology on productivity-diversity relationship

This study's PDR was compared to existing open ocean and mesocosm-specific datasets in three ways: (1) richness vs. POC, (2) H vs. POC, and (3) H vs. productivity rates. These methods were chosen to compare our data to previous PDR studies, as well as investigate the PDR in ways that are more suited to molecular data and use true productivity measurements such as C transport rates. Contrary to the classic unimodal PDR, all three PDRs for the Chesapeake Bay experiments were monotonically negative, though H-based PDRs were not significant (**Fig. 7**).

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While a unimodal relationship may be observed in larger regional studies, it may not apply to local diversity patterns, which can appear monotonic (Rosenzweig 1992). This can be seen when comparing a unimodal global dataset to several localized experiments which displayed both positive and negative monotonic PDRs (**Fig. 7a,b**). The issue of regionality may occur because a range of several orders of magnitude in productivity is often needed to detect a true unimodal relationship or because

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475 taxa appear more cosmopolitan at small scales (Smith 2007). The latter is true for many generalist microbes regardless of study scale and may also explain why unimodal, positive, and negative monotonic relationships are all commonly reported in aquatic microbial PDR studies (Smith 2007; Graham and Duda 2011; Skácelová and Lepš 2014). Aside from unimodal phytoplankton community PDRs, negative monotonic relationships are the most common amongst general observations of natural microbial communities.

480 The consistently negative whole community PDR of this study supported the common interpretation that competitive exclusion, in this case driven by blooming diatoms, produced the latter (negative) half of the canonical unimodal PDR. However, examining this study in isolation, it is unclear if the negative relationship resulted from an absence of low productivity observations or because 18S-based diversity measurements reveal greater than expected diversity in the inoculum community, which should have been limited by the low nutrient in situ conditions.

485 To address this, we compared our study to another whole community 18S-based H vs. productivity rate study (Wang et al. 2021) (**Fig 7c**). Both studies produced negative PDRs, despite including open ocean samples. The combined 18S-based datasets resulted in the best fitting PDR (Pearson, $p = 3.55 \times 10^{-13}$), with greater significance than either dataset in isolation. This combined dataset covered both a broad range of productivity rates and regionality, but did not result in a clear unimodal PDR. Therefore, while our data by themselves displayed the best linear fit in the traditional Richness vs. POC PDR, the H vs. 490 productivity rate PDR may be representative of larger regional PDR trends with 18S-based diversity.

5 Conclusions

This study provides a more direct approach to PDRs by measuring productivity rates concurrent with DNA sampling. Furthermore, it provides additional complexity by analyzing a more comprehensive eukaryotic community, both by using 18S-based diversity analyses and by including non-phytoplankton taxa, over the time course of the bloom. Overall, carboy 495 microcosms reached maximum diversity leading up to the peak bloom, followed by a day 4 transition in the community and increase in specific uptake rates, which led to a day 5 peak in Chl-a, fucoxanthin, and transport rates, and finally resulted in a late-bloom reduction in all phytoplankton signatures and POM accumulation (summarized in **Fig. 6**), which matched potential grazing rates.

500 The use of 18S amplicon sequencing to analyze the development and waning of a phytoplankton bloom thus provides a new lens through which to view the patterns of species succession and relationships between diversity and productivity. Although the 18S data detected a larger portion of the community with more taxonomic resolution than was possible with microscopy or pigments, a general pattern of lower diversity at the height of the bloom was still evident (**Fig. 6**). Diatoms dominated the bloom, and therefore the patterns of nutrient utilization and biomass production. The characteristic relationships between



505 productivity (at high levels of primary production) and diversity held for the diatom component of the bloom, as well as for
the entire community. The most abundant group (diatoms), however, was represented by multiple genera and multiple
“species” within some genera, rather than by one or a few major species. Many other phytoplankton genera, including
dinoflagellates and others, persisted through the bloom, clustering into assemblages that characterized different (early, mid,
late) phases of the bloom. It was not possible to attribute the decline of the bloom to one factor, since grazing and nutrient
510 depletion coincided. Several lines of evidence, including a biogeochemical model, showed that grazing was responsible for
the decline in chlorophyll–biomass, but nutrient depletion had already led to a decrease in growth rates. General reproducibility
on the temporal patterns of uptake rates and succession in the 24–L microcosms, as well as successful simulation by an oceanic
scale model, suggest that these results provide insight into coastal and oceanic phytoplankton bloom dynamics. Understanding
these dynamics along transition zones such as the lower bay is increasingly important as rising sea level pushes high salinity
515 conditions and communities, known to be the most responsive to nutrient loading (Adolf et al. 2006), further up the bay where
they will be more susceptible to nutrient loading (Bilkovic et al. 2019).

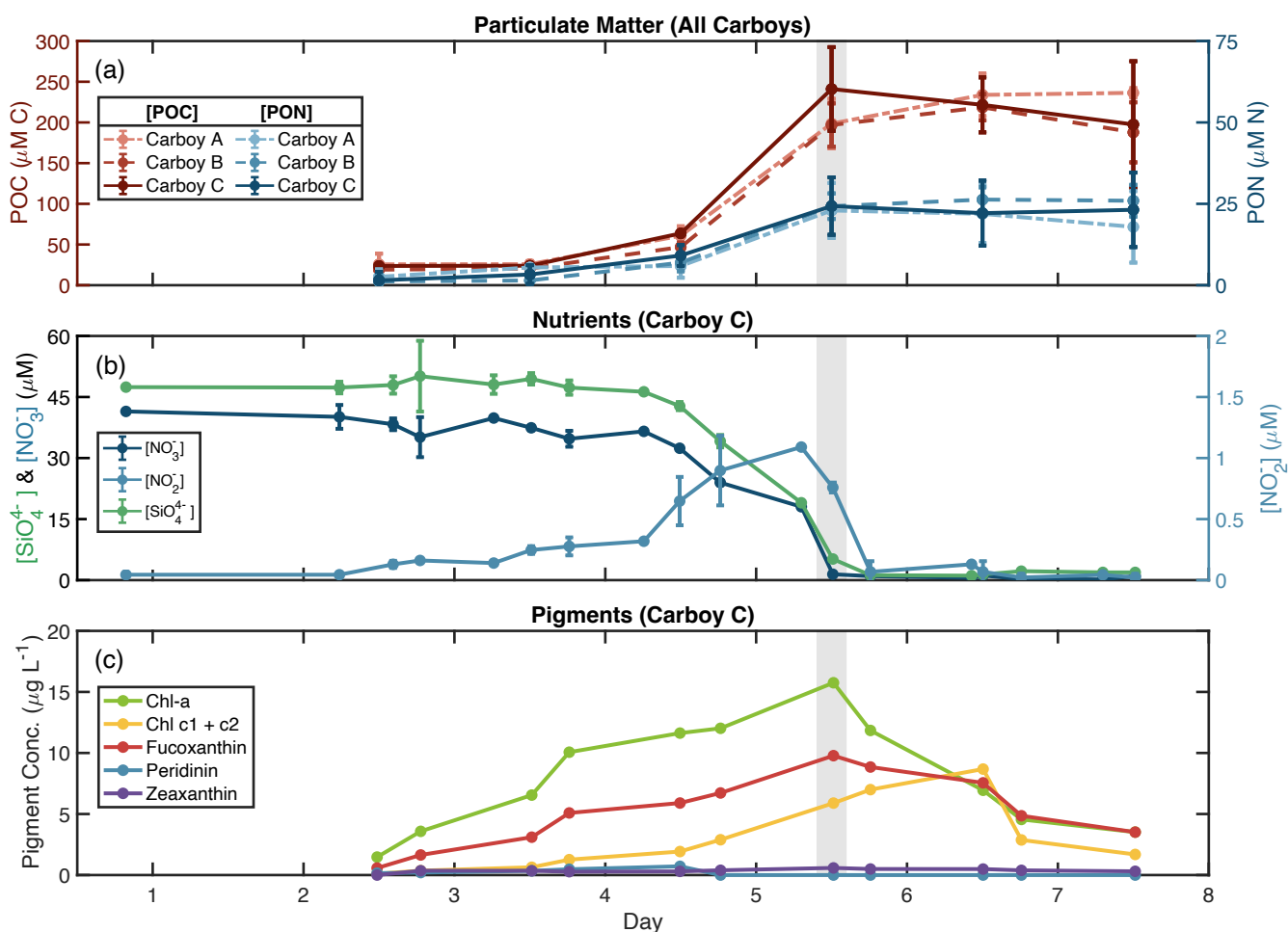
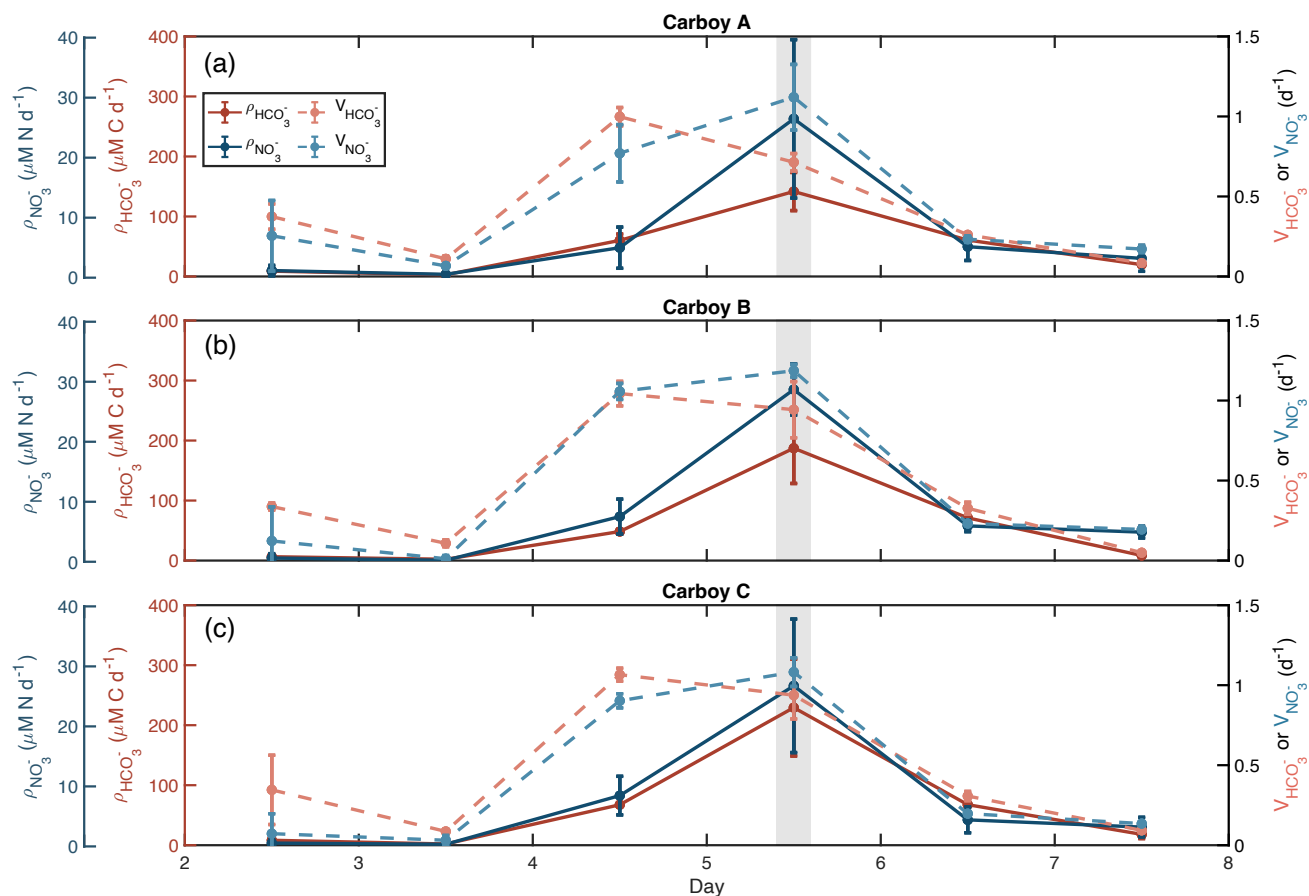


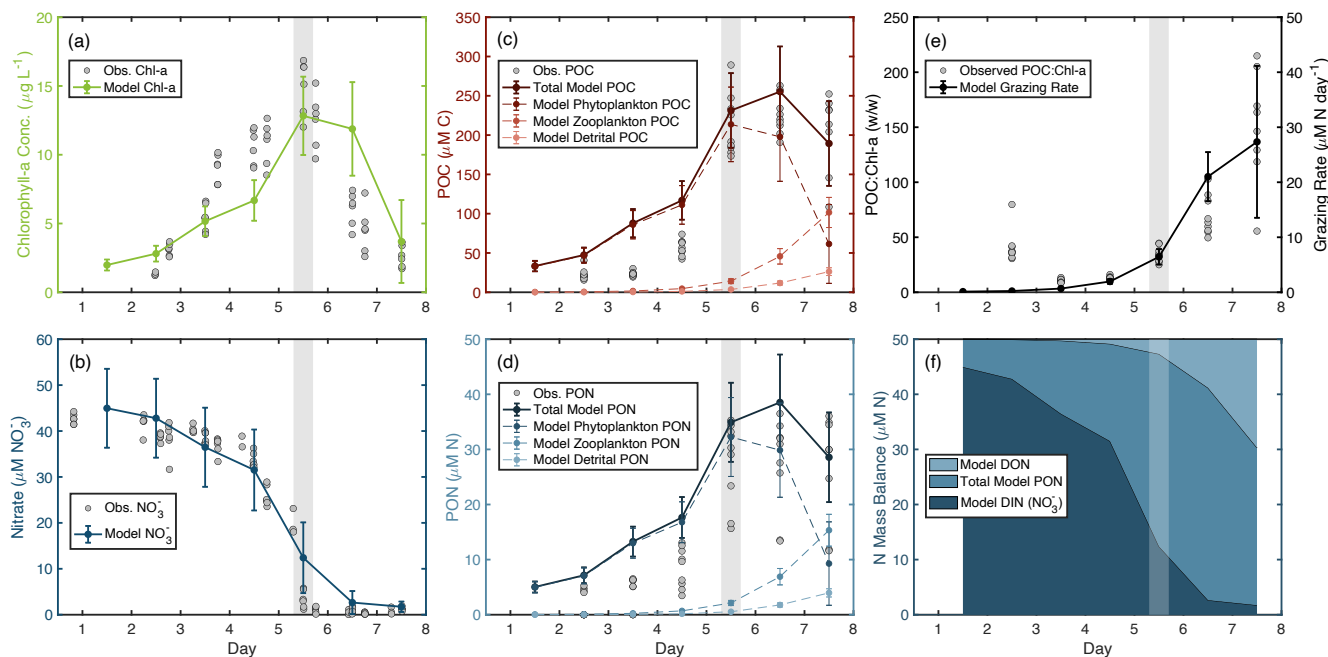
Figure 1: Biogeochemical data indicate a bloom. (a) Particulate organic carbon (POC) and nitrogen (PON) concentrations for all carboys over the course of the bloom simulation. POC is in red/pink, PON is in blue, and line type denotes carboy. Carboy C is shown as a representative example for (b) nutrients and (c) pigments (data for carboys A and B are presented in Fig. S1). (b) Nutrient measurements in carboy C during the bloom simulation are presented as dark blue lines for nitrate ([NO₃⁻]), light blue lines for nitrite ([NO₂⁻]), and purple lines for silicate ([SiO₄⁻]). (c) Pigment concentrations for chlorophyll a (Chl-a), chlorophyll c (Chl c1 + c2), and diagnostic pigments for diatoms (fucoxanthin), dinoflagellates (peridinin), and cyanobacteria (zeaxanthin). Error bars represent standard deviation of sample replicates. No error bars for are plotted pigments as only duplicate samples were taken. Time is shown as days since carboy inoculation. The grey shaded region indicates the peak bloom (~noon day 5).

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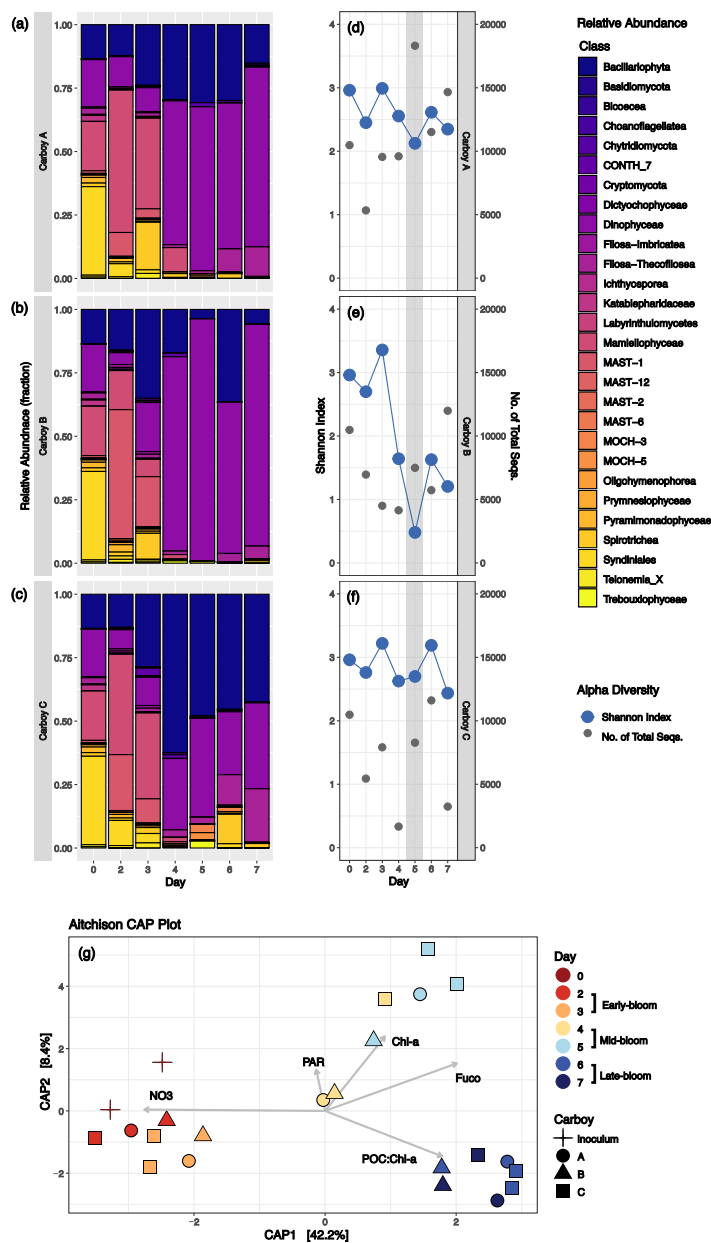


530 **Figure 2: Specific uptake rates sharply increase before the blooms peak.** Carbon (C) and nitrogen (N) uptake rates are shown in blue and red/pink, respectively for each carboy. Dark solid lines are absolute uptake rates (ρ) calculated from labeled ^{15}N - NO_3^- - and ^{13}C - HCO_3^- -sub-incubation experiments. Light dashed lines are specific uptake rates (V), which are equal to ρ normalized to POC for $\text{V}_{\text{HCO}_3^-}$ - and PON for $\text{V}_{\text{NO}_3^-}$ -. Error bars show standard deviation of triplicate samples. The light grey shaded region indicates the peak bloom (~noon day 5).



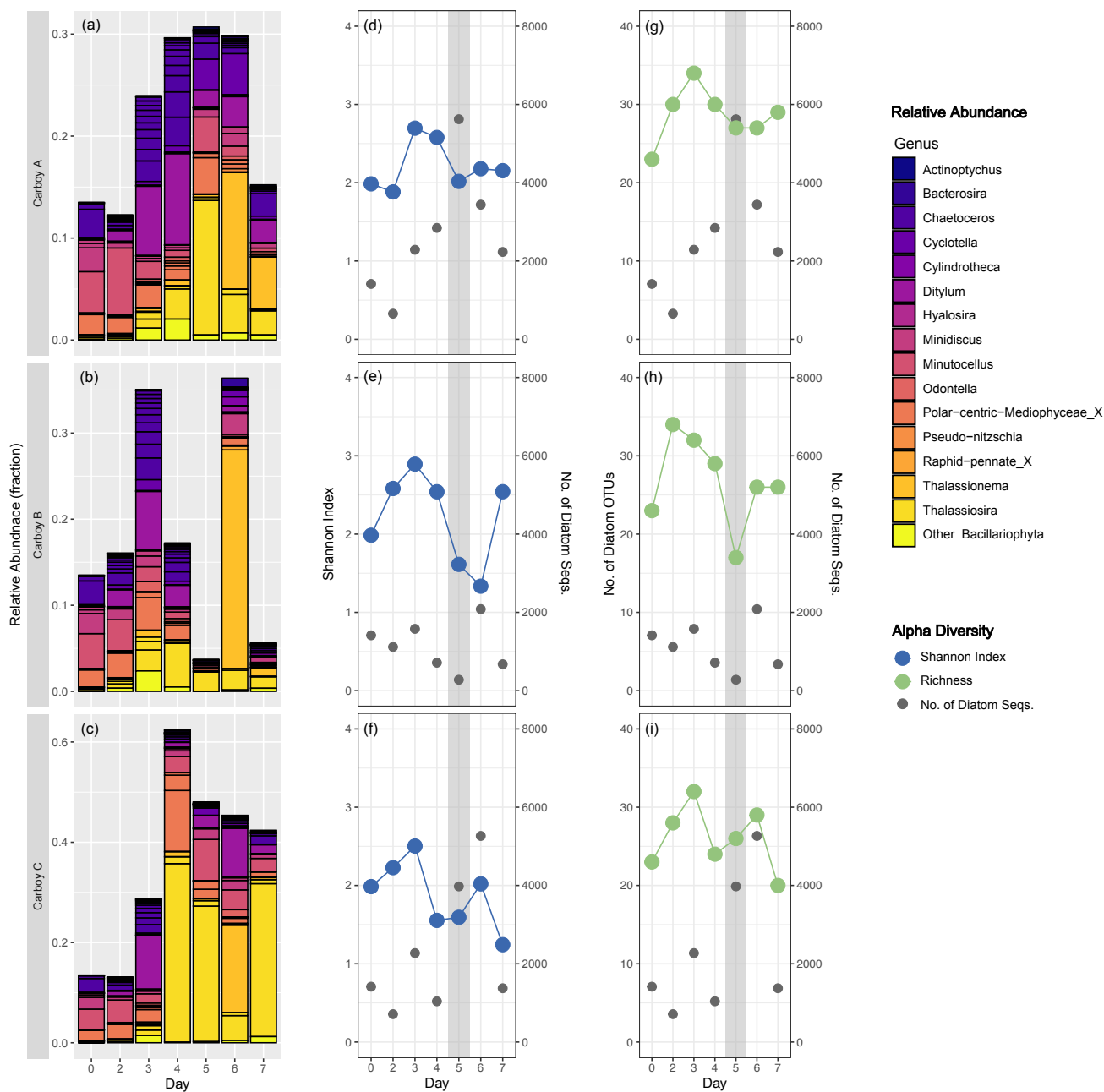
535 **Figure 3: NPZ model outputs indicate potential grazing.** Model outputs are plotted with observations for (a) chlorophyll-a, (b) nitrate ($[\text{NO}_3^-]$), and particulate organic (c) carbon (POC) and (d) nitrogen (PON) concentrations. Potential grazing rates are represented by (e) the modeled grazing rates and compared to observations of POC:Chl-a. And (f) the nitrogen mass balance shows the relative contribution of various nitrogen pools to the total nitrogen budget of the modeled bloom. Observed data are shown as grey dots. The grey shaded area indicates the peak bloom. Modeled data shown as lines with error bars representing the standard deviation of sensitivity testing.

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Figure 4: Community succession and diversity during the bloom. Community composition and diversity measurements over the course of the bloom. (a-c) Relative abundance shown as the fraction of non-metazoan 18S-derived OTUs for each carboy, colored by taxonomic class. Sequence counts for biological duplicates of inoculum and carboy C samples were merged before analysis. “Day” corresponds to the number of days since inoculation, with day 0 for each carboy represented by the same merged inoculum sample. (e-f) Shannon alpha diversity (blue) and the number of non-metazoan sequences (grey) on the same “Day” scale. The grey shaded region indicates the peak bloom. (g) Aitchison distance beta diversity measures. Color corresponds to the number of days since inoculation and shape corresponds to carboy. Beta diversity is plotted alongside relevant environmental parameters: NO₃ (nitrate concentration), PAR (photosynthetically active radiation), Chl-a (Chl-a concentration), Fuco (fucoxanthin concentration), and POC:Chl-a (carbon-to-Chl-a ratio).



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Figure 5: Diatom community succession and alpha diversity. (a-c) Relative abundance of diatom OTUs (fraction of the total non-metazoan community) for each carboy, colored by genus with outlines around individual OTUs. “Day” corresponds to the number of days since inoculation, with day 0 for each carboy represented by the same merged inoculum sample. Y-axis range is variable. (d-i) Shannon alpha diversity (dark blue), number of diatom OTUs present (i.e. richness) (dark green), and the number of diatom sequences (grey) on the same “Day” scale. The grey shaded region indicates the peak bloom.

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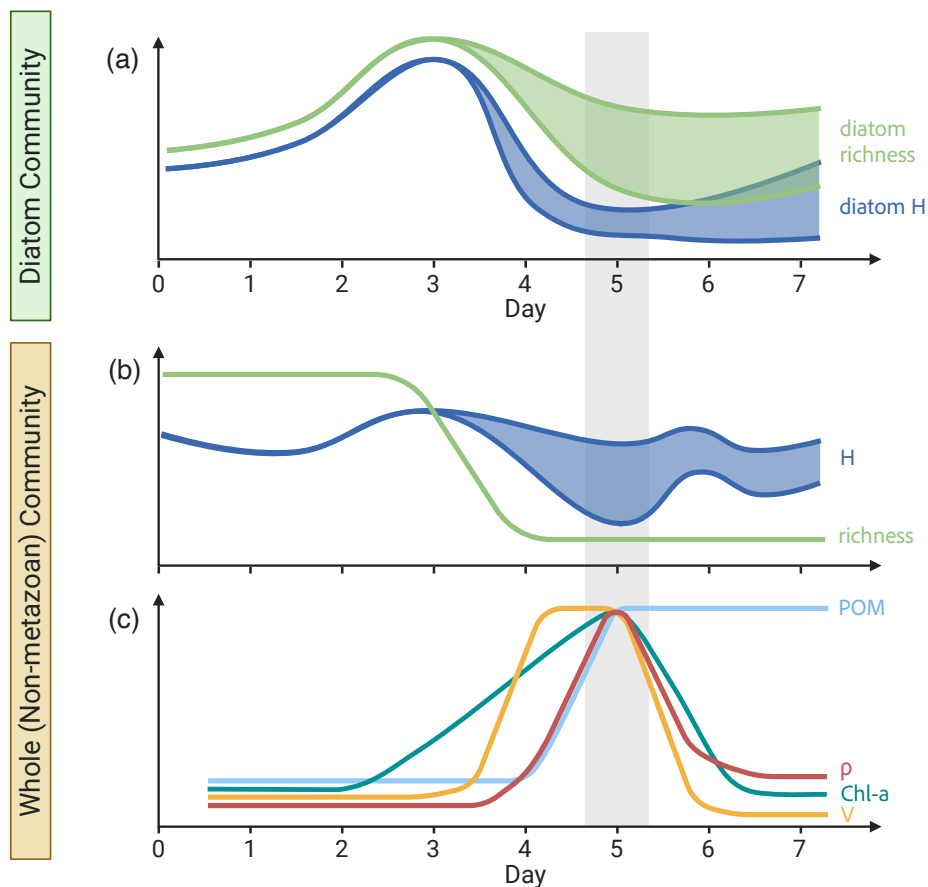
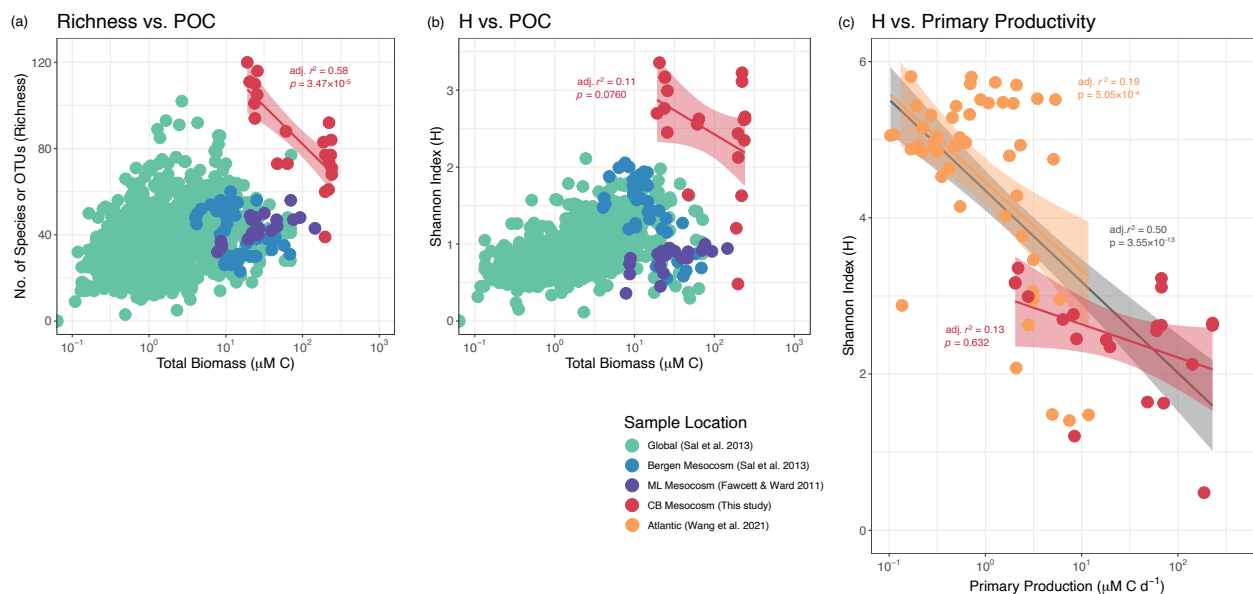


Figure 6: Bloom summary. Schematic diagram of the temporal trends in (a) diatom-specific 18S diversity, (b) whole non-metazoan eukaryotic community 18S diversity, and (c) biogeochemical measurements throughout the bloom experiments. Diversity is displayed as the number of OTUs (richness) and Shannon Index (H). Particulate organic matter (POM), chlorophyll-a concentration (Chl-a), transport rates (ρ) and specific uptake rates (V) were chosen as the primary biogeochemical parameters. Trends are averaged across carboys and shaded regions represent deviations between carboys. This figure was created in BioRender.

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565 **Figure 7: Productivity-Diversity Relationship (PDR).** Several studies were compared using 3 different PDRs. (a,b) The global (Sal et al. 2013) and the Moss Landing (ML, Fawcett and Ward 2011) diversity data were obtained via microscopy and the Bergen mesocosm data was extracted from the global dataset. (c) The Atlantic (Wang et al. 2021) and Chesapeake Bay (CB, this study) diversity data were calculated from 18S-based OTUs. The Atlantic Net Community Production rates were converted to Primary Production rates according to Li & Cassar (2016, eq. 7). Linear regressions are plotted for select datasets. Shaded regions around linear regressions indicate the 95 % confidence interval and the Pearson correlation and significance are listed in the respective color of a given dataset. The grey line in (c) is the linear regression for the combined Atlantic and Chesapeake Bay datasets.

570 Data availability

The 18S data presented in the paper are available at NCBI SRA, BioProject ID: PRJNA1222857 (<http://www.ncbi.nlm.nih.gov/bioproject/1222857>).

The biogeochemical data presented in the paper are available at BCO-DMO, project number: 869541 (<https://demo.bco-dmo.org/project/869541>).

575 Supplement

Detailed methodology and analyses, as well as supplemental tables and figures are available in the Supplementary Material and can be accessed at XXX.

Author contributions (CRediT)

All authors were responsible for writing – review and editing (equal). Additional contributions:



- 580 JAL was responsible for conceptualization (supporting), data curation (lead), formal analysis (lead), investigation (lead), methodology (equal), visualization (lead), and writing – original draft (lead).
JHV was responsible for methodology/software (supporting) and supervision (supporting).
MAP was responsible for methodology/software (supporting) and writing – original draft (supporting).
LR was responsible for funding acquisition (supporting) and supervision (supporting).
- 585 BBW was responsible for conceptualization (lead), funding acquisition (lead), methodology (equal), project administration (lead), resources (lead), supervision (lead), and writing – original draft (supporting).

Competing interests

The authors declare that they have no conflict of interest.

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