Flexible C : N ratio enhances metabolism of large phytoplankton when resource supply is intermittent

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Abstract. Phytoplankton cell size influences particle sinking rate, food web interactions and biogeographical distributions. We present a model in which the uptake, storage and assimilation of nitrogen and carbon are explicitly resolved in different-sized phytoplankton cells. In the model, metabolism and cellular C : N ratio are influenced by the accumulation of carbon polymers such as carbohydrate and lipid, which is greatest when cells are nutrient starved, or exposed to high light. Allometric relations and empirical data sets are used to constrain the range of possible C : N, and indicate that larger cells can accumulate significantly more carbon storage compounds than smaller cells. When forced with extended periods of darkness combined with brief exposure to saturating irradiance, the model predicts organisms large enough to accumulate significant carbon reserves may on average synthesize protein and other functional apparatus up to five times faster than smaller organisms. The advantage of storage in terms of average daily protein synthesis rate is greatest when modeled organisms were previously nutrient starved, and carbon storage reservoirs saturated. Small organisms may therefore be at a disadvantage in terms of average daily growth rate in environments that involve prolonged periods of darkness and intermittent nutrient limitation. We suggest this mechanism is a significant constraint on phytoplankton C : N variability and cell size distribution in different oceanic regimes.

1 Introduction

Through its influence on resource acquisition (Pasciak and Gavis, 1974), growth (Tang, 1995) and food web interactions (Armstrong, 1994), organism size is thought to play a major role structuring marine plankton communities (Chisholm, 1992). A few primary productivity (PP) algorithms (Kameda and Ishizaka, 2005; Hirata et al., 2008; Uitz et al., 2008; Brewin et al., 2010) and several other oceanic ecosystem models (e.g., Blackford et al., 2004; Le Quere et al., 2005) resolve phytoplankton traits as a function of cell size. Hence, there is a need to understand how metabolism and physio- physiological traits scale with organism size.

Due to their relatively high surface area to volume ratio, small cells are thought to be superior competitors for nutrients in oligotrophic environments (Chisholm, 1992; Clark et al., 2013). Furthermore, pigment packaging in large organisms can lead to a reduction in light absorption per unit chlorophyll (Morel and Bricaud, 1981), again conferring an advantage to smaller organisms. The prevalence of large organisms in eutrophic ecosystems is usually explained by enhanced resilience to predation (e.g., Ward et al., 2012), and greater nutrient storage capacity. For example, using a Droop model of algal growth in a model chemostat, Verdy et al. (2009) showed the positive influence on growth of a large internal storage reservoir. Furthermore, Grover (1991a, 1991b, 2011) and Tozzi et al. (2004) have demonstrated the benefit of enhanced storage capacity in environments with
Infrequent nutrient pulses. In general, studies that have assessed the ecological advantage of storage have tended to focus on the benefits associated with an enhanced capacity to store nutrients such as nitrogen, phosphorus and iron. Yet, at high latitude, where there is low average surface irradiance and relatively deep mixing (Fig. 1), phytoplankton growth is likely to be light limited.

With sufficiently high irradiance, many phytoplankton species can accumulate large stores of carbohydrate and lipid (Granum et al., 2002). In darkness, these reserves may be drawn upon both as a source of energy to fuel metabolism, and as a source of organic carbon to incorporate into proteins and cell structure. Vertical mixing and the diurnal cycle cause phytoplankton to regularly experience prolonged exposure to chronically low irradiance or darkness (Dubinsky and Schofield, 2010). Therefore, the ability to store carbon may be critical to survival, and may also be a vital ecological strategy when growth maximization determines fitness.

Storage of carbon in the form of carbohydrate and lipid has a significant influence on the phytoplankton C : N ratio (Geider and La Roche, 2002). Eukaryotic autotrophs such as diatoms and coccolithophores often accumulate significant C reserves under N or P stress, when the ability to fix carbon (and thus to store energy), exceeds rates of protein synthesis (Geider and La Roche, 2002). It is not uncommon for eukaryotes to possess C : N ratios more than double the Redfield C : N (Caperon and Meyer, 1972). Bloom-forming eukaryotes with flexible C : N are extremely prevalent at high latitude, where resource supply is relatively variable (Fig. 1).

Small cell size has been emphasized as a factor contributing to the dominance of picoplankton in oligotrophic waters, because small cells with high surface area to volume ratios have reduced transport-limitation of nutrient uptake, (Clark et al., 2013). Yet, small cell size may also prevent the accumulation of large carbon reserves. If small cell size places a limit on the capacity of organisms to store carbon, they may have relatively narrow ranges of C : N. However, this may not be a problem given the less variable conditions in stratified oligotrophic waters, where the build-up and mobilization of carbon reserves may be less of a constraint on growth rate.

We use an empirically constrained phytoplankton growth model to understand how storage capacity influences growth in environments with intermittent nitrogen supply and photon flux density (PFD). The model uses published allometric relations to constrain the capacity for storage. We begin with an overview of the mathematical relations used to constrain growth and go on to describe the theory and experimental data sets used to constrain model parameters. We demonstrate that the model can be constrained to fit observations of organisms in balanced growth. Finally, we report the influence of cell size and carbon storage on the ability of cells to grow when PFD and nitrogen supply are intermittent, and discuss potential implications of our results for the distribution and biogeochemistry of marine phytoplankton.

2 Methods

2.1 Model overview

The model (Fig. 2) is designed to mechanistically capture intracellular dynamics of nitrogen and carbon using simple, previously established mathematical relations. Photosynthesis and uptake are responsible for additions to internal storage reservoirs of carbon and nitrogen, respectively. Photosynthesis is parameterized with a photoacclimation model that allows allocation to light-harvesting proteins to vary dynamically in response to ambient irradiance conditions. Nitrogen is assumed to enter a subcellular reserve pool as a Michaelis–Menten function of the surrounding substrate concentration. Reserve nitrogen and carbon are converted into proteins via the cell’s biosynthetic apparatus. Protein
synthesis only ceases when internal reserves of nitrogen or carbon are depleted, and reserves only accumulate when either photosynthesis or uptake exceed protein synthesis. Thus, variations in cellular C:N ratio arise when there is an imbalance between photosynthesis, nutrient uptake and the synthesis of functional apparatus.

Allometric relations that constrain nutrient uptake, storage capacity and light absorption were used to parameterize the model. Remaining parameters were tuned to empirical data sets for organisms spanning an appropriate size range. This section contains a detailed overview of the model equations, and a description of the allometric relations and empirical data sets used to constrain parameter values.

2.2 Model equations

The model explicitly resolves intracellular reserve pools of compounds that contain either nitrogen or carbon, but not both (Fig. 2 has a model schematic, and Table 1 has all parameter definitions and units). The reserve nitrogen pool is assumed to consist only of NO$_3^−$. The reserve carbon pool may contain any monosaccharides, non-structural polysaccharides and non-structural lipids. These reserve pools serve as input reservoirs of nitrogen and carbon to a mixed pool. The mixed pool contains all “functional” cellular apparatus that regulate metabolism. It may include, but is not limited to, proteins, pigments, nucleic acids, amino acids and structural lipids. The following four equations parameterize growth in terms of these intracellular pools:

$$\frac{1}{N_F} \frac{dN_R}{dt} = V_n - \mu$$

$$\frac{1}{N_F} \frac{dC_R}{dt} = P_n - \left( \frac{1}{\eta} + \zeta \right) \mu - R_0$$

$$\frac{1}{N_F} \frac{dN_LH}{dt} = \rho_{LH} \mu - F_{LH} R_0.$$

The reserve nitrogen and carbon pools are denoted $N_R$ and $C_R$, respectively. Although here $N_F$ denotes the nitrogen content of the functional pool, we impose a fixed stoichiometry on this pool, so that functional nitrogen and carbon may be related with $N_F = \eta C_R$ where $\eta$ is the imposed N:C ratio in g N (g C)$^{-1}$. The light-harvesting apparatus, denoted here $N_{LH}$, are part of the functional pool, but are nonetheless modeled with a separate state variable (Eq. 4). The synthesis of light-harvesting apparatus is regulated with the function $\rho_{LH}$ (see below), to simulate variations in nitrogen allocation that occur during photoacclimation (McKew et al., 2013). Losses associated with the carbon and energy costs of basal metabolism are encapsulated with the fixed parameter $R_0$. Each term on the right-hand side of Eqs. (1) to (4) is now described in detail. Note that all parameter definitions and units may be found in Table 1.

Inorganic nitrogen (denoted here $S$, for “substrate”) first enters the reserve pool via a Michaelis–Menten style parameterization of uptake:

$$V_n(S, N_R, N_F) = V_m \frac{S}{S + K_S}.$$

In Eq. (5), $V_m$ and $K_S$ are the maximum uptake and half-saturation coefficients of the Michaelis–Menten relationship, respectively. The maximum rate of nitrogen uptake is a linearly decreasing function of the internal nitrogen reserve (e.g., Thingstad, 1987):

$$V_m(N_R, N_F) = \left( 1 - \frac{N_R}{N_R^{max}} \right) V_m.$$

Carbon fixed via photosynthesis enters the reserve pool via the following, photosynthesis—irradiance relationship:

$$P_n(E, N_{LH}, C_R, C_F) = P_m \left( 1 - \exp \left( - \frac{\alpha F_{LH} E}{P_m} \right) \right),$$

where the maximum rate of photosynthesis is a linearly decreasing function of the internal carbon reserve (see Fig. 3):

$$P_m(C_R, C_F) = \left( 1 - \frac{C_R}{C_R^{max}} \right) P_{max}.$$
Table 1. Parameters and variables with associated units. Where appropriate values were found in the literature, the source is indicated. The half-saturations for biosynthesis, $K_c$ and $K_N$, were assumed here to be small, representing a high turnover of internal reserves (e.g., Hama, 1991). Note that the units of $V_{\text{max}}$ were obtained by dividing the units reported by Litchman et al. (2007) by their units for $Q_{\text{min}}$ (see also Table 3).

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
<th>Value</th>
<th>Units</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>$C_R$</td>
<td>reserve carbon</td>
<td>variable</td>
<td>mmol C m$^{-3}$</td>
<td>–</td>
</tr>
<tr>
<td>$C_F$</td>
<td>functional carbon</td>
<td>variable</td>
<td>mmol C m$^{-3}$</td>
<td>–</td>
</tr>
<tr>
<td>$N_R$</td>
<td>reserve nitrogen</td>
<td>variable</td>
<td>mmol N m$^{-3}$</td>
<td>–</td>
</tr>
<tr>
<td>$N_F$</td>
<td>functional nitrogen</td>
<td>variable</td>
<td>mmol N m$^{-3}$</td>
<td>–</td>
</tr>
<tr>
<td>$S$</td>
<td>substrate concentration</td>
<td>variable</td>
<td>µmol L$^{-1}$</td>
<td>–</td>
</tr>
<tr>
<td>$E$</td>
<td>photon flux density (PFD)</td>
<td>variable</td>
<td>mol photons m$^{-2}$ day$^{-1}$</td>
<td>–</td>
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<tr>
<td>$V_a$</td>
<td>nitrogen uptake rate</td>
<td>variable</td>
<td>day$^{-1}$</td>
<td>–</td>
</tr>
<tr>
<td>$V_{\text{max}}$</td>
<td>maximum nitrogen uptake at $N_R$</td>
<td>variable</td>
<td>day$^{-1}$</td>
<td>–</td>
</tr>
<tr>
<td>$K_S$</td>
<td>nitrogen uptake half-saturation</td>
<td>allometric</td>
<td>µmol L$^{-1}$</td>
<td>Litchman et al. (2007)</td>
</tr>
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<tr>
<td>$F_m$</td>
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<tr>
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<td>mmol C (mol N)$^{-1}$ day$^{-1}$</td>
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<tr>
<td>$e_{LH}$</td>
<td>light absorption</td>
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<td>m$^{-2}$ (mol N)$^{-1}$</td>
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<td>$e_{LH}^c$</td>
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<td>490.0</td>
<td>m$^{-2}$ (mol N)$^{-1}$</td>
<td>Morel and Bricaud (1981)</td>
</tr>
<tr>
<td>$\phi_{LH}$</td>
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<td>0.08</td>
<td>mol C (mol photons)$^{-1}$</td>
<td>Falkowski and Raven (2007)</td>
</tr>
<tr>
<td>$\gamma$</td>
<td>taxonomic initial slope factor</td>
<td>see Table 4</td>
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<td>–</td>
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<tr>
<td>$F_{LH}$</td>
<td>fraction of cellular nitrogen allocated to light harvesting</td>
<td>variable</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>$\rho_{LH}$</td>
<td>fraction of cellular nitrogen allocated to synthesis of light-harvesting apparatus</td>
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<td>–</td>
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<tr>
<td>$P_{\text{max}}^L$</td>
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<td>see Table 4</td>
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</tr>
<tr>
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<td>minimum nitrogen allocation to light harvesting</td>
<td>see Table 4</td>
<td>–</td>
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</tr>
<tr>
<td>$F_L$</td>
<td>curvature of allocation to light harvesting</td>
<td>see Table 4</td>
<td>m$^{-2}$ day mol photon</td>
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</tr>
<tr>
<td>$N_L$</td>
<td>Chl : N of light-harvesting apparatus</td>
<td>2.4</td>
<td>g Chl g N$^{-1}$</td>
<td>–</td>
</tr>
<tr>
<td>$K_N$</td>
<td>maximum reserve nitrogen</td>
<td>variable</td>
<td>mmol N m$^{-3}$</td>
<td>–</td>
</tr>
<tr>
<td>$C_R$</td>
<td>maximum reserve carbon</td>
<td>variable</td>
<td>mmol C m$^{-3}$</td>
<td>–</td>
</tr>
<tr>
<td>$s_{\text{tot}}$</td>
<td>maximum reserve nitrogen as fraction of functional pool</td>
<td>0.2</td>
<td>–</td>
<td>Lourenço et al. (1998)</td>
</tr>
<tr>
<td>$K_C$</td>
<td>carbon reserve half-saturation coefficient</td>
<td>0.01</td>
<td>–</td>
<td>–</td>
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<tr>
<td>$K_N$</td>
<td>nitrogen reserve half-saturation coefficient</td>
<td>0.01</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>$\mu_{\text{max}}$</td>
<td>maximum biosynthesis rate</td>
<td>see Table 4</td>
<td>day$^{-1}$</td>
<td>–</td>
</tr>
<tr>
<td>$\zeta$</td>
<td>cost of biosynthesis</td>
<td>3.0</td>
<td>mmol C (mol N)$^{-1}$</td>
<td>Pahlow (2005)</td>
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<tr>
<td>$\eta$</td>
<td>N : C ratio of functional components</td>
<td>0.17</td>
<td>mmol N (mol C)$^{-1}$</td>
<td>Geider and La Roche (2002)</td>
</tr>
<tr>
<td>$R_0$</td>
<td>maintenance respiration</td>
<td>0.01</td>
<td>mmol C (mol N)$^{-1}$ day$^{-1}$</td>
<td>Geider et al. (1998)</td>
</tr>
<tr>
<td>$\alpha_N$</td>
<td>reduction in dark N assimilation</td>
<td>0.59</td>
<td>–</td>
<td>DiTullio and Laws (1986)</td>
</tr>
<tr>
<td>$V$</td>
<td>individual cell volume</td>
<td>see Table 3</td>
<td>µm$^3$</td>
<td>–</td>
</tr>
</tbody>
</table>

In Eq. (7), the initial slope of the photosynthesis–irradiance curve is dependent on the fraction of intracellular nitrogen allocated to light harvesting: $F_{LH} = N_{LH}/N_F$. Because both $N_{LH}$ and $N_F$ are state variables, $F_{LH}$ is a dynamic representation of the cell’s nitrogen allocation to light harvesting. Here we constrain this fraction with the regulatory function $\rho_{LH}$, analogous to the approach of Geider et al. (1997, 1998):

$$\rho_{LH} = F_{LH}^{\text{max}} \frac{1}{1 + F_{LH}^{G} E} \frac{F_{LH}^{\text{min}}}{F_{LH}^{\text{max}}}.$$  

(9)

With Eq. (9), the proportion of newly fixed nitrogen allocated to the synthesis of light-harvesting pigments is a decreasing function of the ambient PFD (see Fig. 4), which enables the investment in light-harvesting apparatus as a function of growth irradiance to be constrained empirically. Thus, the trade-off between nitrogen allocation to light-harvesting and other apparatus such as the photoprotective machinery (Armstrong, 2006; McKew et al., 2013) is not considered in this work.

The flow of resources from reserve pools to the functional pool is parameterized as the minimum between two Michaelis–Menten style functions of the internal reserves:

$$\mu = \min \left\{ \frac{N_{R} / N_{F}}{K_N + N_{R} / N_{F}}, \frac{C_{R} / C_{F}}{K_C + C_{R} / C_{F}} \right\} \mu_{\text{max}}.$$  

(10)

where the reserve concentration is normalized by the concentration of the functional pool. This is an appropriate constraint for situations in which reserves are not significantly
more abundant than enzymes involved in metabolism (Borghans et al., 1996).

There is evidence that dark N assimilation proceeds at a lower rate than in the light (DiTullio and Laws, 1986; Probyn et al., 1996; Ross and Geider, 2009). Reductions in dark N assimilation are assumed to influence \( \mu \) in the following way:

\[
\mu'_{\text{max}} = \begin{cases} 
\mu_{\text{max}} & \text{if } E > 0 \\
\frac{\eta N \mu_{\text{max}}}{V} & \text{otherwise}.
\end{cases}
\]  

(11)

Equation (10) simulates internal conversion of C and N into protein and other functional apparatus. When the flow of carbon and nitrogen into the reserve pool from the surrounding medium is equal to the subsequent rate of removal into the functional pool, the cell is said to be in balanced growth. All data sets used for comparison were of organisms in balanced growth, and so Eq. (10) was treated as the effective growth rate.

When either ambient photons or N supply are limiting, cells are able to draw on at least one internal resource to maintain active metabolism. In such conditions, Eq. (11) is only comparable to the specific growth rate of the non-limiting abiotic resource. It is not directly comparable to the net accumulation of the limiting resource, which is strictly less than \( \mu \).

2.3 Allometry

2.3.1 The package effect

Let \( a^\text{ph}_{\text{pk}} \) denote the theoretical, spectrally integrated absorption cross section of a unit of nitrogen contained in the light-harvesting apparatus (with units \( \text{m}^2\text{mol}^{-1} \)), assuming that the light-harvesting apparatus was in no way influenced by pigment packaging. In other words, it is the absorption cross section of pigment associated with each unit of nitrogen in the light-harvesting apparatus in solution. Furthermore, let \( c_i \) denote the concentration of cellular nitrogen associated with the light-harvesting apparatus (units \( \text{mol N m}^{-3} \)). If \( \eta \) is the N : C ratio of the main functional apparatus and \( V \) is the cell volume, then with knowledge of the cellular carbon quota and the fraction of cellular nitrogen allocated to light harvesting \( (F_{\text{LH}}) \), \( c_i \) may be calculated with

\[
c_i = \left( \frac{F_{\text{LH}} Q^C_{\text{max}} \eta}{V} \right) \times 10^{-24}.
\]  

(12)

Calculations of the package effect require knowledge of the carbon per cell, through the parameter \( Q^C_{\text{max}} \). However, all carbon based variables in the model have units \( \text{mmol C m}^{-3} \) – i.e., they are density units and do not keep track of individual cells. To keep track of a dynamic C cell\(^{-1}\), we would also need to keep track of population cell count. Keeping track of cell count is not difficult. However, there is a very small range in the package effect for individual cells undergoing changes in carbon content. The largest influence of the package effect is between organisms of very different size. Therefore, for simplicity, we represented size-dependent variation in C cell\(^{-1}\) with the fixed parameter \( Q^C_{\text{max}} \).

Following Morel and Bricaud (1981), the actual absorption of pigment packaged within a cell of diameter \( d \) (with units m) with \( c_i \) mols of nitrogen contained in chlorophyll...
(units mol N m\(^{-3}\)) may be calculated with
\[
d_{ph} = \frac{3}{2} d_{ph} \frac{Q(\rho)}{\rho},
\]
where
\[
Q(\rho) = 1 + 2e^{-\rho} \frac{\rho}{\rho^2} + 2(e^{-\rho} - 1)\]
and
\[
\rho = d_{ph} c_{f_1} d.
\]

The initial slope of the photosynthesis–irradiance response curve may then be constrained as a function of cell size, with knowledge of the maximum quantum efficiency of photosynthesis, \(\phi_m\):
\[
\alpha = d_{ph} \phi_m \gamma
\]

Empirical allometric relations suggest the initial slope of the growth–irradiance curve may be negatively correlated with cell size across taxa (Edwards et al., 2014). Yet, there is considerable scatter in the data, probably due in part to different pigment compositions, non-spherical cell shapes, and non-homogeneous intracellular pigment distributions. We include \(\gamma\) in the above relation as a tuning parameter to account for these differences when fitting the model to data of different taxa.

### 2.3.2 Nitrogen storage

Cellular nitrogen quotas are known to change considerably as a function of the external substrate concentration to which cells are acclimated (Droop, 1973; Caperton and Meyer, 1972). The difference in cell quota that occurs under different growth conditions is thought to increase as a function of cell size, when maximal nitrogen quotas scale faster than minimal nitrogen quotas (Verdy et al., 2009). However, changes in nitrogen quota are usually accounted for primarily by changes in cellular protein content in different growth conditions (Dortch et al., 1984). Thus, changes in the total nitrogen quota as a function of cell size cannot be used directly to constrain the size of our nitrogen reserve pool, which may contain only inorganic forms of N.

In different species, inorganic nitrogen may contribute anywhere between 0 (Dortch et al., 1984) and ~40% (Lourenço et al., 1998) of total cellular nitrogen. We do not know of any previously reported studies of the size dependence of stored, inorganic nitrogen. We therefore assumed a maximum capacity for nitrogen storage that is invariant of cell size, such that
\[
N_{R}^{\text{max}} = f_{	ext{stor}} N_{F},
\]
where \(f_{	ext{stor}}\) is the maximum capacity for storage as a fraction of the total functional nitrogen concentration (see Table 1). We acknowledge this treatment may overlook a reduced capacity to store nitrogen in some very small prokaryotes.

### 2.3.3 Carbon storage

To the best of our knowledge, there are insufficient measurements of carbon storage quotas to directly infer allometric relations. We therefore parameterized the maximum capacity for carbon storage in the following way. Carbon contained in the functional pool (that includes pigments, nucleic acid, etc.) is expected to reach a minimum when cells are nutrient starved (Dortch et al., 1984). According to Mei et al. (2011), the minimal carbon quota associated with the functional apparatus is (in mmol C cell\(^{-1}\)) (see also, Shuter, 1978)
\[
Q^{\text{C}}_{F, \text{min}} = 9.9 \times 10^{-12} \nu^{0.72}.
\]

We assume that whole-cell maximal carbon quotas (\(Q^{\text{C}}_{\text{max}}\)) are associated with cells grown under nutrient-replete conditions, and scale as a power law function of cell volume (Menden-Deuer and Lessard, 2000):
\[
Q^{\text{C}}_{\text{max}} = 18 \times 10^{-12} \nu^{0.9}.
\]

Under nutrient limitation, cells divert fixed carbon away from biosynthesis of functional components, and toward synthesis of reserve polymers (Rodolfi et al., 2009). Thus, we assume that differences in the functional carbon cell quota under nutrient limitation, and the maximum carbon quota under nutrient-replete conditions, may be used to approximate the maximum potential capacity for carbon storage:
\[
C^{\text{R}}_{\text{max}} = \left( \frac{Q^{\text{C}}_{\text{max}}}{Q^{\text{C}}_{F, \text{min}}} - 1 \right) C_{F}
\]

The exponent in Eq. (19) is larger than the exponent in Eq. (18), so the capacity for carbon storage is expected to increase as a function of cell volume.

### 2.4 Model parameterization

The parameters in Table 4 were tuned to enable model predictions of growth, Chl: C and C: N to agree with measurements of several species of phytoplankton cultured in photon flux density (PFD) and nitrogen-limiting conditions. In order to test the influence of storage capacity in a range of cell sizes, organisms selected include low-light-adapted Prochlorococcus marinus SS120 (Moore et al., 1995), high-light-adapted P. marinus (MED4) (Bertilsson et al., 2003), Synechococcus WH8012 and WH8103 (Moore et al., 1995), the freshwater strain Synechococcus linearis (Healey, 1985) and the diatom Skeletonema costatum (Sakshaug et al., 1989). All measurements are of organisms in balanced growth.

Most of the remaining model parameters were taken from allometry (Table 3). The carbon cost of nitrogen assimilation (\(\xi\)) and the quantum efficiency of photosynthesis (\(\phi_m\)), were assumed based on previously established theoretical considerations (see Table 1). The reduction in dark N assimilation was constrained with data from DiTullio and Laws (1986) (see Table 2).
Table 2. Diurnal changes in nitrogen assimilation based on \(^{14}\)C incorporated into proteins. Data are from DiTullio and Laws (1986). Data are given as a percentage of total N assimilation calculated with CHN analyses (DiTullio and Laws, 1986). Average reduction in dark N assimilation (i.e., \(\alpha_N\)) is 0.59.

<table>
<thead>
<tr>
<th>Species</th>
<th>Light (12 h)</th>
<th>Dark (12 h)</th>
<th>Ratio</th>
</tr>
</thead>
<tbody>
<tr>
<td><em>P. tricornutum</em> (diatom)</td>
<td>101</td>
<td>74</td>
<td>0.73</td>
</tr>
<tr>
<td><em>P. lutheri</em> (haptophyte)</td>
<td>98</td>
<td>79</td>
<td>0.81</td>
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<tr>
<td><em>Isochrysis sp.</em> (dinoflagellate)</td>
<td>154</td>
<td>95</td>
<td>0.62</td>
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<tr>
<td><em>A. carteri</em> (dinoflagellate)</td>
<td>213</td>
<td>40</td>
<td>0.19</td>
</tr>
<tr>
<td><em>D. salina</em> (halophilic chlorophyte)</td>
<td>119</td>
<td>70</td>
<td>0.59</td>
</tr>
</tbody>
</table>

2.5 Parameterizing resource variability

In the ocean, surface wind and temperature forcing cause vertical transport of phytoplankton due, for example, to deep convection (Backhaus et al., 2003) and turbulent mixing (Huisman et al., 1999). Due to the attenuation of light by water and dissolved and suspended material, cells that undergo such vertical motions experience variation in ambient photon flux density, sometimes over several orders of magnitude. Consequently, the effective “photophase” (i.e., the time period in which cells are in the light) may in some conditions be extremely short. For example, in the North Atlantic, transport due to deep convection may result in cells completing 800 m vertical loops on the order of a day (Backhaus et al., 2003). Assuming a constant average vertical velocity and an euphotic depth of approximately 100 m, cells in such a system would be in the dark for roughly 21 h.

We mimicked the effect of vertical transport on phytoplankton exposure to light by conducting simulations in which hypothetical, model organisms were exposed to intermittent photon flux doses within a 24 h period:

\[
E(t) = \begin{cases} 
200 & \text{if } 0 < t \leq \rho \\
0 & \text{otherwise} \end{cases}
\]

When \(\rho = 0\), the cells are exposed to complete darkness for the whole day. When \(\rho = 1\), the cells are exposed to saturating irradiance for 24 h. By varying \(\rho\) within the range [0, 1], we were able to mimic changes in average photon flux density and photosynthesis, due to changes in photophase. Note that the value 200 was chosen to completely saturate photosynthetic rates within the euphotic zone.

In addition to light, phytoplankton cells may experience variability in nutrient concentration by passing in and out of small-scale nutrient “patches” (Seymour et al., 2009). We accounted for the possibility that organisms may pass in and out of small-scale nutrient patches by applying similar, idealized step changes in the ambient substrate concentration:

\[
S(t) = \begin{cases} 
1 & \text{if } 0 < t \leq \rho \\
0 & \text{otherwise} \end{cases}
\]

In order to test the sensitivity of our results to different assumptions regarding PFD and nutrient variability, we also tested two additional scenarios (see supporting information). One additional set of experiments mimicked multiple visits to the euphotic zone or nutrient patch by allowing multiple intermittence phases within a single 24 h period. The other allowed for much slower transport by testing intermittence phases on the order of a week.

To test the combined effect of nutrient starvation and intermittent light on organism metabolic state, a repeat of all scenarios was performed in which model organisms were pre-acclimated to very low-nutrient conditions. The results of these experiments were contrasted against the main set of experiments, in which there was an initial spin-up time involving exposure to resource replete conditions. The resource sufficient spin-up was imposed by setting the ambient nutrient concentration far higher than the Michaelis–Menten half-saturation constant for nutrient uptake (\(S \gg K_S\)), and the ambient irradiance to be significantly greater than \(E_k = P_m/\alpha\), the saturation point of photosynthesis (\(E \gg E_k\)). In contrast, experiments that tested the combined effect of nutrient starvation and intermittent light on organism metabolic state imposed a very low ambient nutrient concentration (i.e., \(S \ll K_S\)), and a saturating PFD (\(E \gg E_k\)), during the model spin-up.

3 Results

3.1 Model-data comparisons

When cultured in nutrient-replete conditions, the growth rates of *Prochlorococcus marinus* SS120, *Synechococcus WH8103* and *Skeletonema costatum* (abbreviated SS120, WH8103 and *S. costatum*, respectively) all increase as a function of ambient PFD, eventually reaching a maximum at high PFD (Fig. 5). Furthermore, Chl : C declined with increasing growth irradiance in all three organisms. When the parameters in Table 4 were tuned to match experimental observations, the model is able to capture the observed dependence of growth rate and Chl : C on PFD for *Prochlorococcus SS120*, *Synechococcus WH8103* and *S. costatum* (Fig. 5).

Under nitrogen limitation, carbon fixed via photosynthesis is diverted away from protein synthesis, and toward synthesis of carbohydrates and lipids (Rodolfi et al., 2009). Thus, when grown under nitrogen limitation, phytoplankton cultures tend to show increases in cellular C : N at low growth rates (Fig. 6). The model is able to replicate the dependence of C : N ratio on nitrogen-limited growth rate for all species in Figs. 6 and 7.

The allometric relations for carbon storage quota suggest large phytoplankton cells are able to accumulate significantly more carbon reserves than small cells (Table 3). Thus, model predictions suggest large cells that accumulate relatively more storage lipid and carbohydrate should reach
higher nitrogen limited C:N ratios. Model predictions of the size dependence of C:N ratio are supported by data corresponding to *P. marinus* (MED4), *Synechococcus WH8103* and WH8103, *S. linearis* and *S. costatum* (Fig. 6).

### 3.2 Growth in a constant environment

Due to reduced package effects, and their high surface area to volume ratio, small cells are expected to have higher average growth rates than large cells when either PFD or nitrogen supply are limiting. In fact, when interspecific differences in the initial slope of the $P-E$ curve are assumed to arise solely from size-related pigment packaging, the model underpredicts observed growth rates of *S. costatum* (Fig. 5c), which suggests this diatom may only partially be influenced by pigment packaging. The advantage of small cell size is nonetheless evident at low nitrogen supply rates, even when the model is parameterized for *S. costatum* with a maximum growth rate approximately double that of *P. marinus* (SS120) (Figs. 5 and 9a). With sufficiently high PFD and nitrogen supply, *S. costatum* reaches its maximum growth rate, and any advantage of small cell size disappears (Fig. 9a).

### 3.3 Intermittence experiments

The model predicts that organisms with a sufficiently large capacity for storage are able to accumulate carbon reserves under saturating PFD, which may subsequently be used to fuel growth in the dark (Fig. 8). Accumulation and subsequent mobilization of carbon reserves leads to fluctuations in the C:N ratio (Fig. 8a). Even when forced with intermittent PFD, the model predicts relatively invariant C:N ratio of small cells with limited capacity for carbon storage (Fig. 8a). Due to this inability to accumulate reserve carbon, the model predicts that very small cells may be unable to maintain growth in the dark. Thus, model predictions suggest...
the ability to store carbon may confer an advantage to larger organisms under exposure to intermittent PFD (Fig. 8).

When forced with intermittent PFD, the model predicts *S. costatum* may on average grow more than twice as fast as *P. marinus* (SS120), even when the average daily PFD is extremely low (Fig. 9b). The benefit of small cell size nonetheless persists at a very low nitrogen supply rate, even when the model is forced with intermittent PFD (Fig. 9b).

The model predicts that *P. marinus* (SS120) should still grow faster than *S. costatum* at low average nitrogen supply rate, even when forced with intermittent nitrogen supply (Fig. 9c). Indeed, because the capacity for inorganic nitrogen storage is relatively low and invariant with cell size, there is almost no discernible influence of intermittent nitrogen pulses on the modeled balance between *S. costatum* and SS120 growth rates (Fig. 9c).

Phytoplankton carbon storage is expected to reach a maximum when organisms are nutrient starved (Rodolfi et al., 2009). Thus, one might expect the influence of variable PFD...
4 Discussion

We used a model to understand how energy stored in carbohydrates and lipids influences phytoplankton growth rate in environments with ephemeral PFD. The model was parameterized in part using allometric relationships for carbon storage quotas and nutrient uptake rates (Table 3), and in part by fitting to experimental data sets (Table 4 and Fig. 5). This empirical parameterization led to the model prediction that the very smallest phytoplankton cells should have a low capacity to store carbon, which is associated with relatively inflexible

to change depending on phytoplankton nutrient status. Indeed, the modeled benefit of carbon storage in environments with intermittent PFD is greater in experiments that involved prior acclimation to a low nitrogen supply rate, by comparison to experiments that involved prior acclimation to a high nitrogen supply rate (Fig. 9b and d).
C: N ratios (Fig. 8). Our model suggests that an inability to store carbon reduces the capacity for cells to synthesize functional biomass during darkness. In contrast, phytoplankton cells with the ability to accumulate large carbon stores (Griffiths and Harrison, 2009), may continue to synthesize functional biomass in the dark, albeit at a reduced rate (Table 2, Fig. 8).

Our results may have implications for understanding the distribution of very small phytoplankton cells in different oceanic regimes. For example, in environments with deep convection, cells are regularly mixed well below the euphotic depth (Backhaus et al., 1999). Such environments therefore involve prolonged exposure to darkness, and may favor relatively large cells with sufficient capacity for storage. It has been suggested previously that dominance of larger organisms in more variable environments may be linked to the capacity to store nutrients such as phosphorus and iron (Grover, 1991a, 1991b, 2011, Tozzi et al., 2004). The link between cell size, carbon quota, and infrequent PFD has received far less attention.

The prediction that the smallest prokaryotic autotrophs should have a diminished capacity for storage is unsurprising in light of the strong evolutionary pressure toward small cell size in low-nutrient environments, which may also have caused Prochlorococcus to shrink its genome (Partensky and Garczarek, 2010). Prochlorococcus are typically most dominant in relatively stable environments, with very low-nutrient supply rates (Partensky et al., 1999). Pressure to optimize nutrient uptake in stable, low-nutrient environments is likely to subordinate storage requirements, even when photon supply is intermittent (Fig. 9b and d). Larger organisms tend to dominate in environments with relatively high nutrient input, where small cells are intensely grazed and the need to optimize surface area to volume ratios disappears (Chisholm, 1992; Ward et al., 2013). Our model indicates one additional benefit to large cell size in eutrophic ecosystems.

At high latitude phytoplankton may be exposed to many months of darkness during winter (McMinn and Martin, 2013). Without going into resting stages, tolerance of prolonged exposure to darkness is influenced by the capacity for basal respiration, which is also likely to depend on reserve carbon availability (Furusato and Asaeda, 2009). Organisms able to survive prolonged exposure to darkness without going into a resting stage may respond faster when favorable conditions return. Thus, while this work has focused on the benefit of carbon storage to organism growth rates, there may also be ecologically significant benefits to survival associated with flexibility in C:N ratio, and accumulation of carbon reserves.

Not all experimental data used to constrain and interpret our model were of organisms in similar culture conditions. For example, while Sakshaug et al. (1989) cultured S. costatum over a range of day lengths, Moore et al. (1995) grew P. marinus on 10 light–dark cycles. Furthermore, the C:N data of Bertilsson et al. (2003) were of cyanobacteria in batch culture, exposed to P starvation, which may underestimate the nitrogen-starved C:N ratio (Goldman et al., 1979). In addition, none of the data were explicitly of carbohydrate or lipid abundance, and C:N variability was used to infer changes in macromolecular composition. Additional
experimental data to further advance the theory presented here include measurements of the accumulation and consumption of different storage carbohydrates and lipids, under conditions of intermittent photon supply for a range of species cultured under comparable experimental conditions.

We did not include a size dependence of the inorganic reserve N quota. By comparison to carbon, phytoplankton typically do not have large quotas for inorganic N; most of the nitrogen “stored” by large phytoplankton is usually proteinaceous (Geider and La Roche, 2002). High-protein quota may buffer protein degradation, prolonging survival at the individual level. Recycling of nitrogen and carbon contained in proteins may also lead to a more flexible metabolic strategy. Nonetheless, this recycling does not lead to a net gain in the cells’ nitrogen or carbon quota. In contrast, carbohydrates, lipids and inorganic forms of N have no direct metabolic function. Their subsequent assimilation into proteins must lead to an increase in the capacity for assimilating C, N or both.

Accumulation of carbon reserves under PFD fluctuations and nutrient limitation have been widely reported (Handa, 1969; Packer et al., 2011), but the ecological significance of this storage is not well understood. Accumulation of storage compounds is nonetheless responsible for large fluctuations in the C : N ratio. How do these size-dependent constraints on stoichiometry influence large-scale patterns in C : N? A compilation of existing data by Martiny et al. (2013), suggests nutrient-poor, high-light environments have relatively high ratios of particulate organic carbon (POC), to particulate organic nitrogen (PON) (i.e., high POC : PON). By contrast, darker, nutrient-rich waters have lower POC : PON. How can these observations be reconciled with the suggestion here that large cells more likely to dominate at high latitude can have the highest C : N? We hypothesize that, even if more temperate environments favor large cells with the propensity for high C : N, large cells only obtain such high values transiently, when the ability to fix carbon exceeds the rate at which nitrogen may be assimilated. It may therefore not come as a surprise that a compilation of data taken over a large spatio-temporal range indicates that, on average, light limited, nutrient rich environments have relatively low C : N. Cyanobacteria that dominate in the gyres may not have the capacity to accumulate such large carbon reserves, but may well maintain C : N ratios close to their maximum limit in direct response to the local environment.

Phytoplankton stoichiometry is also likely to influence food web dynamics (Loladze et al., 2000). Phytoplankton cells with high carbon relative to other main constituents (N,P), are often less palatable to herbivores (Urabe et al., 2002), although these effects may be offset when predators are able to graze upon multiple food types (Urabe and Waki, 2009). The manner in which prey stoichiometry influences herbivore growth is likely to influence rates of export production (Anderson et al., 2013). We suggest that the model presented here is a useful tool for further investigations of the influence of phytoplankton C : N on ecosystem function.

We have focused here on the benefit of a large carbon reserve to organism growth rates. We nonetheless do not exclude the possibility that “excess” C may be excreted from the cell, forming a protective polysaccharide layer (Wotton, 2004). Furthermore, using reserve carbon to fuel respiratory costs associated with the maintenance of buoyancy (Waite et al., 1997), may also be a valuable survival mechanism when cells are vulnerable to rapid sinking away from the euphotic zone. We anticipate that the model of intracellular C : N dynamics presented here may in the future be expanded to include multiple ecological benefits of a large carbon reservoir.

5 Conclusions

Larger phytoplankton cells able to accumulate a significant amount of reserve carbon polymers may be able to maintain active metabolism in the dark, thereby buffering the effects of prolonged light limitation. While the smallest autotrophs are optimized for nutrient acquisition in oligotrophic environments, they may be less equipped to cope with light limitation often found at high latitude (Fig. 1). We suggest this is one additional factor that influences the distribution of small and large organisms in different trophic regimes. Furthermore, due to accumulation of carbon storage compounds, large organisms may have a higher potential C : N ratio, and are likely to exhibit a wider range of values. We hope that in the future, the model presented may be combined with more detailed descriptions of PFD variability and interspecific interactions, to better understand the influence of carbon storage on large-scale patterns of the C : N ratio, and the distributions of different phytoplankton size classes.

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