



Modeling Net Growth of *Phaeocystis antarctica* Based on Physiological and Optical Responses to Light and Temperature Co-limitation

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Temperature and light are fundamental environmental variables which regulate phytoplankton growth rates when nutrients are in excess. For polar coastal oceans that are undergoing changes in sea ice cover and warming, light, and temperature are particularly important for bloom dynamics. Using colonial Phaeocystis antarctica cultures grown at steady-state, we assessed the combined effect of these two environmental controls on net growth rate (μ_n), chlorophyll-specific absorption of light (a_{nh}^* (λ)), and quantum yields for growth (ϕ_{μ}). Specific net growth rates (μ_n) varied from 0.04 to 0.34 day^{-1} within a matrix of light and temperature ranging from 14 to 542 μ mol quanta $m^{-2} s^{-1}$ and -1.5 to 4°C. Values of a_{nh}^* (λ) varied significantly with light but only slightly with temperature. Values of $\phi_{\rm tr}$ ranged from 0.003 to 0.09 mol C (mol quanta absorbed)⁻¹ with highest values at low light and 4°C. For excess irradiances or low temperatures where growth rate is inhibited, quantum yields were low. The low ϕ_{μ} values are attributed both to increased absorption by photoprotective pigments compared to photosynthetic pigments and thermodynamic control of dark reaction enzymes. The systematic changes in photophysiological properties of P. antarctica in relation to temperature and light were used to develop a series of nested light- and temperature-dependent models for μ_n , a_{nh}^* (λ), and ϕ_{μ} . A model for a_{nh}^* (300–700 nm) was developed that takes into account the systematic changes in $a_{ph}^{*}(\lambda)$ due to pigment packaging effects and cellular concentrations of chlorophylls and photoprotective pigments. Also, a model for ϕ_{μ} was developed based on a cumulative one-hit Poisson probability function. These model parameterizations for absorption and quantum yield are combined into an overall model of net growth that can be applied easily to P. antarctica bloom dynamics using remote sensing data for temperature, light, and chlorophyll a. Furthermore, modeling based on the biophysical variables a_{oh}^{*} (λ), and ϕ_{μ} that are shown to regulate the growth rate provides a more fundamental mechanistic approach compared to other modeling methods that do not explicitly resolve photon flux into the cell or the quantum yield.

Keywords: *Phaeocystis*, phytoplankton, photosynthesis, modeling, co-limitation, photophysiology, quantum yield, absorption

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INTRODUCTION

The colonial prymnesiophyte Phaeocystis can dominate coastal, ice edge, and open ocean blooms in polar and temperate waters with significant implications for carbon export (Smith et al., 1991; DiTullio et al., 2000; van Leeuwe et al., 2007; Pavlov et al., 2017). The importance of light, temperature, and nutrients in governing phytoplankton growth and bloom formation has motivated research on metabolic and cellular characteristics of phytoplankton in response to environmental stressors. Singly, the light-, temperature-, and nutrient- dependence of phytoplankton growth has been well-characterized for a variety of species under a wide range of experimentally controlled laboratory conditions providing the basis for generalized models of phytoplankton growth for temperature, nutrients, and light limitation. Nutrientlimited growth has been described as a function of cell quota which is determined by Michaelis-Menten uptake kinetics (Droop, 1983). These kinetic models have been modified to account for multiple nutrients by invoking Leibig's Law of the minimum (Falkowski and Raven, 2007). Temperature-dependent growth has been modeled with the Arrhenius equation (Li, 1980) and maximal growth has been based on empirical derivations determined from cultures (Eppley, 1972). Light-dependent growth has been modeled as a variety of different formulations which allows growth to increase with irradiance, saturate, and decrease at high irradiances (Goldman and Carpenter, 1974; Laws and Bannister, 1980; Moisan et al., 2002). While nutrient limitation is generally modeled on the basis of the "limiting" nutrient, phytoplankton must also acclimate to large variations in light and temperature (Behrenfeld et al., 2002; Ciotti et al., 2002). Often these regulate growth in an interactive co-limitation matrix (Sosik and Mitchell, 1994; Sunda and Huntsman, 1997, 2011) however there is lacking a mechanistic framework for modeling phytoplankton growth rate based on the biophysical variables of cellular absorption and quantum yield under simultaneous co-limitation.

Blooms of Phaeocystis sp. have been reported in diverse oceanic provinces for temperature ranging from less than 0°C to greater than 35°C and for both shallow and deep mixed layers (Gieskes et al., 2007; Gypens et al., 2007). Data sets required to specify physiological parameters for growth at various environmental conditions are relatively rare. A review of Chl:C ratios in polar and temperate microalgae at different acclimation states is found in Lacour et al. (2017)Various mathematical descriptions for modeling phytoplankton growth that include multiple limitations establish the foundations for our approach (Shuter, 1979; Laws and Bannister, 1980; Rhee and Gotham, 1981a,b; Geider et al., 1998; Behrenfeld et al., 2002; Edwards et al., 2016). These concepts have been extended to the estimates of primary production and growth rates from satellites (Behrenfeld et al., 2005; Arteaga et al., 2014, 2016). Growth rates regulated by irradiance, temperature, and nutrients have been modeled using different approaches based on algal metabolism (Laws and Bannister, 1980), elemental ratios of chlorophyll or carbon (Geider et al., 1998; Graff et al., 2016), or cell carbon quota (Zonneveld et al., 1997). Some of these models have limited ecological applicability because they require parameters that

are not easily measured in natural systems. We model steadystate growth based on the mechanistic biophysical parameters of chl-a specific spectral absorption and photosynthetic quantum yeield $(a_{ph}^{*}(\lambda) \text{ and } \phi_{\mu})$, which are directly regulated by light, temperature, and nutrients (Kiefer and Mitchell, 1983; Sakshaug et al., 1989; Cullen, 1990; Moisan and Mitchell, 1999). For brevity, all symbols are defined in Table 1. For steady-state growth, a* ph (λ) and ϕ_{μ} vary in response to temperature, light, and nutrient regulation of growth (Mitchell and Kiefer, 1988; Nelson and Prézelin, 1990; Sosik and Mitchell, 1991, 1995; Moisan and Mitchell, 1999; Westberry et al., 2008). Combining estimates of phytoplankton absorption (Kishino et al., 1985) with rates of carbon fixation allows for validation of models based on $a_{ph}^{*}(\lambda)$ and ϕ_{μ} for natural communities (Cleveland et al., 1989; Smith et al., 1989; Marra et al., 1995; Sosik and Mitchell, 1995; Sosik, 1996). Thus, the model we describe here to predict $a_{ph}^{*}(\lambda)$ and ϕ_{μ} for light and temperature co-limitation can easily be applied and validated for natural communities in polar regions using routine methods and is most applicable when blooms dominated by Phaeocystis are forming and nutrients are in excess. The model is most likely not as robust during pre-bloom situations with a mixed community assemblage and for post bloom conditions when nutrients are exhausted.

Modeling growth rates of phytoplankton over large geographical scales will require parameterizations that take into account the taxonomic diversity of the phytoplankton community and variability driven by physiological acclimation (Moisan et al., 2017). In this paper, we have developed a model of growth rate using the colonial prymnesiophyte, P. antarctica, an ecologically and biogeochemically important organism in the polar and subpolar carbon cycle (Stuart et al., 2000; Smith et al., 2003; Pabi and Arrigo, 2006; Smith et al., 2006; Gieskes et al., 2007) and is responsible for significant amounts of new production (Arrigo et al., 1999). Temperature, light, and iron are major limiting factors that control the initiation and development of phytoplankton blooms in the Southern Ocean. We describe the range of variability and relationships of μ_n , a_{ph}^* (λ), and ϕ_{μ} for *P. antarctica* for a realistic range of temperature, light, and nutrient conditions that occur during the Antarctic spring bloom. Using steady-state experimental data, we developed a series of nested models that together form a photophysiological model of net growth. Our model is based on contemporary bio-optical models for photosynthesis (Kiefer and Mitchell, 1983; Cullen, 1990; Moisan and Mitchell, 1999) which utilize $a_{ph}^{*}(\lambda)$ and ϕ_{μ} as photophysiological variables to describe phytoplankton net growth,

$$\mu_n = \frac{chl \, a}{C} \int_{350nm}^{700nm} a_{ph}^* (\lambda) E_0(\lambda) \phi_\mu \, d\lambda. \tag{1a}$$

We determined the environmental control of a_{ph}^{*} (λ), and φ_{μ} by assessing their magnitude for co-limitation using controlled light and temperature experiments. Our model for φ_{μ} is based on a cumulative one-hit Poisson probability distribution and introduces the concept that both light and temperature regulate the maximal quantum yield of growth for any acclimated state.

λ	Wavelength (nm)					
chl a	Chlorophyll a (mg m $^{-3}$)					
Т	Temperature (°C)					
P _{net}	Net primary production					
μ_n	Specific rate of net growth (day ⁻¹)					
μ_m	Maximal specific rate of net growth (day^{-1})					
С	Particulate carbon (mg m $^{-3}$)					
$E_O(\lambda)$	Spectral quantum scalar irradiance (mol quanta m $^{-2}$ s $^{-1}$ nm $^{-1}$)					
PAR	Photosynthetically available radiation defined as $\int_{350nm}^{700nm} E_o(\lambda) d\lambda \ (\mu mol quanta m^{-2} s^{-1})$					
F _{DCMU}	In vivo DCMU-enhanced fluorescence (relative units)					
a _{ph} (λ)	Absorption by phytoplankton (m^{-1})					
a _{ph} (λ)	Chl-specific absorption of phytoplankton (m ² mg chl a^{-1})					
a _{ps} (λ)	ChI-specific absorption due to photosynthetic pigments $(m^2 mg chl a^{-1})$					
a _{pp} (λ)	Chl-specific absorption due to photo-protective pigments $(m^2 mg chl a^{-1})$					
ϕ_{μ}	Quantum yield for growth [mol C fixed (mol quanta absorbed) ⁻¹]					
φm	Maximal quantum yield [mol C fixed (mol quanta absorbed) ⁻¹]					
$\phi_{mE_0,T}$	Maximal quantum yield at a steady state irradiance and temperature [mol C fixed (mole quanta absorbed) $^{-1}$]					
Eku	Photoacclimation parameter (m ² / μ mol quanta m ⁻² s ⁻¹)					
a, b	Coefficients of power function fit for $\phi_{mE_0,T}$					
$C_0,C_{E,}C_T$	Coefficients for intercepts, irradiance and temperature multiple-linear regression fit for $a_{ph}^{*}(\lambda)$					
α, μ _S ^β , β	Coefficients for fitting light-dependent growth					
α1, α2	Coefficients for temperature-dependence of α					
P _{sb}	Curve fitting parameter for a P vs. E curve					

In general, the maximal quantum yield for any acclimated state should be modeled as a function of the limiting growth conditions (Moisan and Mitchell, 1999) rather than using the theoretical maximum (Sakshaug et al., 1989). Our formulation is based on Cullen (1990) and we extend our previous model of ϕ_{μ} (Moisan and Mitchell, 1999) as a variable dependent on both light and temperature,

$$\phi_{\mu} = \phi_{mE_o,T} \frac{1 - \exp^{-E_o/E_{k\mu}}}{E_o/E_{k\mu}}$$
(1b)

where ϕ_{μ} is the quantum yield of growth, $\phi_{m,Eo,T}$ is the maximal quantum yield for growth for an acclimated state within the light and temperature limitation matrix, $E_{k\mu}$ is the photoacclimation parameter (Falkowski and Raven, 2007), and E_0 is photosynthetically available radiation.

The culture experiments were designed to assess the broad range of temperature and light that Antarctic phytoplankton experience in nature that is relevant to formation of blooms. Since a bloom cannot form unless there are excess nutrients to support the growth of the bloom, we do not model nutrients although clearly macronutrients and iron can regulate the crop size and productivity (Sedwick, 2007; van Leeuwe and Stefels, 2007). By characterizing the light and temperature response of phytoplankton absorption and quantum yield over a realistic environmental range, we have parameterized a biooptical model for μ_n for application to ocean color satellites and buoys. This effort to describe and model the variability in the biophysical parameters that mediate temperature and light control of phytoplankton growth provides important details related to phytoplankton acclimation when multiple environmental variables are regulating growth simultaneously.

MATERIALS AND METHODS

Cultures of colonial *P. antarctica* (CCMP 1374) were grown semi-continuously at -1.5, 0, 2, and 4°C for 5–8 generations in f/2 medium (Gulliard and Ryther, 1962) under continuous blue light ranging from 14 to 542 mol quanta m⁻² s⁻¹. The spectrum of the light source was similar to the light quality observed in the upper ocean (Moisan and Mitchell, 1999). Cultures were maintained at ~30% of the carrying capacity of the culture medium. This protocol results in a steady-state acclimated condition at the time of experimental sampling. Details of the illumination source, blue filter and culturing system and our method to estimate spectral quantum flux in the cultures can be found in Moisan and Mitchell (1999).

Specific Growth Rate (µn)

Specific growth rate was estimated by a linear regression of \log_e transformed daily determinations of *in vivo* fluorescence intensity (n = 2) measured with a Turner Model 10 fluorometer at the same time each day. POC data was also collected during mid-log phase growth at the time of the experiments, allowing an estimate of the carbon-based growth rate for each experimental culture.

Fluorometric Estimates of Chl a

Chl *a* concentrations (n = 3) were estimated fluorometrically using a Turner Model 10-AU fluorometer according to (Welschmeyer, 1994).

High Performance Liquid Chromatography (HPLC)

Duplicate samples (n = 2) were concentrated onto Whatman GF/F filters and extracted in cold 100% acetone by grinding with a teflon-tip grinding rod. After centrifugation, the supernatant was mixed with 20% (v/v) HPLC-grade water and analyzed on a Shimadzu LC10-AD HPLC system (Wright et al., 1991). Pigment concentrations were based on absorption at 440 nm (Dynamax Model UV-1). Integrated HPLC peak area was quantified with external standards. Canthaxanthin was used as an internal standard. A spectral diode detector provided spectral absorption of peaks which were used to confirm the individual pigments by reference to a spectral library of standards. Duplicate samples (N = 2) differed by less than 2% of each other on replicate filtered samples. Note that there was a small difference (2–20%) between HPLC samples and fluorometric samples. For this reason, we have averaged the HPLC and fluorometric samples.

ChI-Specific Absorption Coefficient $(a_{ph}^{*}(\lambda))$

For experiments at -1.5, 2, and 4°C, *in vivo* whole cell absorption spectra were determined (n = 4) on concentrated samples from 300 to 800 nm at 1 nm intervals with an integrating sphere accessory in a dual beam Perkin Elmer Lambda 6 UV/Vis spectrophotometer. Fresh f/2 medium was used as a reference and a blank. For cultures grown at 0°C, the integrating sphere failed and we estimated the absorption coefficient with the glass fiber filter method of Mitchell (1990) which has been shown to agree well with the integrating sphere methods after corrections for pathlength amplification (Mitchell et al., 2000). The chlspecific absorption coefficient was estimated by dividing loge absorption, aph (λ), by the corresponding chl *a* value:

$$a_{\rm ph}^*(\lambda) = a_{\rm ph} \left(\lambda\right) \left[{\rm chl} \, a \right]^{-1}. \tag{2}$$

Particulate Carbon and Nitrogen Content

Samples (n = 3) were filtered onto pre-combusted GF/F filters, stored at -20° C and run on a Carlo Erba carbon analyzer. Samples of acetanilide were used to validate the system's calibration. See Moisan and Mitchell (1999) for details.

Cell Concentrations

Cell concentrations were estimated in Palmer Maloney chambers both fresh (unfixed) and observed at a magnification of 400X. See Moisan and Mitchell (1999) for details.

Quantum Yield for Growth

Quantum yield for growth was based on the carbon-specific net growth rate, whole cell *in vivo* spectral absorption, and the spectral irradiance in each treatment (Sosik and Mitchell, 1991; Moisan and Mitchell, 1999) and is a rearrangement of Equation (1a).

Non-linear Curve Fitting

Coefficients for P-E curves, and temperature and light dependent equations for growth and quantum yield for growth were fit using a Marquardt-Levenberg non-linear least squares minimization routine (Marquardt, 1963). See Moisan and Mitchell (1999) for details.

RESULTS AND DISCUSSION

The Relationship between Growth Rate, Temperature, and Light

Growth rates of *P. antarctica* varied ~8-fold ranging from 0.04 day⁻¹ to 0.34 day⁻¹ under conditions which were limiting to inhibiting for growth (**Figure 1A**). Specific net growth rates (μ_n) increased in response to light and approached photoinhibition at higher irradiances (**Table 2**). The initial slope of the light-dependent growth was lowest at 4°C (0.007 average± S.E. 0.001) and highest at 0°C (0.030 average ± S.E. 0.006). A photoacclimation parameter ($E_{k\mu}$) was derived utilizing a curve-fitting scheme of all of the light treatments for a single temperature and other studies. Values of $E_{k\mu}$ were both temperature- and light- dependent (**Figure 1A**).

Relative growth rates (μ_n/μ_m) for our treatments exhibited a strong dependence on temperature, with maximal values between 0 and 2°C except at photoinhibiting intensities (**Figure 1B**, Tilzer and Dubinsky, 1987). In general, μ_n had a parabolic response to temperature, with a 1.5-fold change between minima and maxima with the exception of the treatments at 14 µmol quanta $m^{-2} s^{-1}$. At low temperatures under moderate light intensities, *P. antarctica* achieved maximal growth similar to those grown at high light at 4°C. For example, the maximal growth rates, μ_m (0.98 ± 0.05, average ± S.E.) achieved at 37 µmol quanta $m^{-2} s^{-1}$ at temperatures <2°C were lower than those observed at inhibiting growth conditions at 542 µmol quanta $m^{-2} s^{-1}$ at 4°C (**Figure 1A**, *t*-test; p > 0.05).

Empirical Model of Growth Rate

The systematic dependence of net growth rate (μ_n) on temperature and light allowed for the derivation of a simple empirical model for μ_n (Figure 1C). We modeled the light-dependence of growth with a hyperbolic tangent equation commonly used for photosynthesis-irradiance relationships (Platt et al., 1980, Figure 1A) which allows μ_n to increase with irradiance and become saturated or photoinhibited at higher irradiances (Table 3). We modeled μ_m as a function of temperature (Eppley, 1972). Values of α were fit as linear functions of temperature and light ($R^2 = 0.85$) as has been observed for Antarctic phytoplankton (Lizotte and Sullivan, 1991). Net growth for *P. antarctica* can be estimated with knowledge of light and temperature using the following equations,

$$\mu_{\rm m} = \log(2)0.851^* (1.066^{\rm T}), \tag{3a}$$

$$\alpha = \alpha 1 + (\alpha 2T), \tag{3b}$$

$$\mu_{n} = \mu s^{b} [1 - \exp(\frac{-\alpha PAR}{\mu s^{b}})] \exp(\frac{-\beta PAR}{\mu s^{b}})$$
(3c)

$$\mu s^{b} = \frac{\mu_{\max}[\alpha + \beta/\alpha]}{[\beta/\alpha + \beta]^{\beta/\alpha}}$$
(3d)

where μ_n is net growth rate, μ_m is the maximal growth rate (day⁻¹), T is temperature (°C), α is the initial slope of the μ_n vs. irradiance relationship, β is the inhibition of growth at high irradiance, $\alpha 1$ and $\alpha 2$ are the y-intercepts and slope for α with temperature, respectively, μs^{b} (analogous to P_{sb} in Platt et al., 1980) is a fitting parameter and PAR is photosynthetically available radiation (350-700 nm). Equations (3c,d) represent a growth equation originally derived by Platt et al. (1980) as a function of P vs. I. parameters. The values for $\alpha 1$ and $\alpha 2$ are 0.0069 \pm 0.001 and (-0.001 \pm 0.0003), respectively (p < 0.00001). We have very few treatments where irradiance was sufficient to inhibit growth. However, other investigators have clearly documented that phytoplankton growth is inhibited at higher irradiances (Goldman and Carpenter, 1974; Laws and Bannister, 1980; Platt et al., 1980). Therefore, we allow β to be constrained in a relatively large range to accomodate photoinhibition at high light under low temperature. These dynamics are in general agreement with increases in photoinhibition of short-term photosynthetic carbon fixation at high light (Platt et al., 1980).



rate (μ_n/μ_m) with respect to temperature for irradiances in units of μ mol quanta m⁻² s⁻¹. (C) An empirically-derived model for growth rate based on combined limitation of light and temperature. The continuous surface is the model prediction and discrete points are experimental observations. (D) Modeled vs. observed μ (day^{-1}) based on Equation (4).

This simple empirical model has an excellent ability for predicting μ_n based only on light and temperature as input (**Figure 1D**; $r^2 = 0.89$, p < 0.05). However, it does not include explicit information about mechanisms of physiological acclimation. Therefore, we have also developed a model to predict growth using a series of nested models that predict the biophysical parameters, a_{ph}^* (λ) and ϕ_{μ} , based on acclimation to temperature and light.

Temperature and Light Dependence of a_{ph}^{*} (λ): Observations and Modeling of Variability

The *in vivo* whole cell chlorophyll-specific absorption properties, a_{ph}^* (λ), of *P. antarctica* varied with both irradiance and temperature (**Figures 2A,B**). Values of a_{ph}^* (λ) increased with increasing irradiance (**Figures 2C–E**). Values of a_{ph}^* (λ) at 436 and 676 nm generally increased with decreasing temperature but the magnitude of the temperature dependence was less than the irradiance dependence (**Figures 2C,D**, ANOVA p > 0.05). Values

of a_{ph}^* (330), due to mycosporine-like amino acids (Moisan and Mitchell, 2001), had less dependence on temperature than estimates at 436 and 676 nm (**Figure 2E**). Our observations are within the range of a_{ph}^* (λ) observed during *Phaeocystis*dominated blooms (Cota et al., 1994; Arrigo et al., 1998; Reynolds et al., 2001).

Pigment package effects, changes in pigment ratios (**Table 2**), and thylakoid stacking played an important role in determining both the spectral shape and large changes in the magnitude of a_{ph}^* (λ) (Moisan and Mitchell, 1999; Moisan et al., 2006). Ratios of a_{ph} (330) to a_{ph} (676) increased linearly from 1.8 to 13 due to changes in MAA to Chl *a* concentrations as a photo-protective response to increased irradiance (Moisan and Mitchell, 2001).

We explored the dependence of a_{ph}^* (λ) on irradiance and temperature in an attempt to specify a simple parameterization for spectral absorption that could accurately capture the changes in both shape and magnitude caused by changes in pigment ratios and packaging. Singly, a_{ph}^* (676) was highly predictable by a linear regression with either irradiance ($r^2 = 0.62$, **Figure 3A**) or temperature ($r^2 = 0.63$, data not shown). However, the accuracy

Irradiance μmol quanta m ⁻² s ⁻¹	μ	Chl <i>a</i> Pg (cell) ⁻¹	Ratio to HPLC Chl a					Total Pigment Per	PS/total Pig	PP/ total Pig of	
			Chl (c1+c2+c3)	19'hex	Fuco	19'but	β-car	DD+DT	cell	(w/w)	total (w/w)
$T = 4^{\circ}C$											
14	0.04	0.38	0.415	0.609	0.008	ND	0.003	0.015	0.76	0.99	0.01
37	0.12	0.45	0.407	0.640	0.031	ND	0.003	0.016	0.92	0.99	0.01
85	0.19	0.45	0.307	0.268	0.010	ND	0.010	0.070	0.82	0.95	0.05
151	0.24	0.35	0.325	0.368	0.012	ND	0.011	0.099	0.60	0.94	0.06
259	0.25	0.37	0.380	0.386	0.009	ND	0.011	0.077	0.69	0.95	0.05
400	0.34	0.28	0.264	0.298	0.042	ND	0.016	0.120	0.50	0.92	0.08
542	0.29	0.17	0.204	0.230	0.015	ND	0.028	0.371	0.38	0.78	0.22
$T = 2^{\circ}C$											
14	0.07	0.10	0.503	0.725	0.217	0.010	0.006	0.038	0.22	0.98	0.02
37	0.27	0.31	0.528	1.254	0.076	0.028	0.005	0.053	0.74	0.98	0.02
85	0.34	0.36	0.519	1.585	0.052	0.035	0.010	0.189	1.04	0.94	0.06
106	0.30	0.30	0.489	1.529	0.057	0.039	0.012	0.199	0.86	0.94	0.06
214	0.31	0.06	0.525	1.624	0.070	0.039	0.010	0.183	0.17	0.94	0.06
$T = 0^{\circ}C$											
14	0.06	0.26	0.508	1.305	0.061	0.019	0.005	0.085	0.65	0.97	0.03
37	0.28	0.37	0.511	1.079	0.079	0.020	0.006	0.065	0.85	0.97	0.03
85	0.28	0.41	0.473	0.933	0.023	0.025	0.009	0.097	1.15	0.96	0.04
106	0.33	0.34	0.470	0.834	0.027	0.023	0.006	0.087	0.65	0.96	0.04
214	0.21	0.05	0.402	0.797	0.022	0.017	0.011	0.180	0.11	0.92	0.08
T = −1.5°C											
14	0.05	0.02	0.461	0.529	0.164	0.014	0.004	0.028	0.05	0.99	0.01
37	0.17	0.03	0.529	1.425	0.048	0.044	0.009	0.159	0.77	0.95	0.05
85	0.20	0.27	0.533	1.528	0.051	0.042	0.008	0.135	0.79	0.96	0.04
106	0.25	0.30	0.539	1.340	0.047	0.042	0.008	0.133	0.96	0.95	0.05
214	0.16	0.10	0.475	1.386	0.038	0.046	0.012	0.321	0.26	0.90	0.10

TABLE 2 | Growth and pigmentation of *P. antarctica* grown under nutrient-replete conditions.

Photosynthetic pigments (PS) include Chlorophyll (Chl), 19' hexanoyloxyfucoxanthin (19' hex), and fucoxanthin (Fuco). Photo-protective pigments (PP) include diadinoxanthin (DD), β -carotene (β -car), and diatoxanthin (DT). Chl a values are the average of HPLC and fluorometric (Fluor) values. HPLC values (N = 2) were within 2% of each other.

of modeled estimates of $a_{ph}^{*}(\lambda)$ was improved by using a multiple linear regression with light and temperature,

$$a_{\rm ph}^*(\lambda) = C_{\rm o} + C_{\rm E} PAR + C_{\rm T} T, \qquad (4)$$

where PAR is in units of μ mol quanta m⁻² s⁻¹, temperature (T) is in °C, Co is the intercept of the multiple linear regression, C_E and C_T are the multiple linear regression coefficients for PAR and T. The multiple regression fit for a_{ph}^{*} (676) for temperature and light had an $r^2 = 0.82$ (Figure 3B), a considerable improvement compared to the fit for only irradiance $(r^2 = 0.62, \text{ data not shown})$. A comparison of modeled (10 nm resolution) and observed (1 nm hyperspectral) $a_{\text{ph}}^{*}\left(\lambda\right)$ are shown in Figures 3C,D. Our modeled values are in excellent agreement with measured $a_{ph}^{*}(\lambda)$ in the visible region. Variability in a_{ph}^{*} (λ) between 300 and 400 nm is dominated by irradiance (Moisan and Mitchell, 2001) with negligible dependence on temperature (data not shown). For consistency, a single parameterization based on PAR and T is used for the full spectrum (300-700 nm). Coefficients for the model based on light and temperature are given in Table 4. The model does an excellent job of describing a_{ph}^{*} (λ) as illustrated for an extreme range of acclimation states representing high (**Figure 3C**) and low (**Figure 3D**) degrees of pigment packaging and changes in photosynthetic and photoprotective pigments relative to chlorophyll *a*. High relative error in the UV for the high light sample (**Figure 3C**) is caused in part because a_{ph}^{*} (330) is ~10 times smaller than the low light example (**Figure 3D**) caused by changes in cellular concentrations of mycosporine-like amino acids relative to chlorophyll a (Moisan and Mitchell, 2001). There was insignificant irradiance in the system below 350 nm so the large variance in the a_{ph}^{*} (λ) model for data between 300 and350 nm did not affect the estimates of ϕ_{μ} or μ_{n} ; see section Performance of the Biophysical Model of Net Growth for *P. antarctica* for dertails on model application and **Figures 5C,D** for results modeling ϕ_{μ} and μ_{n} .

Temperature and Light Dependence of the Quantum Yield for Growth: Observations and Modeling

The quantum yield for growth, ϕ_{μ} , is the molar ratio of carbon fixed for net growth to the quanta absorbed. Understanding

TABLE 3 | Growth fitting parameters of *Phaeocystis* grown between -1.5 and 4° C.

Temperature °C	αμ	μnm	Ε _k μ	
-1.5	0.026 ± 0.003	0.95 ± 0.06	37	
0	0.030 ± 0.006	0.99 ± 0.01	34	
2	0.026 ± 0.006	1.06 ± 0.12	41	
4	0.007 ± 0.001	0.93 ± 0.06	128	

how φ_{μ} varies as a function of environmental conditions that are easily measured either in situ or remotely is critical in the parameterization of photosynthetic bio-optical models that have ecological relevance. We have chosen to estimate ϕ_{μ} in relation to light and temperature because they limit bloom formation, control cellular absorption, and enzymatic activity. Although phytoplankton may exhibit chromatic adaptation that optimizes the absorbing pigments to selectively absorb the modified under water light spectrum (Dubinsky and Stambler, 2009), the quantum yield of photosynthesis (oxygen evolved or carbon fixed per photon absorbed) is relatively flat from about 400-650 nm (Luning and Dring, 1985). In the classic review Photosynthesis by Rabinowitch and Govindjee (1969) they show the spectral quantum yield determined by Emerson and Lewis (1943) for Chlorella pyrenoidosa ranged from 0.07 to 0.09 over the spectral range 400-680 nm with the lowest values in the area of photoprotective carotenoids 480-510 nm. While this classic work shows very modest spectral variation in the spectral quantum yield of $\pm 15\%$ from 400 to 680 nm, there are other articles that show larger variations more like $\pm 30\%$, including our work on the chlorophyll-a fluorescence yield of P. antarctica (Moisan and Mitchell, 2001) with a similar dip in the photoprotective carotenoid band and very low yields below 350 nm where mycosporine amino acids absorb. The work by (Luning and Dring, 1985) for 4 genera of macro algae indicated quantum yield varied by $\pm 15\%$ from 400 to 670 nm. There is spectral variation in the quantum yield that we did not determine in this study, but for the PAR region below 680 nm the variations are modest, depending on which study is cited.

We spectrally quantified total photon absorption and total carbon growth so we do not have resolution of the spectral dependence of φ_{μ} . The quantum yield for growth at steady state for experimental samples was estimated by rearranging Equation (1a),

$$\phi_{\mu} = \frac{\mu_n}{Chl : C \int_{350nm}^{700nm} a_{ph}^*(\lambda) E_0(\lambda) d\lambda}.$$
(5)

For our study, ϕ_{μ} varied by 30-fold (**Figure 4A**) ranging from 0.003 to 0.09 mol C (mol quanta absorbed)⁻¹. Values of ϕ_{μ} were highest at low irradiances and decreased at high light intensity but with a large difference in the relationship with light for different temperatures. For a given light level, ϕ_{μ} generally decreased with decreasing temperature (**Figure 4A**); the curve fitted to the 4°C was consistently higher than all other temperatures. Values of ϕ_{μ} for the 0 and 2°C treatments were similar to each other for the corresponding light treatments

(*t*-test, p > 0.05). Our values for ϕ_{μ} are in the range previously reported for *Phaeocystis*-dominated blooms (SooHoo et al., 1987; Cota et al., 1994; Carder et al., 1995; Vaillancourt et al., 2003).

The quantum yield for net photosynthesis has previously been modeled as a product of the maximal quantum yield for net photosynthesis, ϕ_m and a Poisson probability function that an open photosynthetic unit will be hit. This model was originally formulated and validated with experiments on O₂ evolution by short-term flashes of light (Dubinsky et al., 1986; Falkowski and Raven, 2007) and adopted to model the quantum yield for growth (Sakshaug et al., 1989; Cullen, 1990). Typically, ϕ_m has been set for all acclimation states to the theoretical maximal value which is equal to 0.125 mol C (mol quanta absorbed)⁻¹ (e.g., Sakshaug et al., 1989). However, as reviewed by Laws et al. (2002), absorption by non-photosynthetic pigments will automatically result in quantum yields of photosynthesis less than the theoretical maximum. To address this important modeling issue, Moisan and Mitchell (1999) proposed a novel approach where the maximal quantum yield (ϕ_m) for each acclimated state was dependent on the irradiance. Here, we extend this concept of an environmentally controlled maximal quantum yield by also parameterizing it based on temperature which controls the thermodynamics of photosynthetic enzymes and therefore affects the quantum yield for growth. Thus, as we have shown for a_{ph}^* above, here we present a mechanistic control of ϕ_m based on co-limitation by light and temperature.

We found lower quantum yield for growth values at low light (<20 μ mol quanta m⁻²s⁻¹) whereas the cumulative hit Poisson Probability does not allow this roll off at low light. Johnsen and Barber (2003)have shown excellent experimental data that demonstrates that there are different photophysiological mechanisms at reduced light levels which contribute to a reduced quantum yield. Despite low quantum yields for growth and photosynthesis, there is experimental evidence that photosynthetic efficiency is not always maximal at low irradiances (Johnsen and Barber, 2003). Low photosynthetic quantum yield has been attributed to the Kok effect, S-state decay (S3 to S2), and imbalances in photosystem excitation (Diner and Mauzerall, 1973). Furthermore, it is expected that the ratio of respiration to photosynthesis may increase at very low light, leading to the observed reduction in quantum yield at very low light.

Given observed values of φ_{μ} , a function for $\varphi_{mEo,T}$ was derived which exponentially decreased from low to high irradiances (**Figure 4B**). The equations used to predict $\varphi_{mEo,T}$ are given below and the final estimate of φ_{μ} compared to observed values are plotted in in **Figure 5C**. Values of $\varphi_{mEo,T}$ were described as a function of irradiance (**Figure 4B**) and the temperature dependence was incorporated into the coefficients a and b which are modeled as linear functions of temperature (**Table 5**),

$$\phi_{mEo,T} = ae^{-bEo},\tag{6a}$$

$$a = 0.0125T + 0.058$$
 $r^2 = 0.70$ (6b)

$$b = -0.0010T + 0.0075 \quad r^2 = 0.85 \tag{6c}$$





where T is temperature in $^\circ C$ and E_o is in units of μ mol quanta $m^{-2}~s^{-1}.$ Values of a and b which are in Equation (6a) are linear relationships of parameters that rely on temperature. Values of $\varphi_{mEo,T}$ and for $Ek\mu$ as estimated below were introduced into Equation (1b) to estimate $\varphi_{\mu}.$

Performance of the Biophysical Model of Net Growth for *P. antarctica*

To test the performance of the biophysical model for net photosynthesis, we predicted $a^*_{ph}~(\lambda)$ and φ_μ using the

temperature and light dependent model parameterizations as described above. Overall, the predicted values of a_{ph}^* (λ) underestimate the observed values by about 10% (**Figures 3C,D**, **5A**) with greatest accuracy achieved in the region between 350 and 700 nm. Relatively more scatter was seen in the UV region between 300 and 350 nm. **Figure 5B** is a compilation of net growth normalized to the maximum value of the experiment vs. irradiance for the data presented here, Moisan and Mitchell (1999), and data from the literature for the genus *Phaeocystis*. To estimate an $E_{k\mu}$ value to introduce into Equation (1b), we used a non-linear curve fit (Platt et al., 1980) for the full data synthesis in



FIGURE 3 | Modeled vs. observed values of a_{ph}^* at 676 nm based on a linear regression using (A) light and (B) a multiple linear regression based on temperature and light. Comparison of a_{ph}^* (λ) observed (1 nm resolution continuous line) vs. modeled values (10 nm resolution) based on temperature and light for cultures grown at (C) 4°C 542 μ mol quanta m⁻² s⁻¹ and -1.5°C and (D) 37 μ mol quanta m⁻² s⁻¹. Note scale difference in a_{ph}^* in (C,D). Standard error bars are shown for model estimates.

Figure 5B. This fit resulted in a single value of μ_m and α for the full data set in **Figure 5B** that we used to estimate a generalized value of $E_{k\mu}$ (μ_m/α) for *Phaeocystis* spp. This resulted in a $E_{k\mu}$ value of 110 μ mol quanta m⁻² s⁻¹ overall fit with $r^2 = 0.83$ (**Figure 5B**). We also explored modeling using $E_{k\mu}$ determined from each individual growth vs. irradiance data set at the different temperatures (**Figure 1A**) but that did not improve the overall variance in our biophysical model so we kept the model simpler by using one generalized value determined from our data and literature synthesis (**Figure 5B**).

We predicted values of ϕ_{μ} using a cumulative one hit Poisson probability parameterization where $\phi_{mEo,T}$ in Equation (1b) is predicted from temperature and light (Equation 5). The ϕ_{μ} model also requires an input variable for $E_{k\mu}$. **Figure 5B** combines data for *Phaeocystis* spp. presented in Moisan and Mitchell (1999) and the new data for -1.5, 0, and 2°C treatments that are presented here. Using the combined data set, we have derived an $E_{k\mu}$. Our sensitivity analysis shows that the ϕ_{μ} is the most sensitive to $\phi_{mEo,T}$ as compared to $E_{k\mu}$. We prefer the use of a composite $E_{k\mu}$ at this time although in principle it should depend on environmental acclimation, in this case the steady-state conditions we imposed in our experiment. Although we acknowledge that in nature both light and temperature may fluctuate on time scales faster than division rates and hence impact acclimation, our experiments were not designed to resolve that additional, and potentially important, variability. We conducted a sensitivity analysis of ϕ_{μ} that indicates that the relative magnitude of the error is greatest at high intensities; however, the absolute error is minor. These assumptions are justified on the basis that the entire equation is dependent on

TABLE 4 Coefficients for multiple-linear regression model for $a_{ph}^{*}(\lambda)$ based on
temperature and light.

λ (nm)	C ₀	CE	CT	Standard error of estimate	R ²
300	0.016	3.25E-04	0.0050	0.000715	0.79
310	0.020	2.96E-04	0.0027	0.000609	0.77
320	0.022	3.84E-04	0.0011	0.000834	0.79
330	0.022	5.42E-04	0.0007	0.001282	0.83
340	0.022	5.72E-04	-0.0001	0.001190	0.85
350	0.023	4.41E-04	-0.0007	0.000553	0.88
360	0.019	2.07E-04	-0.0002	0.000121	0.88
370	0.016	1.09E-04	0.0008	4.69E-05	0.85
380	0.017	9.93E-05	0.0008	3.71E-05	0.86
390	0.019	1.08E-04	0.0006	3.37E-05	0.89
400	0.021	1.11E-04	0.0004	2.83E-05	0.91
410	0.026	1.19E-04	-0.0004	3.72E-05	0.87
420	0.028	1.24E-04	-0.0006	3.27E-05	0.89
430	0.030	1.29E-04	-0.0010	3.21E-05	0.90
440	0.032	1.32E-04	-0.0011	3.05E-05	0.90
450	0.032	1.21E-04	-0.0013	2.74E-05	0.90
460	0.033	1.14E-04	-0.0014	2.07E-05	0.91
470	0.033	9.90E-05	-0.0015	1.66E-05	0.90
480	0.030	8.94E-05	-0.0015	1.12E-05	0.92
490	0.023	8.21E-05	-0.0012	8.63E-06	0.93
500	0.019	6.05E-05	-0.0010	5.40E-06	0.91
510	0.017	3.95E-05	-0.0011	3.92E-06	0.86
520	0.015	2.77E-05	-0.0011	4.49E-06	0.74
530	0.012	1.87E-05	-0.0010	3.63E-06	0.65
540	0.010	1.12E-05	-0.0008	2.64E-06	0.54
550	0.008	6.36E-06	-0.0007	2.42E-06	0.41
560	0.007	4.36E-06	-0.0006	2.21E-06	0.34
570	0.006	4.46E-06	-0.0005	2.76E-06	0.23
580	0.007	4.16E-06	-0.0005	2.76E-06	0.22
590	0.008	5.02E-06	-0.0006	2.45E-06	0.34
600	0.007	4.29E-06	-0.0004	2.41E-06	0.19
610	0.006	4.70E-06	-0.0003	1.79E-06	0.21
620	0.006	6.50E-06	-0.0004	1.57E-06	0.33
630	0.007	6.18E-06	-0.0004	1.57E-06	0.33
640	0.007	5.83E-06	-0.0003	1.68E-06	0.26
650	0.007	6.40E-06	-0.0002	1.13E-06	0.34
660	0.009	1.23E-05	-0.0003	2.28E-06	0.49
670	0.017	2.79E-05	-0.0008	3.04E-06	0.80
680	0.015	2.57E-05	-0.0009	4.18E-06	0.72
690	0.005	7.53E-06	-0.0004	3.77E-06	0.16
700	0.000	8.76E-07	-0.0004	9.27E-07	-0.03

See Equation (6) for abbreviated terms.

 φ_{max} . We do observe some variability in $E_{k\mu}$ between the lower temperature treatments and the 4°C treatment (Figure 1A). We derived a composite $E_{k\mu}$ from the intersection between the initial slope of the growth curve and the maximal growth rate (Figure 5B). The use of a generic $E_{k\mu}$ is appropriate because of the overall variability in growth curves with respect to irradiance for various *Phaeocystis* spp. Also, a sensitivity analysis of $E_{k\mu}$ within the model revealed that φ_{μ} was relatively insensitive to the changes that we observed in $E_{k\mu}$. Using an $E_{k\mu}$ value of 110 μ mol quanta m⁻² s⁻¹ and a temperature-light dependent model of $\varphi_{mEo,T}$ results in an excellent prediction of



 ϕ_{μ} (Figure 5C, $r^2 = 0.80$). Using these individual component models for a_{ph}^* (λ) and $\phi_{mEo,T}$ in Equations (1a,b), a comparison between observed and predicted values of μ_n (Figure 5D; $r^2 = 0.66$). The simple empirical model using only temperature and light resulted in a slightly better estimate ($r^2 = 0.70$; Figure 1C) however that empirical approach does not include the mechanistic dynamics of changes in a_{ph}^* (λ) and $\phi_{mEo,T}$ that represent how light and temperature co-limitation regulate the cellular physiology.

Although we used a single value of $E_{k\mu}$ a sensitivity analysis proved that using the estimated $E_{k\mu}$ for the different temperatures did not improve our model fit. Also, our quantum yields for growth are a net process of growth after respiration which can be important in nature (Grégori et al., 2002; Siegel et al., 2002; Marra and Barber, 2004; López-Urrutia et al., 2006).



FIGURE 5 | (A) The relationship between modeled a_{ph}^* (λ) based on temperature and light (Equation 6, **Table 4**) and observed values at individual wavelengths. **(B)** Relative growth rates measured from various *Phaeocystis* species grown under a variety of temperature and irradiance regimes. Note that there are overlapping data points. **(C)** Modeled ϕ_{μ} vs. observed values. There are overlapping points. **(D)** Relationship between the observed carbon-specific growth rate and model predictions of carbon-specific rate of net growth (μ_n). Model is based on Equation (1) and **Table 4**.

 $\ensuremath{\mathsf{TABLE 5}}\xspace$] Model variables for maximal quantum yield based on irradiance at discrete temperatures.

a value	b value	r ² value
0.0377	-0.0095	0.90
0.0600	-0.0061	0.93
0.0447	-0.0025	0.09
0.0958	-0.0014	0.32
	0.0377 0.0600 0.0447	0.0377 -0.0095 0.0600 -0.0061 0.0447 -0.0025

Values of φ_m are described as an exponential function of irradiance with the form, $\varphi_{mE_0=}ae^{-bE_0}.$

Our quantum yield for net growth calculations can be defined by:

$$\phi_{\mu_n} = \frac{\mu_n C}{E_{absorbed}} = \frac{P - R}{E_{absorbed}}$$

Although respiration is often modeled as a fixed ratio to growth rate, or a basal a rate plus a fixed fraction of growth rate, this likely is not so simple. There are relatively few data sets in the literature for phytoplankton respiration over a range of growth conditions and we did not estimate respiration for our study. Respiration data that is commonly used in

models is very noisy (Laws and Bannister, 1980; Geider et al., 1998) and is mostly limited to a modest range of intensity from about 50-200 μ mol quanta m⁻² s⁻¹ so the fidelity of respiration models may not be robust over the full range of light intensity that is relevant to phytoplankton communities. One expects that the ratio of respiration to net growth (R/μ_n) will actually go to infinity at the limit as light goes to zero where there is no photosynthesis but basal respiration continues. Also, for the growth inhibiting irradiances greater than about 150 μ mol quanta m⁻² s⁻¹ we observed at lower temperatures (Figure 1A), we speculate that the ratio of R/μ_n also must increase. Models that assume a basal respiration rate plus a fixed fraction of growth rate will not predict this increase in R/μ_n at higher growth inhibiting irradiances but will instead predict lower respiration as growth rate decreases. This seems flawed since inhibiting irradiances would cause damage from oxidative stress and hence very likely a greater respiration rate, not a lower rate. These two thought experiments based on simple logic indicate that it is important to obtain more data that includes respiration at very low light that and also at high growth inhibiting irradiances, but such data is lacking in the literature.

Our biophysical model includes explicit descriptions of the regulation of absorption and quantum yield by environmental controls which results both in an accurate model of net growth and captures details of light and temperature regulation of the biophysical variables that mediate light utilization. Our quantum yields for growth were calculated in a traditional way of modeling based on carbon, chlorophyll a, phytoplankton absorption, and spectral irradiance flux. However, our estimates are relatively high compared to other phytoplankton; our high C: chl a ratios are above the typical ratio of 75-150 for phytoplankton (Falkowski and Raven, 2007). This is likely due to the fact that Phaeocystis can accumulate a carbohydrate/glucan-rich mucilage that is not metabolically active in addition to metabolically active cellular carbon. Mathot et al. (2000) estimates that mucus-related carbon at peak bloom is about $14 \pm 4\%$ of total colony carbon but mucus-related carbon may be as high as 90% for a senescent bloom (Verity and Medlin, 2003). The colonial matrix acts as a spillover mechanism probably for storage of low molecular weight compounds (Lancelot and Mathot, 1985) and hence may accumulate when there is excess light.

Although natural systems may have substantial variations in both light and temperature over time-scales that are much less than the doubling times that will affect acclimation of algae in a more complicated way than the simple steady-state experiments we conducted, it is very time consuming to try to add that complexity to the matrix of growth and acclimation control. There is a relatively small literature for the steady-state co-limitation of light, nutrients and temperature acclimation of cellular pigments, absorption, quantum yield and growth (Laws and Bannister, 1980; Kiefer and Mitchell, 1983; Sosik and Mitchell, 1991, 1994; Geider et al., 1997). There is even more limited knowledge for cold water obligate psychrophiles like *P. antarctica*, for example the study of Luxem et al. (2017) who explored light and iron limitation for this organism. In the area of applied phycology aimed at commercial production, recent work with a similar experimental design for light and temperature as ours, for the seaweed Gracilaria, Nejrup et al. (2013), reported growth rates and metabolic capacity. Grobbelaar et al. (1996) have explored microalgae acclimated to different light levels, then exposed to light fluctuations that would be found in commercial cultivation. However, neither of these studies quantified spectral absorption or quantum yield, that ultimately underlie the responses for growth and bulk PvsE relationships. Our work advances the state of knowledge with greater detail of how light and temperature combined affect acclimation of the fundamental biophysical parameters light absorption and quantum yield of algae and provides a level of detail not previously reported for the ecologically important prymnesiophyte, Phaeocystis antarctica, that is a key structural component of the Southern Ocean and that may show changes in its role as stratification and heating of the Southern Ocean change both light and temperature climate in the future.

CONCLUSIONS

Understanding the contribution of *Phaeocystis* in the overall global carbon cycle is important because this genus can dominate mesoscale blooms and make a significant contribution to the

carbon and sulfur cycle for open and coastal oceans as well in the vicinity of ice edges. The strain of P. antarctica we used for these experiments exhibits flexibility in its ability to absorb light due to changes in cell size and the cellular concentrations of photoprotective and photosynthetic pigments that are regulated by light and temperature co-limitation. These cellular changes have consequences for how P. antarctica captures light required for carbon fixation. Photo-protective pigments that divert absorbed photons from the photosynthetic reaction centers will automatically result in photosynthetic quantum yields below the theoretical maximum. Using data from our controlled laboratory experiments, we have developed both empirical and biophysical models that predict growth rates based on knowledge of chlorophyll a, light, and temperature. It is important to note that unlike some proposed models that depend only on easily measured variables our biophysical model, $a_{ph}^{*}(\lambda)$ and ϕ_{μ} is parameterized based on light and temperature since nutrients are not regulating during the onset and development of blooms. We propose a conceptually important improvement over previous models by accepting the fact that the maximum quantum yield for any acclimated state may be considerably less than the theoretical maximum (Moisan and Mitchell, 1999) and that it should be regulated by environmental control, in this case light and temperature co-regulation. In general, a_{ph}^{*} (\lambda), E_{km} and $\varphi_{\mu},$ should all be modeled as variables determined by the simultaneous interaction of light, temperature, and nutrients. Furthermore, future ocean color missions with hyperspectral ability, such as PACE, may be capable of remote sensing phytoplankton spectral shape that we document changes significantly with different acclimation states (Figure 2). Such a capability might allow a link between community structure, photosynthetic processes and the biophysical model for absorption, cellular pigmentation and pigment packaging that is dependent on light and temperature co-limitation (see also Wang et al., 2016; Moisan et al., 2017). While this level of detail remains aspirational for our community, an accurate understanding of how biophysical variables (Equation 1) depend on the full matrix of environmental forcing requires these essential but timeconsuming laboratory studies for representative taxa over the full range of environmental variability experienced by phytoplankton to advance our ability to apply advanced remote sensing methods to coastal processes of primary production and the carbon cycle.

AUTHOR CONTRIBUTIONS

TM: Supervised all experiment, sample analysis and data analysis; TM and BGM: Contributed equally to writing, editing, graphics, and tables.

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Conflict of Interest Statement: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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