Photophysiological and light absorption properties of phytoplankton communities in the river-dominated margin of the northern Gulf of Mexico

Sumit Chakraborty1, Steven E. Lohrenz1, and Kjell Gundersen2

1School for Marine Science and Technology, University of Massachusetts-Dartmouth, New Bedford, Massachusetts, USA,
2Institute of Marine Research, Bergen, Norway

Abstract Spatial and temporal variability in photophysiological properties of phytoplankton were examined in relationship to phytoplankton community composition in the river-dominated continental margin of the northern Gulf of Mexico (NGOM). Observations made during five research cruises in the NGOM included phytoplankton photosynthetic and optical properties and associated environmental conditions and phytoplankton community structure. Distinct patterns of spatial and temporal variability in photophysiological properties were found for waters dominated by different phytoplankton groups. Photophysiological properties for locations associated with dominance by a particular group of phytoplankton showed evidence of photoacclimation as reflected by differences in light absorption and pigment characteristics in relationship to different light environments. The maximum rate of photosynthesis normalized to chlorophyll ($P_{\text{Bmax}}$) was significantly higher for communities dominated ($\geq 60\%$ biomass) by cyanobacteria + prochlorophyte (cyano + prochl). The initial slope of the photosynthesis-irradiance (P-E) curve normalized to chlorophyll ($z^B$) was not clearly related to phytoplankton community structure and no significant differences were found in $P_{\text{Bmax}}$ and $z^B$ between different geographic regions. In contrast, maximum quantum yield of carbon fixation in photosynthesis ($\Phi_{\text{Bmax}}$) differed significantly between regions and was higher for diatom-dominated communities. Multiple linear regression models, specific for the different phytoplankton communities, using a combination of environmental and bio-optical proxies as predictor variables showed considerable promise for estimation of the photophysiological parameters on a regional scale. Such an approach may be utilized to develop size class-specific or phytoplankton group-specific primary productivity models for the NGOM.

Plain Language Summary This study examined the relationships between phytoplankton community composition and associated optical properties as key factors in explaining the variability of photosynthesis-light relationships in the dynamic and complex continental margin of the northern Gulf of Mexico. Photoacclimation of phytoplankton in different light environments, from the highly turbid Mississippi River delta to relatively oligotrophic offshore waters, was evident and considered to be a factor regulating the efficiency of carbon fixation in photosynthesis. Our findings were unprecedented in revealing significant differences in photosynthetic parameters between the major phytoplankton groups in northern Gulf of Mexico waters. This enabled us to build an empirical model to predict photosynthetic parameters for the major phytoplankton groups across the entire study area. These findings provide a basis for further efforts to apply this approach for wider-scale modeling of primary production in the northern Gulf of Mexico.

1. Introduction

Environmental variability can act at the level of physiology (e.g., photoacclimation and nutritional status) within phytoplankton groups or species as well as through effects on species composition to cause variations in photosynthesis-irradiance (P-E) relationships. Two important parameters that describe the P-E relationship are $z^B$ (mg C (mg Chl)$^{-1}$ h$^{-1}$ (μmol photon m$^{-2}$ s$^{-1}$)$^{-1}$), which is the initial slope of the photosynthesis-irradiance relationship and $P_{\text{Bmax}}$ (mg C (mg Chl)$^{-1}$ h$^{-1}$), which is the maximum photosynthetic rate under light-saturated conditions. Variability in these two parameters occurs due to changes in
phytoplankton physiology and community structure and may contribute to uncertainty in estimates of primary production derived using photosynthesis-irradiance models. Therefore, an understanding of the basis for variations in photophysiological properties of phytoplankton communities is crucial. Relationships between photophysiological properties and phytoplankton community structure have been characterized in different parts of the world ocean [Cermen et al., 2005; Claustre et al., 2005; Hashimoto and Shiomoto, 2002; Suggett et al., 2009; Uitz et al., 2008]. In addition to community structure, other key factors that influence photophysiological parameters include environmental variables such as irradiance, temperature, nutrient availability, as well as other biological and ecological factors [e.g., Sakshaug et al., 1997; Behrenfeld et al., 2002; Claustre et al., 2005; Uitz et al., 2008; Xie et al., 2015].

The NGOM is an optically complex and highly productive continental margin (as high as 400 g C m\(^{-2}\) yr\(^{-1}\)) [Lohrenz et al., 2014; Heilman and Rabalais, 2008]. Discharge from both Mississippi and Atchafalaya Rivers strongly influences the distribution of dissolved and particulate material [D’Sa and DiMarco, 2009; D’Sa et al., 2007], nutrients [Lehrter et al., 2013; Turner et al., 2007], and the availability and spectral properties of light [Schaefur et al., 2011]. In conjunction with the large environmental gradients, our prior research has demonstrated substantial variations in phytoplankton communities across different water mass types and over seasonal time scales in these continental shelf waters [Chakraborty and Lohrenz, 2015]. A recent study [Zhao and Quigg, 2015] in a shallow inner shelf region of the NGOM also highlighted the importance of phytoplankton community composition in understanding the diel patterns of photosynthesis and photoacclimation under the complicated light and nutrient conditions.

Primary productivity (PP) is highly variable over space and time, and prior studies in the NGOM [Sklar and Turner, 1981; Lohrenz et al., 1999; Lehrter et al., 2009; Quigg et al., 2011] have documented strong relationships of regional PP to physical and meteorological factors that modulate the spatiotemporal changes in freshwater inputs, light, nutrients, temperature, and phytoplankton biomass. Over the years, our understanding of variability in regional PP has improved. However, the photophysiological basis of this regional variability remains poorly understood. Several studies [e.g., Uitz et al., 2008, 2010] have shown considerable promise in using phytoplankton size class-specific estimates of photosynthesis-irradiance (P-E) parameters to improve estimates of PP using remote sensing algorithms. However, such phytoplankton size class-specific or group-specific information is limited for the continental margin of the NGOM. An earlier study [Lohrenz et al., 1994] in the Mississippi River plume (MRP) observed spatial and temporal differences in the P-E parameters that were partially attributed to variations in river discharge, depth, daily PAR, and temperature. That work mainly considered the variation in P-E parameters due to the environmental conditions without explicitly examining the potential linkages between the composition of the phytoplankton community and the associated P-E parameters.

Here we build on previous studies with the overarching goal of improving our understanding of how variations in photosynthetic properties are related to phytoplankton community composition and associated absorption properties in continental shelf waters of the NGOM. Additionally, we sought to identify patterns of photoacclimation in waters having different dominant phytoplankton groups. Finally, we present an empirical approach for estimation of P-E parameters in the NGOM, specific for the different phytoplankton communities, taking into account the relationships to environmental variables and to bio-optical proxies of phytoplankton size class and pigment composition. Such an approach has the potential to improve regional biogeochemical models [Fennel et al., 2011; Xue et al., 2013], which are requisite to better understanding of carbon cycling and ecosystem processes in the northern Gulf, and may also lead to improved estimates of primary productivity from remotely sensed data.

### 2. Materials and Methods

Water samples were collected on board the R/V Cape Hatteras for (Gulf Carbon 1–3 and 5) and R/V Hugh R. Sharp (Gulf Carbon 4) during five cruises that took place in January, April, July, October 2009, and March 2010. Eight transects were made across the NGOM shelf (Figure 1), occupying contrasting water mass types from freshwater (S ≤ 15) riverine end members dominated by the Mississippi-Atchafalaya River system to oligotrophic oceanic (S > 33) waters. Water samples were collected at each station using 10 L Niskin bottles mounted on a rosette and CTD (SeaBird SBE911 plus) profiling system. Discrete water samples were collected and subsequently filtered for particulate absorption, phytoplankton pigment analysis, nutrients, and
P-E experiments. Mixed layer depth ($Z_m$, m) was determined at each station using the criterion of [Mitchell and Holm-Hansen, 1991] as the first depth where the density ($\sigma-t$) change over a 5 m interval was $\geq 0.05$ kg m$^{-3}$. Profiles were obtained of hyperspectral downwelling irradiance, $E_d$ ($\lambda$, $z$), in units of W m$^{-2}$ and where $\lambda$ is wavelength and $z$ is the depth, using a Satlantic HyperPRO free-falling optical profiler equipped with a surface downwelling irradiance ($E_s$) reference to correct for temporal variability in $E_s$ during profiling. The maximum depth of the optical profiles ranged from a few meters in nearshore waters to 50 m in oligotrophic waters. The value of $E_d$ ($\lambda$, $z$) at just below the sea surface was obtained by extrapolation of logarithm-transformed data from the 1 to 4 m depth interval. Typically, three to five profiles were averaged at a given station. Photosynthetically active radiation (PAR, mol photons m$^{-2}$ d$^{-1}$) was calculated versus depth by integrating $E_d$ ($\lambda$, $z$) over 400–700 nm. The attenuation of PAR in the water column, $K_d$ (PAR), m$^{-1}$, was determined as the slope of the least squares regression fit to logarithm-transformed $E_d$ (PAR) as a function of depth. Following Lehrter et al. [2009], we chose the depth of the euphotic zone ($Z_{eu}$) as the depth at which PAR had decreased to 1% of the surface value,

$$Z_{eu} = \frac{\ln (0.01)}{-K_d}.$$

### 2.1. Photosynthesis-Irradiance Curve Measurements

Photosynthesis versus irradiance (P-E) curves were determined from $^{14}$C-HCO$_3$ in vitro incubations similar to that described by Lewis and Smith [1983] and Lohrenz et al. [1994]. Eighteen 10 mL subsamples with a final specific activity of approximately 1 µCi mL$^{-1}$ were incubated at 0.03–10 mol quanta m$^{-2}$ h$^{-1}$ irradiance and each incubation was terminated by filtration after 0.5 h followed by treatment of the filters with 200 µL 0.1N HCl to eliminate residual inorganic $^{14}$C. Samples for dissolved inorganic carbon (DIC) and chlorophyll $a$ (chl $a$) analyses were collected in conjunction with the P-E samples and used to estimate the chlorophyll-specific rate of carbon fixation. The observed P-E relationships were fitted to a mathematical expression [Platt et al., 1980], which was chosen because it was found to perform well in representing patterns in the data. Derived parameters included the initial slope of the light-saturated curve ($a_B$, g C (g chl $a$ h)$^{-1}$ (µmol photons m$^{-2}$ s$^{-1}$)$^{-1}$), the specific photosynthetic rate at optimal light ($P_{max}$, mg C mg chl $a^{-1}$ h$^{-1}$), maximum potential light-saturated photosynthetic rate ($P_B^S$), and the rate of photoinhibition ($b_B$, g C (g chl $a$ h)$^{-1}$ (µmol photons m$^{-2}$ s$^{-1}$)$^{-1}$). From these parameters, we estimated the light saturation index ($E_{sat}$ = $P_{max}$/$a_B$, mol C (mol photons)$^{-1}$), which is considered the threshold for light limitation of photosynthesis [Platt et al., 1980]. Notation for optical variables and photosynthetic parameters is given in Table 1.
The maximum quantum yield of CO₂ fixation was determined using the following equation:

\[ \Phi_{\text{cmax}} = 12,000 \alpha_{\text{a}} \left( \frac{\int_{400}^{700} \alpha_{\text{ph}}(\lambda) d\lambda}{\int_{400}^{700} \alpha_{\text{ph}}(\lambda) d\lambda} \right) \]

where 12,000 is the molar weight (mg) of carbon and \( \alpha_{\text{a}} \) is the light-limited slope of the P-E curve, normalized to chl \( a \). In addition, a normalized spectral slope of the \( \alpha_{\text{ph}} \) spectrum between 488 and 532 nm was obtained by normalizing \( \alpha_{\text{ph}}(\lambda) \) by chl \( a \) concentrations. The maximum quantum yield of CO₂ fixation was determined using the following equation:

\[ \Phi_{\text{cmax}} = 12,000 \alpha_{\text{a}} \left( \frac{\int_{400}^{700} \alpha_{\text{ph}}(\lambda) d\lambda}{\int_{400}^{700} \alpha_{\text{ph}}(\lambda) d\lambda} \right) \]

The \( \alpha_{\text{ph,slope}} \) was used to examine the relationships of photoprotective pigments to the shape of the phytoplankton absorption in the blue-green spectral region and served as a proxy for photoacclimation and/or changes in pigment composition [Eisner et al., 2003].

### 2.3. Pigment Analyses

For pigment analyses, seawater samples with volume ranging from 0.15 to 1.5 L at the shallow estuarine end member station to 2 – 5 L at the deep offshore slope waters were filtered through 47 mm Whatman GF/F glass-fiber filters. Filters were immediately frozen and stored in liquid N₂ until analysis. The details of the HPLC analyses are described in Chakraborty and Lohrenz [2015]. The pigment data were further organized into the two categories of accessory pigments: (i) photosynthetic carotenoids or PSC—the sum of fucoxanthin, peridinin, 19'-hexanoyloxyfucoxanthin, and 19'-butanoyloxyfucoxanthin and (ii) photoprotective carotenoids or PPC—the sum of zeaxanthin, diadinoxanthin, alloxanthin, and \( \beta \)-carotene. CHEMTAX software v...
1.95 [Mackey et al., 1996; Roy et al., 2011] was used to determine the relative contributions of phytoplankton groups to chl \(a\). The CHEMTAX-derived phytoplankton community data set was also separated into three dominant phytoplankton groups. Stations were given phytoplankton community designations as (1) diatom dominated, when the relative percentage of diatoms was \(\geq 60\%\), (2) as cyanobacteria and prochlorophyte (cyano + prochl) dominated, or (3) as mixed groups, when neither diatoms nor cyano + prochl were dominant. Major contributors to the mixed group were chlorophytes, cryptophytes, and haptophytes. Chlorophytes and cryptophytes were particularly abundant in stations near the delta and shallow inner shelf waters while haptophytes were generally more abundant in areas away from the direct influence of the river discharge and in offshore waters [see Chakraborty and Lohrenz, 2015]. These operational designations provide a context for examining the bio-optical and photophysiological properties of the phytoplankton communities.

2.4. Measurements of Suspended Particulate Matter (SPM)
Seawater samples were collected by draining an entire Niskin bottle into a 20 L carboy. Prior to withdrawing samples for filtration, the carboy was agitated to ensure uniform distribution of sample. Water samples of 0.05–3.5 L were filtered onto pretared, 0.45 \(\mu\)m pore size, 45 mm diameter Poretics polycarbonate membrane filters. Filtered samples were stored at \(-20^\circ C\) in a small plastic petri dish until returning to the lab. Samples were dried for 24 h at 60\(^\circ\)C and weighed on a Lettler Precision Analytical Balance. This was repeated over 3–4 days until weight was stable.

2.5. Colored Dissolved Organic Matter (CDOM) Absorption Measurements
Seawater samples were filtered under low vacuum through 0.22 \(\mu\)m polycarbonate filters prerinsed with 50 mL Milli-Q water. The filtrate was immediately stored at 4\(^\circ\)C in acid cleaned and Milli-Q water rinsed 250 mL amber glass (Teflon-capped) bottles. Prior to analysis, the samples were allowed to come to room temperature to reduce the chance of any bias occurring due to temperature difference between the sample and the Milli-Q water reference. CDOM absorbance of the filtered water was measured at 1 nm intervals from 250 to 800 nm in a 10 cm quartz cuvette using a bench top spectrophotometer (Cary 300). A baseline correction was made by subtracting the mean absorbance between 650 and 680 nm from the spectrum to remove instrument baseline drift and refractive effects. The measured absorbance (\(A(\lambda)\)) values were converted into absorption coefficients, \(a_{CDOM}(\lambda)\) (\(m^{-1}\)) according to the following:

\[
a_{CDOM} (\lambda) = \frac{2.203 \cdot A(\lambda)}{I},
\]

where \(I\) was the path length of the cuvette. The spectral slope (SCDOM) for each spectrum was calculated by applying a nonlinear, least squares fit to the measured \(a_{CDOM}(\lambda)\) values between 350 and 500 nm [Babin et al., 2003]. The fit was performed using the raw (i.e., nonlog-transformed) data [Twardowski et al., 2004]:

\[
a_{CDOM} (\lambda) = a_{CDOM} (\lambda_r) e^{-SCDOM(\lambda - \lambda_r)}
\]

2.6. Nutrients
Samples for nutrients were initially filtered through Whatman 25 mm GF/F filters and refrigerated in acid-washed, polyethylene bottles until analysis on shore. Nutrient samples were analyzed for nitrate (NO\(_3\)), nitrite (NO\(_2\)), ammonium (NH\(_4\)), silicate (SiO\(_3\)), and phosphate (PO\(_4\)). Fluorometric methods were used for nitrogen species and spectrophotometric methods for PO\(_4\) and SiO\(_3\). All nutrient analyses were performed using an Astoria-Pacific A2 + 2 nutrient auto-analyzer (methods #A179, A027, A205, and A221; Astoria Pacific International).

2.7. Statistics
Statistical analyses were conducted using SPSS v24 software. Relationships between P-E parameters and environmental, biological, and optical variables were examined using Spearman correlation coefficients (\(r\)). The significance of \(r\) was evaluated with the two-sample \(t\) test for correlation (Table 2). Kolmogorov-Smirnov and Shapiro-Wilk tests were employed to test the normality of the distribution for each of the variables. Data were log-transformed prior to statistical analyses according to Campbell [1995]. In the case of nonnormal distributions, the nonparametric Kruskal-Wallis test was used, which is analogous to an ANOVA.
The variables used for the classification of the regions were carried out using SPSS v.24 software. City block distances were calculated using Ward’s minimum variance diate, far field, and offshore, Figure 1) were identified using hierarchal cluster analysis (HCA). The HCA was evaluated. Multicollinearity of variables was tested using the variance inflation factor. Homoscedasticity of error assumption of multicollinearity of variables used in the model and homoscedasticity of errors was also eval-

Differences in P-E parameters between different geographical regions were assessed using nonparametric Kruskal-Wallis ANOVA. The associations between environmental and bio-optical predictor variables and P-E response variables were analyzed using multiple-regression analysis. A stepwise multiple linear regression was used to determine a subset of variables that explained the largest amount of variation in the P-E parameters using a significance cutoff of $p < 0.05$. From this, we selected key factors for the regression (Table 3a). The assumption of independence of error of the multiple linear regressions was verified using the Durbin-Watson statistic and the statistical significance of the model was assessed using the $F$-ratio. The assumption of multicollinearity of variables used in the model and homoscedasticity of errors was also evaluated. Multicollinearity of variables was tested using the variance inflation factor. Homoscedasticity of error

| Variables | $p_{\text{max}}^0$ | $\Phi$ | $E_s$ |
|-----------|-------------------|--------|-------|
| Environmental | mg C mg chl $a^{-1}$ h$^{-1}$ | (n = 61) | mg (mg chl $a h^{-1}$) | (n = 61) | mol C | (mol photons$^{-1}$) | (n = 61) | $\mu$mol photons m$^{-2}$ s$^{-1}$ | (n = 61) |
| $S$ | 0.36832 | 0.0282 | -0.47995 | 0.15331 |
| $T$ | 0.48063 | -0.00127 | -0.40367 | 0.46675 |
| $\text{chl }a$ | -0.50367 | -0.03078 | 0.62911 | -0.35847 |
| DIN | -0.0934 | 0.19454 | 0.45014 | -0.09574 |
| $\text{SiO}_4$ | -0.17995 | 0.07062 | 0.36895 | -0.31064 |
| $\text{PO}_4$ | 0.08887 | 0.04151 | 0.2565 | -0.02997 |
| MLD | 0.25735 | 0.16037 | -0.33889 | -0.11338 |
| $Z_{\text{env}}$ | 0.46032 | -0.04232 | -0.5837 | 0.419 |
| $K_d$ | -0.43546 | 0.03451 | 0.58027 | -0.39167 |
| SPM | -0.32953 | 0.16078 | 0.42075 | -0.34174 |

To account for the optical complexity in NGOM and the influence of the large rivers (Mississippi and the Atchafalaya Rivers) on photophysiology and bio-optical properties, four geographical zones (delta, intermediate, far field, and offshore, Figure 1) were identified using hierarchal cluster analysis (HCA). The HCA was carried out using SPSS v.24 software. City block distances were calculated using Ward’s minimum variance method. The variables used for the classification of the regions were $T$, $S$, chl $a$, $a_{\text{phi}}(440)$, $K_d$, $Z_{\text{env}}$, SPM, $a_{\text{DOM}}(412)$, and $a_{\text{MFP}}(440)$ (supporting information Figure S1). HCA yielded four major groups and they were similar to the geographical zones identified by several recent studies [Fennel et al., 2011; Laurent et al., 2013; Xue et al., 2013] in the region. The term “geographical zones” refers to the different water types based on their different physicochemical and optical properties and that corresponded to different locations along the shelf.

Differences in P-E parameters between different geographical regions were assessed using nonparametric Kruskal-Wallis ANOVA. The associations between environmental and bio-optical predictor variables and P-E response variables were analyzed using multiple-regression analysis. A stepwise multiple linear regression was used to determine a subset of variables that explained the largest amount of variation in the P-E parameters using a significance cutoff of $p < 0.05$. From this, we selected key factors for the regression (Table 3a). The assumption of independence of error of the multiple linear regressions was verified using the Durbin-Watson statistic and the statistical significance of the model was assessed using the $F$-ratio. The assumption of multicollinearity of variables used in the model and homoscedasticity of errors was also evaluated. Multicollinearity of variables was tested using the variance inflation factor. Homoscedasticity of error

| Groups (Phytoplankton) | Intercept | Temperature (°C) | $a_{\text{phi}}(676)$ (m–1) | $a_{\text{phi}}(440) / a_{\text{phi}}(676)$ | $a_{\text{phi}}$ slope | RMSE | APD |
|------------------------|-----------|-----------------|----------------------------|-------------------------|---------------------|-------|-----|
| Diatom                 | 4.961     | -0.076          | 461.24                     | -1.39                   | 52.683              | 1.5   | 8.39|
| Mixed                  | -4.160    | 0.165           | 140.199                    | 2.975                   | -20.743             | 2.65  | 49.05|
| Cyanobacteria + prochl | -0.676    | 0.291           | 40.016                     | 2.124                   | 124.211             | 1.39  | 4.99|
| All data               | -3.735    | 0.288           | 141.08                     | 1.637                   | 73.035              | 2.55  | 26.89|

*Bold values indicate significant relationships (i.e., $p < 0.05$).
distributions was assessed by plotting the standardized residuals of the regression against the unstandardized predictor variables.

3. Results

3.1. Regional Environmental Conditions

Here we briefly summarize the environmental and optical properties of the geographic zones during the study, and subsequently describe the relationship between phytoplankton, light absorption properties, photophysiology, and community composition. Detailed descriptions of the regional variability of optical and environmental properties and their relationship with the phytoplankton community are given elsewhere [Chakraborty, 2013; Chakraborty and Lohrenz, 2015].

As expected, the delta and the intermediate zones were directly influenced by the large rivers, characterized by low salinity waters (overall mean salinity of 25.8 ± 7.1 standard deviation). Seawater temperatures were relatively high during summer and water columns were highly stratified, while lower temperatures and vertically mixed conditions prevailed during winter 2009 and spring 2010 (supporting information Table S1). Highest mean (0.58 ± 0.25 m−1) light attenuation coefficients (Kd) and shallowest euphotic depths (Zeup) (10.85 ± 6.66 m) were observed during spring 2010. Average values of SPM, aaph(440), and aCDOM(412) were significantly higher (p < 0.05) in the delta and intermediate waters than the far-field and the offshore waters. SPM, aaph(440), and aCDOM(412) values exhibited significant relationships with salinity (ANOVA, p < 0.05, not shown), decreasing with increasing salinity away from the direct influence of rivers. Variations of dissolved nutrients were strongly correlated with river discharge (not shown).

In the far-field zone, away from the direct influence of the rivers and associated freshwater inputs, we observed smaller amplitude in the seasonal variation in the environmental properties. Average salinity (34.7 ± 1.56) in far field was much higher than the delta and intermediate zones, with the exception of spring 2010 (31.04 ± 2.98), when high river discharge (supporting information Figure S2) and strong winds out of the northwest resulted in an extended river plume [Huang et al., 2013; Chakraborty and Lohrenz, 2015]. Average values of SPM and CDOM and nonalgal particulate absorption were much lower in the far field than observed in the delta and intermediate zones (supporting information Table S1), and as a result light availability was much higher in the far-field zone. Kd was significantly lower (K-S test, p < 0.05), while Zeup was significantly higher (K-S test, p < 0.05) than observed in the delta and intermediate zones. Average nutrient concentrations were also low (supporting information Table S1).

Over the continental slope of the NGOM, the offshore zone had oligotrophic characteristics. Typical salinity was 36.07 ± 0.95 and nutrient concentrations were generally low; however, atypically low salinities and relatively high nutrient concentrations were observed during spring 2010 (34.2 ± 2.48) and in summer 2009 (<31), which coincided with offshore extension of the river plume as previously described in other studies [Feng et al., 2012]. Spring 2010 was also a period of unusually low temperatures, ~2–3°C lower when compared to means from 2002 and 2011 [Huang et al., 2013].

3.2. Variability in Phytoplankton Light Absorption, Community Composition, Size, and Pigments

The chlorophyll-specific phytoplankton absorption (aaph(676)) varied on both seasonal and spatial scales (Figure 2a), a pattern seen in numerous studies [Bricaud and Stramski, 1990; Bricaud et al., 2004]. The aaph(676) spectra had characteristic absorption maxima at 440 and 676 nm and varied by orders of magnitude across different zones (Figure 2a). The CHEMTAX-derived phytoplankton diatom group generally dominated the delta and intermediate areas (>60% of total chl a). However, significant (Kruskal-Wallis, p < 0.05) seasonal differences were observed during summer and fall when the community was dominated by cyanobacteria + prochlorophyte (cyano + proch) and other phytoplankton groups [Chakraborty and Lohrenz, 2015]. Relative

### Table 3b. Coefficients of the Stepwise Multiple Regression for $a_{676}$ (DIN = Dissolved Inorganic Nitrogen)

| Groups (Phytoplankton) | Intercept | DIN $q$ (μM) | $a_{440}$ (676) (m−1) | $a_{440}$ (676) | RMSE | APD |
|------------------------|-----------|---------------|-------------------------|-----------------|------|-----|
| Diatom                 | −0.005    | 0.002         | 0.858                   | 0.010           | 0.007 | 10.64 |
| Mixed                  | 0.008     | 0.002         | 0.746                   | −0.002          | 0.009 | 23.66 |
| Cyan + prochl          | −0.010    | 0.002         | 2.081                   | −0.001          | 0.008 | 5.23  |
| All data               | 0.011     | 0.001         | 0.755                   | −0.001          | 0.008 | 15.43 |




proportions of cyano + prochl were systematically higher in offshore waters (Figure 2b), with highest chl \(a\) proportions observed during spring and fall 2009 (50 \(\pm\) 15.4%). In contrast, their relative contribution to chl \(a\) in the offshore waters was exceptionally low (ranged 1.9–8.7%) in spring 2010. Such low abundance of cyano + prochl during spring 2010 in offshore waters was attributed to the offshore extension of a river plume feature as previously discussed. The high availability of nutrients (due to strong NW and upwelling-favorable winds) [Huang et al., 2013; Chakraborty, 2013] resulted in conditions suitable for the proliferation of larger phytoplankton groups [Chakraborty and Lohrenz, 2015].

Major differences in bio-optical properties were evident for stations dominated by different phytoplankton groups and among the geographic domains. Concentrations of chl \(a\) varied over orders of magnitude from high chl \(a\) delta and intermediate waters (range 5.0.36–16 and 0.2–29 mg m\(^{-3}\), respectively, Figure 3a) to low values in far-field and offshore waters (0.2–3.4 and 0.04–3.8 mg m\(^{-3}\), respectively). The mean chl \(a\) specific absorption of phytoplankton \((a_{ph}^{a})\) was significantly lower \((p < 0.05, K-S test)\) in the delta region (Figure 3b), while values increased in far-field and offshore waters and highest values were observed in the offshore region. The cyano + prochl group was consistently associated with significantly higher values of \(a_{ph}^{a}\), while for diatoms the values of \(a_{ph}^{a}\) ranged lower and the mixed group had intermediate values. This pattern was consistent among all the domains (Figure 3). The ratio of \(a_{ph}^{(440)}:a_{ph}^{(676)}\) was low (median = 2.19, Figure 3c) in the delta and intermediate zones generally dominated by diatoms and other larger groups, while higher \(a_{ph}^{(440)}:a_{ph}^{(676)}\) values (range = 1.8–4.57, median 2.9, Figure 3c) were associated with far-field and offshore areas where smaller phytoplankton (mainly cyanobacteria and prochlorophytes) dominated. Values of \(a_{ph}^{(440)}:a_{ph}^{(676)} > 2.5\) can be considered as representing the picophytoplankton group, while \(a_{ph}^{(440)}:a_{ph}^{(676)} < 2\) are indicative of microphytoplankton-dominated communities [Bricaud et al., 2004; Stramski et al., 2001, Stramski and Morel, 1990]. Highest values of the normalized slope over 488–532 nm \((a_{ph, slope})\) were generally found in delta and intermediate zones (Figure 3d). The decrease in \(a_{ph, slope}\) (increasingly negative) from delta to offshore waters likely reflected changes in photoacclimation in relationship to higher available irradiance in offshore waters along with associated changes in phytoplankton composition. This was further supported by the finding that \(a_{ph, slope}\) varied inversely in relationship to the ratio of PPC:PSC, with decreasing (negative) values of \(a_{ph, slope}\) and increasing PPC:PSC from delta to offshore \((p < 0.05, r = −0.7; \text{supporting information Table S1})\).

### 3.3. Variability of P-E Parameters

Differences in surface \(P_{\text{max}}^{g}\) values between the different geographical zones were not significant (Figure 4a, \(p > 0.05, K-S \text{ test}\)). However, differences in \(P_{\text{max}}^{g}\) were observed for stations dominated by different phytoplankton groups (Figure 4a). Average \(P_{\text{max}}^{g}\) was \(\sim 54\%\) lower at diatom-dominated (>60%) stations relative to cyano + prochl-dominated (>60%) locations. A positive relationship was found between the optical...
proxy for phytoplankton size, $a_{ph}(440):a_{ph}(676)$, and $P_{max}$ (Figure 5 and Table 2, also see supporting information Figure S3). The relationship of $P_{max}$ to the CHEMTAX-derived fraction of chl $a$ associated with the diatom group was negative (Table 2), while a positive relationship was seen for cyanobacteria (Table 2). A more complex situation was found for the mixed assemblage group (Table 2 and supporting information Figure S3); both the maximum (25.62 mg C (mg chl $a$)$^{-1} h^{-1}$) and minimum (2.92 mg C (mg chl $a$)$^{-1} h^{-1}$) values of $P_{max}$ were associated with stations having mixed assemblages. Significant correlations were found between $P_{max}$ and environmental variables (Table 2). Specifically, the correlation was positive with temperature (Figure 5a and Table 2), while relationships to $K_d$ and chl $a$ (Figure 5b and Table 2) were negative. No significant relationship was seen for DIN and $P_{max}$ (Figure 5c). Highest $P_{max}$ values were observed in the midsalinity range (25–32), and a weak positive correlation existed with salinity (Table 2). Values of the initial slope $d_1$ varied widely, over orders of magnitude (range $= 0.006$–$0.103 g C (g chl $a$ h)$^{-1}$ (μmol photons m$^{-2}$ s$^{-1}$)$^{-1}$) within the zones (Figure 4b) and among phytoplankton groups and no significant differences were evident. No significant correlations were found between $d_1$ and ambient nutrients, temperature, $K_d$, and chl $a$ (Figures 5d–5f and Table 2).

Unlike $P_{max}$ and $d_1$, the maximum quantum yield of carbon fixation in photosynthesis ($\Phi_{cmax}$) differed significantly (K-S test, $p < 0.05$) between the geographic zones (Figure 4c). Values of $\Phi_{cmax}$ decreased along the progression from the river-influenced, light-limited delta to the more oligotrophic waters offshore. Average $\Phi_{cmax}$ values in the delta were about 39% greater than that of the offshore waters (Figure 4c and supporting information Table S1). The relative fraction of total chl $a$ associated with the diatom group was positively correlated with $\Phi_{cmax}$ (Table 2) and the mean $\Phi_{cmax}$ for diatom-dominated stations was significantly higher ($p < 0.05$, K-S test) in comparison to the cyanobacteria-dominated communities (Figures 4c and 5g–5i). In general, $\Phi_{cmax}$ was significantly correlated with environmental variables ($S$, $T$, nutrients, chl $a$, and $K_d$).
2). Among the environmental variables, chl \( a \) and \( K_d \) accounted for the greatest percentage of the variation in \( \Phi_{c_{\text{max}}} \) (Table 2). Furthermore, values of \( \Phi_{c_{\text{max}}} \) were negatively correlated with blue to red ratios \( \alpha_{\text{bp}}(440):\alpha_{\text{bp}}(676) \) and had a positive relationship to \( \alpha_{\text{bp}} \_\text{slope} \) (Table 2).

The light saturation index (\( E_k \)) also varied among water masses and phytoplankton groups and photoacclimation state. Significantly higher (\( p < 0.05 \), K-S test) values of \( E_k \) in the cyano + prochl-dominated waters were observed ranging between 197.8 and 779.4 \( \mu \text{mol m}^{-2} \text{s}^{-1} \) in comparison to a range of 150.0–463.5 \( \mu \text{mol m}^{-2} \text{s}^{-1} \) for stations dominated by diatoms. The higher values of \( E_k \) corresponded with more negative values of \( \alpha_{\text{bp}} \_\text{slope} \) (Table 2). \( E_k \) was also strongly related to the light environment, with significant correlations observed for \( Z_{\text{eq}} \) and \( K_d \) (Table 2).

4. Discussion

4.1. Factors Regulating Photophysiological Parameters in NGOM: Role of Phytoplankton Community Composition and Environmental Variables

A major focus of this study was to improve understanding of how the variability in P-E parameters was associated with differences in phytoplankton community composition and in relationship to the environmental variables in different geographical zones (delta, intermediate, far field, and offshore, Figure 1). Major differences in \( P_{\text{max}} \) were observed among the dominant phytoplankton groups. Significantly (\( p < 0.05 \)) lower values were observed for diatom-dominated populations than in the mixed and the cyano + prochl-dominated communities, and this pattern was consistent across the different geographic zones. Diatoms, as expected, dominated the high biomass, river-influenced and comparatively low-light deltaic and intermediate zones. The low \( P_{\text{max}} \) and associated low \( \alpha_{\text{bp}}(440):\alpha_{\text{bp}}(676) \) observed in those zones were indicative of increased pigment packaging [Mitchell-Innes and Walker, 1991; Marra et al., 2007]. The diatom-dominated communities in low-light environments (high \( K_d \)), particularly in the delta and river-influenced intermediate zones, would be expected to exhibit lower carbon:chl \( a \) ratios and, therefore, have lower \( P_{\text{max}} \), associated

Figure 4. Variability of photosynthetic properties among the dominant phytoplankton groups across geographic zones, (a) maximum photosynthetic rate, \( P_{\text{max}} \), (b) initial slope of the P-E curve, \( \alpha^0 \), (c) maximum quantum yield of carbon fixation, \( \Phi_{c_{\text{max}}} \), and (d) light saturation index, \( E_k \).
with low-light acclimation [MacIntyre et al., 2002]. Cyano + prochl communities dominated waters near the delta during summer [Chakraborty and Lohrenz, 2015], which could be attributed to the combination of strongly stratified, high-light, and nutrient-limited conditions, that would favor high carbon:chl \( a \) ratios and hence high \( P_{\text{max}}^b \) (Figure 4a). The greatest range of variation in \( P_{\text{max}}^b \) was observed in the mixed phytoplankton group in the far-field and offshore zones. Within the mixed group, haptophytes were a major phytoplankton taxon and are known for their ability to successfully adapt to a wide range of environmental conditions [Liu et al., 2009; de Vargas et al., 2015]. Offshore waters were dominated by the cyano + prochl group. However, \( P_{\text{max}}^b \) was lower in the cyano + prochl fraction for the offshore waters than that of the cyano + prochl-dominated delta communities. The cyano + prochl in the delta zone, although referred to as cyano + prochl, was mainly composed of \textit{Trichodesmium} and \textit{Synechococcus} and other unicellular cyanobacteria [Ren, 2010]. Divinyl chlorophyll \( a \) (diagnostic pigments for prochlorophytes) was rarely found in our pigment samples from the delta region. The offshore assemblage of cyano + prochl in NGOM was a combination of \textit{Synechococcus} and \textit{Prochlorococcus} [Chakraborty and Lohrenz, 2015; Wawrik and Paul, 2004], and \textit{Prochlorococcus} has been reported to exhibit lower \( P_{\text{max}}^b \) and \( a_{\text{ph}} \) than \textit{Synechococcus} [Shimada et al., 1996]. Therefore, the variation of \( P_{\text{max}}^b \) observed within the delta and offshore cyano + prochl communities was likely a combined effect of physiological differences among taxa as well as differences in environmental conditions.

The significant relationship of \( P_{\text{max}}^b \) to temperature (\( p < 0.05 \), Figure 5a) observed in our study was consistent with that of previous work [Sukenik et al., 1987; MacIntyre and Geider, 1996; Bouman et al., 2005]. Correlation between \( P_{\text{max}}^b \) and temperature has been reported numerous times from temperate regions to higher latitudes [Harrison and Platt, 1986; Davison, 1991; Sakshaug et al., 1997; Bouman et al., 2005]. Significantly higher seawater temperatures were associated with the cyano + prochl-dominated as compared to diatom-dominated communities (K-S test, \( p < 0.05 \), Figure 5a). Falkowski and Raven [1997] reported that increasing
temperatures can stimulate photosynthesis through a direct effect on Calvin cycle enzymatic activity up to an optimal temperature after which rates in photosynthesis can decline due to inactivation and denaturation of enzymes [Raven and Geider, 1988]. Observed relationships between $P_{\text{max}}^B$ and temperature, particularly in the diatom-dominated stations in our study, were consistent with this concept. Below 21°C, we observed a positive correlation ($r = 0.75$, supporting information Figure S4) between $P_{\text{max}}^B$ and temperature for diatom-dominated stations, while $P_{\text{max}}^B$ values showed a decline at higher temperatures (>21°C). $P_{\text{max}}^B$ was generally negatively correlated with temperature for diatom-dominated stations (Table 2), whereas $P_{\text{max}}^B$ and temperature were positively correlated for cyanobacterial-dominated stations. These correlations of temperature with $P_{\text{max}}^B$ among the phytoplankton groups may also reflect covariation with other variables. For example, in temperate and subtropical environments, stratification of the water column is mainly temperature driven. In an earlier study [Chakraborty and Lohrenz, 2015], it was shown that water temperature, stratification, and mixed layer depths were important environmental factors related to phytoplankton community composition. Highly stratified, low nutrient, and high temperature (i.e., summer) conditions favored picophytoplankton groups, including cyanobacteria and prochlorophytes, while eukaryotic phytoplankton were more prevalent under mixed or less stratified conditions. The correlation of $P_{\text{max}}^B$ and temperature can partly be due to the fact that small cells with higher $P_{\text{max}}^B$ are typically found in stratified oligotrophic conditions (offshore waters, Figure 5), while larger cells with lower $P_{\text{max}}^B$ are dominant during colder periods, when the water column was vertically mixed [Chakraborty and Lohrenz, 2015]. The results of this study are in agreement with the assertion of Platt et al. [2005] that under light-saturated conditions $P_{\text{max}}^B$ is negatively proportional to cell size.

The range of values of $P_{\text{max}}^B$ and $x^B$ in the shelf (delta, intermediate, and far field) during this study was similar to that observed by previous studies [Lohrenz et al., 1999; Lehrter et al., 2009, John et al., 2012; Zhao and Quigg, 2015]. Variations in $x^B$ showed no apparent relationship to environmental variables, and there was no clear spatial or temporal pattern in observed values. This finding was consistent with that of Lehrter et al. [2009]. Their study also did not find significant differences in $P_{\text{max}}^B$ and $x^B$ values along the continental shelf of the NGOM for different geographic regions. Higher $P_{\text{max}}^B$ associated with smaller phytoplankton has been reported in several previous studies [Malone and Neale, 1981; Côté and Platt, 1983; Sathyendranath et al., 1999; Bouman et al., 2005]. However, some studies [Cermeno et al., 2005; Claustre et al., 2005; Ulitz et al., 2008] have also reported higher $I_{\text{max}}^B$ values for larger phytoplankton, such as diatoms, leaving considerable ambiguity in $P_{\text{max}}^B$ and phytoplankton size class relationships.

The range of $\Phi_{\text{cmax}}$ for our study was also consistent with the range for natural algal communities under optimum physiological conditions (0.06–0.08 mol C (mol photon)$^{-1}$) as previously reported [Bannister, 1974; Babin et al., 1996]. In contrast to $P_{\text{max}}^B$ and $x^B$, both geographical and taxonomical differences were evident in $\Phi_{\text{cmax}}$ values. Similar to the findings of this study, high $\Phi_{\text{cmax}}$ in diatom-dominated populations and other large-sized phytoplankton in low-light (delta zone) environments have been observed elsewhere in the ocean [Cermeno et al., 2005]. Claustre et al. [2005] also reported that large phytoplankton composed mainly of diatoms had relatively high $\Phi_{\text{cmax}}$ in comparison to that for small phytoplankton in the North Atlantic. High pigment packaging in diatoms may result in low $a_{\text{ph}}(440)/a_{\text{ph}}(676)$ and lower $a_{\text{ph}}^B(x)$ which could explain, at least partially, the high $\Phi_{\text{cmax}}$. Values of $\Phi_{\text{cmax}}$ were lower for the smaller phytoplankton groups. The picophytoplankton (generally <2 mm in size), particularly cyanobacteria (e.g., Synechococcus and Trichodesmium) have been observed to be the predominant picoprykaryotes in the region [Chakraborty and Lohrenz, 2015; Wawrik and Paul, 2004]. Significant seasonal differences in $\Phi_{\text{cmax}}$ existed; values of $\Phi_{\text{cmax}}$ in summer were lower in comparison to other periods (supporting information Table S1), reflecting a combination of changes in community composition, pigment composition, and environmental conditions. DIN:PO4 values have been shown to be lower in summer in our study region [Dagg and Breed, 2003; Lohrenz et al., 2008] (also see supporting information Table S1 for nutrient concentrations), leading to speculation that nutrient limitation [Dortch and Whitledge, 1992; Justic et al., 1995] may be a factor in the lower observed $\Phi_{\text{cmax}}$. Other reasons for the lower $\Phi_{\text{cmax}}$ could be attributed to the relative increase of nonphotosynthetic (photoprotective) pigments such as zeaxanthin [Chakraborty and Lohrenz, 2015]. Zeaxanthin is a diagnostic pigment for cyanobacteria and it along with other photoprotective pigments are associated with the cell wall [Fujuta et al., 1994] and contribute to the absorption of PAR, but dissipate the absorbed excitation energy rather than transferring it to the photosynthetic apparatus [Falkowski and Raven, 1997]. High PPC:PSC ratios during summer [Chakraborty and Lohrenz, 2015] would result in a depression of $\Phi_{\text{cmax}}$ [Babin et al., 1996] and could
also explain the negative correlation between \( \Phi_{cmax} \) and temperature (Table 2). Therefore, it can be concluded that diatom-dominant populations possessed higher light utilization efficiency for photosynthesis and that observed differences in \( \Phi_{cmax} \) reflect a variety of differences related to community and pigment composition and environmental variables.

We observed a significant relationship between \( \Phi_{cmax} \) and \( \alpha_{ph \_slope} \) (Table 2), which was consistent with photoacclimation as an additional factor influencing variations in \( \Phi_{cmax} \). The \( \alpha_{ph \_slope} \) decreased (increase in negative values) from the low-light delta to high-light environments of far field and offshore (Figure 6). The decrease in \( \Phi_{cmax} \) in the far-field and offshore zones could be due the increase in PPC:PSC, and \( \alpha_{ph \_slope} \) is inversely proportional to PPC:PSC ratio (mentioned earlier). Key factors influencing the variability in \( \Phi_{cmax} \) include relative abundance of nonphotosynthetic pigments, the number of functional reaction centers, and/or the cyclic electron flow around photosystem I or II [Babin et al., 1996]. A study in North Atlantic has also shown significant negative relationships between \( \Phi_{cmax} \) and \( \alpha_{ph}(440) : \alpha_{ph}(676) \) and to the ratio of nonphotosynthetic pigments to chl \( a \) concentrations [Stuart et al., 2000]. High values of PPC are generally found in high irradiance-acclimated cells [Morel and Bricaud, 1981]. In our study, we surmised that the significant relationship observed between \( \Phi_{cmax} \) and \( \alpha_{ph \_slope} \) was a consequence of phytoplankton assemblages acclimating to ambient light by regulating the relative amounts of photosynthetic or nonphotosynthetic pigments. This would, in turn, influence the phytoplankton absorption spectra and thereby the pattern of variation in \( \Phi_{cmax} \). Phytoplankton acclimated to low-light conditions tend to have lower \( P_{max} \) and \( E_k \) than in regions with high light. The PPC group of pigments mainly functions by dissipating the absorbed energy as heat under high-light conditions and so plays a photoprotective role in the cell [Falkowski and Raven, 1997].

In contrast to some previous studies [Zhao and Quigg, 2015; Lohrenz et al., 1994], a strong covariance in \( P_{max} \) and \( \alpha_b \) was found in this study. The strong positive covariation between \( P_{max} \) and \( \alpha_b \) (Figure 7) is typical of \( E_k \)-independent variability that has been frequently observed in other ocean regions [Behrenfeld et al., 2004, and references therein]. However, the underlying basis for the \( E_k \)-independent variability remains largely unresolved. This covariation has been linked to the diversion of reductants produced by the light reactions of photosynthesis to reactions other than carbon fixation. Among potential causes for this diversion of reductants is nutrient limitation [e.g., Côté and Platt, 1983; Behrenfeld et al., 2004]. A
recent study [John et al., 2012] in the Mississippi River plume observed E_k-independent variability in the larger fraction (>2μm) of phytoplankton and suggested nutrient-related stress probably resulted in the E_k-independent variability in those groups. The P_{max}^B and a^B relationship observed in our study (Figure 7) was associated with differences between the diatom-dominated and cyanoprochlorophytoplankton. Therefore, an alternative explanation for the P_{max}^B and a^B relationship is that it reflects underlying differences in the photometabolism of photosynthetic prokaryotes and eukaryotes (e.g., diatoms and the mixed group) [Behrenfeld et al., 2004].

**4.2. Estimation of P-E Parameters: Implication for Primary Production Estimates in NGOM**

An examination of differences in P_{max}^B and a^B for each phytoplankton group did not reveal significant differences between the different geographic regions. There were significant differences in parameters between the phytoplankton groups, however. Hence, we sought to develop empirical relationships of P-E properties to optical indices of phytoplankton absorption and environmental variables that were specific to each subset of stations dominated by the different phytoplankton groups for the entire study area. We used a stepwise multiple-regression model with the predictor variables listed in Table 3a to develop a relationship to estimate P-E parameters. For all data, the stepwise multiple regression explained >60% of the variation for P_{max}^B and ~46% for a^B (Figure 8). Temperature, q_{ph}(676), q_{ph,slope}, and q_{ph}(440):q_{ph}(676) were the best predictors for P_{max}^B, while DIN, q_{ph}(676), and q_{ph}(440):q_{ph}(676) were best predictors for a^B. Using the combination of environmental variables and phytoplankton group-specific indices to estimate P-E parameters, we obtained a mean absolute percent error for the estimate of P_{max}^B was 26.9% (Figure 8a) and a^B was 15.4% (Figure 8b). The percent root-mean-square error (RMSE) of model estimates was generally low; however, larger errors were associated with the mixed phytoplankton-dominated stations. The maximum error and large variability observed in the mixed groups likely reflects the photophysiological differences among taxa within the mixed community.

The idea of using phytoplankton class-specific indices to improve estimates of P-E parameters has been previously implemented by Uitz et al. [2008], and the results of this study provide further support for such an approach. The predictor values of q_{ph}(676), q_{ph,slope}, and q_{ph}(440):q_{ph}(676) can be acquired or derived on synoptic scales through satellite ocean color imagery, and may be used to refine estimates of P_{max}^B (and a^B) in the NGOM region. Accounting for variations in phytoplankton community composition should in principle allow for improved modeling of primary production for the river-influenced continental margin of the NGOM as well as other systems in which there are substantial, spatial gradients in phytoplankton community composition.

**5. Conclusions**

This study characterized the photophysiological and bio-optical properties of the phytoplankton community in different geographical zones (light and nutrient regimes, physicochemical conditions) of the NGOM.
Our results highlight the importance of phytoplankton community composition and associated optical properties as key factors in explaining the variability of P-E curve parameters in the dynamic-complex continental margin of NGOM. Photoacclimation of phytoplankton in different light environments, including delta, intermediate, far field, and offshore, was evident and considered to be a factor regulating the efficiency of carbon fixation in photosynthesis (quantum yield). Our findings were unprecedented in revealing significant differences in photosynthetic parameters between the major phytoplankton groups in northern Gulf of Mexico waters. This enabled us to build an empirical model based on multiple linear regression for the major phytoplankton groups across the entire study area. The empirical model presented in this study, specific for subsets of stations dominated by different phytoplankton groups, generated satisfactory estimates of the $P_{\text{max}}$ and $\alpha$. These findings are encouraging for further efforts to apply this approach for wider-scale modeling of primary production.

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