Associate Editor Initial Decision: Publish subject to minor revisions (Editor review) (12 Jan 2015) by Prof. Katja Fennel

Comments to the Author:

Thank you for revising your manuscript further. There are two remaining issues which have to be addressed before your manuscript can be published.

Response: We are very grateful to your helpful advices. We also thank particularly Prof. Boss and the other two anonymous reviewers. From the revising of our manuscript, we understood SCMLs and related scientific problems more deeply than before.

In your additions to section 4.2, please remove the sentence:

"This assumption has been proved by Mellard et al. (2011)."

Often there is a significant amount of plankton in the surface layer even if chlorophyll is low (see e.g. Fennel and Boss 2003 for examples using BATS data).

Response: Agree. We delete this sentence in the revision.

I also don't agree with this sentence:

"If nutrient input directly to the mixed layer due to riverine inputs, surface runoff, or atmospheric deposition, Chl a concentration within the surface mixed layer will be sustained, while a SCM by itself will be not possible (Mellard et al. 2011)."

Subsurface chlorophyll maxima (SCM) do occur in coastal regions with riverine inputs, etc. I have seen numerous examples in observations. Just because your highly simplified model breaks down in these cases doesn't mean there are no SCM in reality. Please change this sentence accordingly.

Response: Thank you for this comment. This sentence is modified to

"If nutrient input directly to the mixed layer due to riverine inputs, surface runoff, or atmospheric deposition, a positive Chl a concentration within the surface mixed layer will be sustained (Mellard et al., 2011). In this case, the general Gaussian function can be modified with a superimposition of a background which is either a constant (Platt et al., 1988; Richardson et al., 2003) or a function of water depth (Uitz et al., 2006; Mignot et al., 2011)."

A List of what we changed in the revised version of manuscript:

Revised version 5	Revised version 6			
Line 480-481	Line 480: Delete the sentence "This assumption has been			
Lille 460-461	proved by Mellard et al. (2011)."			
Line 489-492	Line 488-493: Modify the sentence to "If nutrient input			
	directly to the mixed layer due to riverine inputs, surface			
	runoff, or atmospheric deposition, a positive Chl a			
	concentration within the surface mixed layer will be sustained			
	(Mellard et al., 2011). In this case, the general Gaussian			
	function can be modified with a superimposition of a			
	background which is either a constant (Platt et al., 1988;			
	Richardson et al., 2003) or a function of water depth (Uitz et			
	al., 2006; Mignot et al., 2011)."			
References	Add 2 references in line 739-741 and line 790-791,			
	respectively.			

Steady-state solutions for subsurface chlorophyll maximum in stratified water columns with a bell-shape vertical profile of chlorophyll X. Gong, J. Shi, H. W. Gao, X. H. Yao Key Laboratory of Marine Environment and Ecology (Ministry of Education of Marine Environment and Ecology)

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Abstract:

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A bell-shape vertical profile of chlorophyll a (Chl a) concentration, conventionally referred to as Subsurface Chlorophyll Maximum (SCM) phenomenon, has frequently been observed in stratified oceans and lakes. This profile is assumed to be a general Gaussian distribution in this study. By substituting the general Gaussian function into ecosystem dynamical equations, the steady-state solutions for SCM characteristics (i.e., SCM layer depth, thickness, and intensity) in various scenarios are derived. These solutions indicate that: 1) The maximum concentration of Chl a occurs at or below the depth of maximum growth rates of phytoplankton located at the transition from nutrient limitation to light limitation, and the depth of SCM layer deepens logarithmically with an increase in surface light intensity; 2) Thickness and intensity of the SCM are mainly affected by nutrient supply, but independent of surface light intensity; 3) The intensity of SCM layer is proportional to the diffusive flux of nutrients from below, getting stronger as a result of this layer being shrunk by a higher light attenuation coefficient or a larger sinking velocity of phytoplankton. In addition, the limitation and potential application of the analytical solutions are also presented.

1 Introduction

- Vertical profiles of chlorophyll a (Chl a) concentration in lakes, coastal seas and open 26 oceans are highly variable. However, a bell-shape vertical profile of Chl a, 27 conventionally referred to as Subsurface Chlorophyll Maximum (SCM) phenomenon, 28 has been frequently observed in stratified water columns, e.g., it occurs through the 29 whole year in tropical and subtropical oceans while it exists only during summer in 30 temperate and high latitude oceanic zones. The subsurface biomass maxima (SBMs) 31 are also common in stratified water columns. The chlorophyll-to-biomass ratio 32 generally increases with depth in the euphotic zone. Thus, SCMs may not necessarily 33 34 represent SBMs (Cullen, 1982; Fennel and Boss, 2003) and are usually deeper than SBMs (Fennel and Boss, 2003; Hodges and Rudnick, 2004). However, both the 35 subsurface maxima in chlorophyll and biomass are usually formed in certain regions 36 of the water column where two opposing resource (light and nutrient) gradients 37 38 combined with turbulent mixing is amenable for survival of phytoplankton. Thus, SCMs are approximately equal to SBMs in many studies (Klausmeier and Litchman, 39 40 2001; Sharples et al., 2001; Huisman et al., 2006; Raybov et al., 2010). Fennel and Boss (2003) reported that the photoacclimation of phytoplankton can be another 41 42 important reason for forming a SCM in oligotrophic waters. The SCM phenomenon can be characterized by the thickness, depth, and intensity of 43 SCM layer (SCML) (Beckmann and Hense, 2007). On-site observations (Platt et al., 44 1988; Sharples et al., 2001; Dekshenieks et al., 2001; Mellard et al., 2011) showed 45 that the SCML occurred relatively shallow (1-50 m) and was thin (several centimeters 46 to a few meters) in lakes and coastal seas, but the concentration of Chl a was high 47 (1-100 mg/m³). In open oceans, the SCML was deeper (80-130 m) and thicker (tens 48 of meters) while the concentration of Chl a was relatively low (<1 mg/m³) (Anderson, 49 50 1969; Platt et al., 1988). 51 SCMs have attracted much attention because of the significant contribution of SCML 52 to the total biomass and primary production in the whole water column (Cullen and
- to the total biomass and primary production in the whole water column (Cullen and Eppley, 1981; Weston et al., 2005; Siswanto et al., 2005; Hanson et al., 2007; Sullivan et al., 2010). Pérez et al. (2006) showed that 65-75% of the total Chl a in a water column of the Atlantic subtropical gyres was presented in SCML and the layer thickness was approximately 50 m. Weston et al. (2005) reported that the SCML

- accounted for 58% of the water column primary production in the central North Sea,
- although the layer thickness was less than 5 m. Sullivan et al. (2010) found that the
- fraction of Chl a in the SCML (thickness <3 m) out of the total water column ranged
- from 33% to 47% in the Monterey Bay.
- 61 Many numerical studies have been conducted to link the thickness, depth and
- 62 intensity of the SCML to various environmental parameters (Jamart et al., 1979;
- Varela et al., 1994; Klausmeier and Litchman, 2001; Hodges and Rudnick, 2004;
- Huisman et al., 2006; Beckmann and Hense, 2007). The thickness of the SCML
- 65 mainly depends on the degree of vertical mixing in lakes (Klausmeier and Litchman,
- 66 2001). In oligotrophic oceans, light attenuation coefficient is the key factor in
- determining the SCML depth (Varela et al., 1994; Hodges and Rudnick, 2004;
- 68 Beckmann and Hense, 2007) and the intensity of the SCML depends strongly on
- sinking velocity of phytoplankton and/or detritus and vertical diffusivity rather than
- 70 growth rate of phytoplankton (Hodges and Rudnick, 2004; Beckmann and Hense,
- 71 2007). However, the thickness, depth and intensity of SCML are very sensitive to
- variations of environmental parameters. Therefore, the relationships obtained from a
- particular case may not be applicable for other cases. To understand the general
- 74 relationships between SCM phenomenon and environmental parameters, the
- analytical solution for dynamic ecosystem equations is needed.
- The algae game theoretical model, pioneered by Klausmeier and Litchmann (2001),
- was perhaps the first one to derive the depth and intensity of SCML, although the
- 78 SCML is assumed to be infinitely thin. They adopted a delta function to approximate
- 79 the phytoplankton distribution in this thin layer. Yoshiyama et al. (2009) used this
- model to examine more than one species competing for limiting nutrients and light
- below the surface mixed layer. Mellard et al. (2011) included stratification into this
- model. However, the SCML was still confined to an infinitely thin layer. In fact,
- many observations showed that the thickness of SCML can reach as high as 100 m in
- oceans (Platt et al., 1988). For those cases, the assumption of an infinitely small
- 85 thickness of SCML is contradictory to the observations.
- In this study, we assume that the vertical profile of Chl a can be approximately treated
- as a general Gaussian function, instead of a delta function. This parameterizing
- approach was proposed firstly by Lewis et al. (1983), and has been widely used to fit
- vertical profiles of Chl a (Platt et al., 1988; Weston et al., 2005; Ardyna et al., 2013).

By incorporating the general Gaussian function into the ecosystem dynamical equations, we derive the steady-state solutions for the thickness, depth, and intensity of SCML in various scenarios and examine their dependence on environmental parameters, such as light attenuation coefficient, vertical diffusivity, sinking velocity of phytoplankton.

2 Methods

2.1 Models

The SCML occurs below the surface mixed layer, where the light attenuated from above and nutrients supplied from the deep water result in the maximal value of phytoplankton growth rate (Fig. 1). The partial differential equations for phytoplankton and nutrients dynamics in which light and nutrients are two major limiting factors (Eqs. 1 and 2) (Riley et al., 1949; Lewis et al., 1986; Gabric and Parslow, 1989; Huisman et al., 2006; Liccardo et al., 2013) were adopted in this study.

$$\frac{\partial P}{\partial t} = \mu_{m} \min(f(I), g(N)) P - \varepsilon P - w \frac{\partial P}{\partial z} + \frac{\partial}{\partial z} \left(K_{v} \frac{\partial P}{\partial z}\right), \tag{1}$$

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$$\frac{\partial N}{\partial t} = -1.59 \mu_m \min(f(I), g(N)) P + 1.59 \alpha \varepsilon P + \frac{\partial}{\partial z} \left(K_v \frac{\partial N}{\partial z}\right), \tag{2}$$

where P denotes the Chl a concentration, N is the limiting nutrient concentration. The photo-acclimation of phytoplankton is not considered here and the Chl a distribution is supposed to represent the distribution of phytoplankton biomass. This is a significant simplification. In fact, phytoplankton increases inter-cellular pigment concentration when light level decreases (Cullen, 1982; Fennel and Boss, 2003). Usually, the unit of Chl a concentration is mg m⁻³, the concentrations of phytoplankton and the limiting nutrients are in unit of mmol N m⁻³. A ratio of 1.59 g chlorophyll per mol nitrogen (Cloern et al., 1995; Oschlies, 2001) is thereby used for unit conversion. μ_m is the maximum growth rate of phytoplankton, ε is the loss rate of phytoplankton (including respiration, mortality, zooplankton grazing), α is the recycling rate of dead phytoplankton ($0 \le \alpha \le 1$). w is the sinking velocity of phytoplankton, which is non-negative in the chosen coordinate system and assumed to be constant with depths. K_{ν} is the vertical turbulent diffusivity and it is much

- larger within the surface mixed layer than that beneath. Here, K_{ν} depends on depth
- in the following way (Hodges and Rudnick, 2004; Mellard et al., 2011):

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$$K_{v} = \begin{cases} K_{v1} & 0 < z < z_{s}, \\ K_{v2} & z_{s} < z < z_{b}, \end{cases}$$
 (3)

- where z_s is the depth of surface mixed layer, z_b is the location where the Chl a
- concentration reduces to nearly zero in a sufficiently deep water column. We assume
- 124 K_{v1} , K_{v2} are constant and K_{v1} is large enough to homogenize the Chl a and nutrient
- concentrations in the surface mixed layer.
- A gradual transition from the surface mixed layer to the deep one written in terms of a
- generalized Fermi function is adopted (Ryabov et al., 2010), that is, $K_v(z) = K_{v2} +$
- 128 $\frac{K_{v1}-K_{v2}}{1+e^{(z-z_s)/l}}$, where parameter *l* characterizes the width of the transition layer. In our
- study, we assumed this transition layer is finitely thin.
- The growth limited function $\min(f(I), g(N))$ for light I and nutrients N is:

$$\min(f(I),g(N)) = \min\left(\frac{I(z)}{K_I + I(z)}, \frac{N(z)}{K_N + N(z)}\right),\tag{4}$$

- where K_I and K_N denote the half-saturation constants of light and nutrients,
- respectively. The net growth rate, $\mu_m \min(f(I), g(N)) \varepsilon$, is positive only if both the
- light limiting term $\mu_m f(I)$ and nutrient limiting term $\mu_m g(N)$ are larger than the
- 135 loss rate ε .
- Light intensity is assumed to decrease exponentially with depth according to
- 137 Lambert-Beer's law, i.e.,

$$I(z) = I_0 \exp(-K_d z), \tag{5}$$

- where I_0 is the surface light intensity and K_d is the light attenuation coefficient (Morel,
- 140 1988). Assuming a constant K_d , we ignore the effects of the self-shading and the
- dissolved and particulate material on the attenuation coefficient.
- The zero-flux boundary condition for the phytoplankton at the surface is used. Like
- the study reported by Ryabov et al. (2010), we also set the chlorophyll concentration
- approaches zero at the bottom boundary z_b , i.e., $P \rightarrow 0$ for $z \rightarrow z_b$. Fennel and Boss
- 145 (2003) used an infinite depth as z_b . Furthermore, we assume a zero-flux boundary

146 condition for nutrients at the surface, while nutrients are replenished from below.

147 That is,

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$$\begin{cases} K_{v1} \frac{\partial P}{\partial z} = 0, & K_{v1} \frac{\partial N}{\partial z} = 0, & at \ z = 0, \\ P(z_b) = 0, & K_{v2} \frac{\partial N}{\partial z} = K_{v2} \frac{\partial N}{\partial z} \Big|_{z=z_b}, & at \ z = z_b. \end{cases}$$
(6)

In addition, Lewis et al. (1983) first proposed a general Gaussian distribution function

- (Eq. 7) to model the nonlinear feature of observed vertical Chl a profiles. In this study,
- this function is adopted to represent the bell-shape vertical distribution of Chl a (Fig.
- 152 1).

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$$P(z) = P_{\text{max}} e^{\frac{-(z - z_m)^2}{2\sigma^2}} \quad 0 \le z \le z_b,$$
 (7)

- where P(z) is Chl a concentration as a function of depth z, and $P_{\text{max}} = \frac{h}{\sigma\sqrt{2\pi}}$. The
- three Gaussian parameters (h, z_m, σ) can vary to characterize the SCM phenomenon.
- Thus h is the vertical integrated Chl a over the entire water column, z_m is the depth of
- the maximum Chl a (the peak of the bell-shape), and σ is the standard deviation of
- Gaussian function, which controls the thickness of the SCML.
- 159 *2.2 Three SCM characteristics*
- The thickness of SCML can characterize the vertical extent of Chl a distribution
- below the surface mixed layer. It is still debatable how to best define the thickness of
- SCML. One easy definition is to use the width between two locations below and
- above the Chl a peak, where Chl a is a certain fraction (e.g. 50%, 100(e^{-1/2})%) of the
- maximum Chl a (Platt et al., 1988; Pérez et al., 2006). Some studies bounded the
- layer by sharp vertical gradients in Chl a above and below the peak (Prairie et al.,
- 166 2011). Others defined the upper and lower boundary of SCML by ad hoc choices.
- Pedrós-Alió et al. (1999) proposed the SCML from the depth of the surface mixed
- layer to the lower maximum gradient in the slope of the Chl a profile. Hanson et al.
- 169 (2007) defined that the upper boundary of the SCML was the minimum gradient
- criterion of 0.02 mg Chl a m⁻¹ and the lower was the base of the euphotic zone.
- Beckmann and Hense (2007) proposed to define the boundaries of SCML by the
- existence of two community compensation depths in the water column, which were
- located at the depths of two maximum phytoplankton gradients in phytoplankton

- 174 biomass.
- Building on the study by Beckmann and Hense (2007), the locations of the maximum
- phytoplankton gradients are defined as the boundaries of SCML in this study. That is,

$$\frac{d^2P}{dz^2}\Big|_{z=z_u,z_l} = 0,$$
(8)

- where z_u and z_l are the upper and lower boundary of SCML, respectively.
- By substituting Eq. (7) into this equality, we obtain $z_u = z_m \sigma$, $z_l = z_m + \sigma$. Thus,
- the thickness of SCML can thereby be expressed as 2σ .
- From Eq. (8) and the steady state of Eq. (1), one gets the following equality at the
- boundaries of SCML:

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$$\left(\mu_m \min(f(I), g(N)) P - \varepsilon P - w \frac{dP}{dz} \right) \Big|_{z=z_u, z_l} = 0.$$
 (9)

- That is, the boundary of SCML is located at the depth where there is the balance
- between phytoplankton growth and all losses (including the divergence of the sinking
- flux $w \frac{dP}{dz}$ and the loss ε due to mortality, respiration, and grazing), named the
- community compensation depth (Ono et al., 2001). Thus, this definition reflects the
- physical-biological ecosystem dynamics associated with SCML.
- As described in Eq. (7), the depth of the SCML is defined as z_m , that is, the location
- of the point-wise maximum value of Chl a.
- The third quantity, i.e. the intensity of SCML, refers to the maximum value of Chl a
- 192 $(P_{\text{max}} \text{ in Eq. 7})$ in the water column.
- 193 *2.3 Approach used in this study*
- Previous numerical studies (Huisman et al., 2006; Ryabov et al., 2010) showed that
- the ecosystem dynamical model (Eqs. 1 and 2) can approximately reproduce the
- bell-shape feature of the vertical Chl a profile (Fig. 1). We assume a general
- Gaussian function of P(z) (Eq. 7) is the solution for the Eqs. (1) and (2) at
- steady-state to derive explicit relationships between three characteristics of SCM and
- the environmental parameters. If nutrient input to the mixed layer due to riverine
- 200 inputs, surface runoff, or atmospheric deposition, is considered in the ecosystem, the
- surface concentration of Chl a should be positive (Mellard et al. 2011). Thus, the

- 202 general Gaussian function is not an exact solution, at best, an approximate solution of
- 203 the dynamical Eqs. (1) and (2) by ignoring external nutrient input.
- Firstly, by substituting the general Gaussian function of P(z) with the steady-state
- version of Eq. (1), we obtain that below the surface mixed layer the net growth rate of
- 206 phytoplankton can be expressed as follows

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$$\mu_{m} \min(f(I), g(N)) - \varepsilon = -\frac{K_{v2}}{\sigma^{4}} \left(z - z_{m} + \frac{w\sigma^{2}}{2K_{v2}}\right)^{2} + \frac{w^{2}}{4K_{v2}} + \frac{K_{v2}}{\sigma^{2}}.$$
 (10)

- Letting $\mu_m \min(f(I), g(N)) \varepsilon = 0$, we get the two compensation depths, z_{c1} , z_{c2} ,
- 209 by solving Eq. (10):

$$z_{c1} = z_m - \frac{w\sigma^2}{2K_{v2}} - \sqrt{\left(\frac{w\sigma^2}{2K_{v2}}\right)^2 + \sigma^2}, \ z_{c2} = z_m - \frac{w\sigma^2}{2K_{v2}} + \sqrt{\left(\frac{w\sigma^2}{2K_{v2}}\right)^2 + \sigma^2}. \tag{11}$$

- 211 From the property of quadratic function with pointing downward (the right-hand
- 212 terms in Eq. 10), we know that for $z_{c1} < z < z_{c2}$ the inequality
- 213 $\mu_m \min(f(I), g(N)) \varepsilon > 0$ is satisfied. This indicates that the subsurface net
- 214 production occurs only between the two compensation depths where the growth rate
- 215 $\mu_m \min(f(I), g(N))$ equals the loss rate ε . Beckmann and Hense (2007) found similar
- 216 results by numerical modeling and emphasized the often overlooked fact that an
- 217 SCML has to have two compensation depths.
- From Eq. (11), we obtain $z_{c1} \le z_m \sigma$ and $z_m \le z_{c2} \le z_m + \sigma$ (Fig. 1). In particularly,
- 219 $z_{c1} = z_m \sigma$, and $z_{c2} = z_m + \sigma$ when the sinking velocity of phytoplankton w is too
- small to affect the chlorophyll profile significantly. This result is identical to that of
- Beckmann and Hense (2007) for neglecting sinking velocity of phytoplankton.
- Hence, according to the property of quadratic function, there exists a depth z_0
- between the two compensation depths,

$$z_0 = z_m - \frac{w\sigma^2}{2K_{v2}},\tag{12}$$

such that the net growth rate of phytoplankton is at its maximum, i.e.,

$$\max\left(\mu_{m}\min(f(I),g(N))-\varepsilon\right)\Big|_{z_{0}} = \frac{K_{v2}}{\sigma^{2}} + \frac{w^{2}}{4K}.$$
 (13)

- In other words, the maximum in net growth rates of phytoplankton occurs at the
- depth of z_0 .
- We define $T = \sigma^2/K_{\nu 2}$ as the characteristic vertical mixing time scale in the SCML of
- thickness σ (Bowdon, 1985; Gabric and Parslow, 1989). Let the length scale be
- 231 $L=2K_{v2}/w$, which determines the scale height of the phytoplankton distribution
- 232 (Ghosal and Mandre, 2003). Thus, the right hand terms of Eq. (13) can be rewritten
- as 1/T+w/(2L). In other words, the maximum net growth rate of phytoplankton,
- $\max(\mu_m \min(f(I), g(N)) \varepsilon)$, is determined by the vertical mixing time scale (T) and the
- time taken by a phytoplankton sinking (w) through lengths (2L).
- Equation (12) also shows that $z_m \ge z_0$, that is, the depth of SCML lies at or below
- the depth for phytoplankton having the maximum growth rate. Observations in the
- Southern California Bight have supported this (Cullen and Eppley, 1981).
- Particularly, $z_m = z_0$ approximately holds when either the sinking velocity (w) or
- Gaussian parameter σ is very small. For non-sinking phytoplankton, i.e., $w \rightarrow 0$,
- numerical modeling can support this equality (Beckmann and Hense, 2007). When
- parameter σ is assumed to be infinitely thin, the equality is obviously correct, which
- has been used to solve for the equilibrium depth and intensity of an infinitely thin
- layer (Klausmeier and Litchman, 2001; Yoshiyama et al., 2009; Mellard et al., 2011).
- In this special case $(z_m = z_0)$, some studies found that the depth of SCML is at the
- location of equal limitation by nutrients and light (Klausmeier and Litchman, 2001;
- Yoshiyama et al., 2009; Mellard et al., 2011). In this study, we further infer that when
- 248 $z_m > z_0$, the depth of SCML is located at where phytoplankton growth is limited by
- 249 light (Appendix A).
- According to Eqs. (12) and (A2), the growth of phytoplankton is light-limited at and
- below the depth of SCML. Therefore, for $z = z_m$ and $z = z_m + \sigma$, the net growth rate
- of phytoplankton (Eq. 10) can be expressed as following, respectively:

$$\mu_{m} f(I)|_{z=z_{m}} -\varepsilon = K_{v2} / \sigma^{2}$$
(14)

$$\mu_{m} f(I)|_{z=z_{m}+\sigma} -\varepsilon = -w/\sigma \tag{15}$$

255 At the depth of z_m , the net growth rate of phytoplankton (Eq. 14) is determined by

- 256 the vertical mixing time, T, while the time taken by phytoplankton sinking through
- half-length of SCML, w/σ , controls the net growth rate of phytoplankton (Eq. 15) at
- the lower boundary of SCML $(z_m + \sigma)$.
- In addition, from Eqs. (12) and (A2) we obtain that the upper compensation depth, z_{c1} ,
- is the location where the growth limited by nutrients, $\mu_m g(N)$, equals the loss rate, \mathcal{E} ,
- while the lower compensation depth, z_{c2} , represents the depth where the growth
- limited by light, $\mu_m f(I)$, equals the loss rate, ε .

263 3 Results

- 3.1 Analytic solutions of three SCM characteristics
- By substituting the growth limitation function for light (Eqs. 4 and 5) into Eqs. (14)
- or (15), we obtain the expression of parameter z_m , i.e.,

$$z_{m} = \frac{1}{K_{d}} \ln \left[\left(\frac{\mu_{m}}{\varepsilon + K_{v2}/\sigma^{2}} - 1 \right) \frac{I_{0}}{K_{I}} \right]$$
 (16)

268 or

$$z_{m} = \frac{1}{K_{d}} \ln \left[\left(\frac{\mu_{m}}{\varepsilon - w/\sigma} - 1 \right) \frac{I_{0}}{K_{I}} \right] - \sigma.$$
 (17)

- The occurrence for a SCM requires $z_m > 0$. Requiring a positive solution for Eq.
- 271 (16), we obtain $\left(\frac{\mu_m}{\varepsilon + K_{v2}/\sigma^2} 1\right) \frac{I_0}{K_I} > 1$, i.e., $\left(\mu_m f(I_0) \varepsilon\right) \sigma^2 > K_{v2}$. For any $\sigma > 0$, we
- get $\mu_m f(I_0) > \varepsilon$. That is, the necessary condition for the existence of SCM is
- $\mu_{m}f\left(I_{0}\right)>\varepsilon$, which is identical with the result of Fennel and Boss (2003) when
- vertical sinking is constant as a function of depth in their model.
- Subtracting Eqs. (16) and (17), and rearranging, we obtain the expression of
- 276 parameter σ :

$$\left(\frac{\mu_m}{\mu_m - \varepsilon + \frac{w}{\sigma}} - 1\right) e^{K_d \sigma} = \frac{\mu_m}{\mu_m - \varepsilon - \frac{K_{v2}}{\sigma^2}} - 1$$
(18)

- Thus far, we have obtained the theoretical relationships between Gaussian parameter
- σ , z_m and environmental parameters (Eqs. 16-18). To derive the relationship between

Gaussian parameter h and environmental parameters, we now return to Eqs. (1) and (2). In steady state, adding these two equations leads to:

$$(1-\alpha)\varepsilon P + w\frac{dP}{dz} = \frac{d^{2}(K_{v}P)}{dz^{2}} + \frac{1}{1.59}\frac{d^{2}(K_{v}N)}{dz^{2}}$$
(19)

- Note that this relationship holds irrespective of the form of growth limiting function.
- Integrating this equation from the surface to bottom boundary (z_b) and using
- boundary conditions (Eq. 6) gives:

$$1.59(1-\alpha)\varepsilon \int_0^{z_b} P(z)dz = K_{v2} \frac{dN}{dz} \Big|_{z=z_b}$$
 (20)

- 287 When the recycling processes do not immediately convert dead phytoplankton back
- into dissolved nutrients below the surface mixed layer, i.e., $\alpha \neq 1$ (For $\alpha = 1$, the
- detailed derivation for the intensity of SCML is presented at Appendix B), one gets
- 290 the total Chl a in the water column:

$$h = \frac{K_{v2} \frac{dN}{dz} \Big|_{z=z_b}}{1.59(1-\alpha)\varepsilon}$$
 (21)

- This equality indicates that the total Chl a in the water column (h) is independent of
- the sinking velocity of phytoplankton. Both Ryabov et al. (2010) and Mellard et al.
- 294 (2011) obtained a similar result.
- 295 The intensity of SCML is

$$P_{\text{max}} = \frac{K_{v2} \frac{dN}{dz} \big|_{z=z_b}}{1.59\sqrt{2\pi}\sigma(1-\alpha)\varepsilon}$$
 (22)

- Obviously, both the total Chl a in the water column and the intensity of SCML are
- proportional to the flux of nutrients from below $(K_{v2} \frac{dN}{dz}|_{z=z_b})$, which is determined
- by the diffusivity below the surface mixed layer and the nutrients gradient at the
- 300 bottom of water column. Varela et al. (1994) also found a similar result by
- 301 simulations.
- 3.2 *Influences of environmental parameters on SCM characteristics*
- We now investigate how the steady-state thickness, depth, and intensity of SCML
- depend on environmental parameters. Because the analytic solutions for SCML depth

- and intensity depend on Gaussian parameter σ and environmental parameters, we first
- examine the influence of environmental parameters on parameter σ .
- Equation (18) shows that the thickness of SCML is independent of sea surface light
- intensity (I_0) . This is consistent with numerical simulations (Beckmann and Hense,
- 309 2007). This result also suggests that seasonal variation of SCML thickness has no
- 310 relation with light intensity. Thus, it is not surprising that the empirical model poorly
- predicted parameter σ by using season as an important factor (Richardson et al.,
- 312 2003).
- To illustrate the effects of other model parameters $(K_d, K_{v2}, \mu_m, \varepsilon, w)$ on the parameter
- 314 σ , we need to obtain informative algebraic expression of σ . To simplify, by Taylor
- expanding $e^{K_d \sigma}$ at $\sigma = 0$ and truncating the Taylor series after the linear term, i.e.,
- 316 $e^{K_d \sigma} = 1 + K_d \sigma + o(\sigma^2)$, Eq. (18) can thereby be rewritten as:

317
$$\sigma^{3} - \frac{w}{\varepsilon} \sigma^{2} - \frac{\varepsilon K_{d} K_{v2} + \mu_{m} w}{\varepsilon K_{d} (\mu_{m} - \varepsilon)} \sigma = \frac{K_{v2} (\mu_{m} / K_{d} - w)}{\varepsilon (\mu_{m} - \varepsilon)}.$$
(23)

- According to the properties of a cubic function, we know that Eq. (23) has one and
- only one positive real root σ , when $\frac{K_{v2}(\mu_m/K_d-w)}{\varepsilon(\mu_m-\varepsilon)} \ge 0$. Because $\mu_m f(I_0) > \varepsilon$ and
- 320 $0 \le f(I_0) \le 1$, so $\mu_m > \varepsilon$. Thus, when the maximum phytoplankton growth rate (μ_m)
- within one penetration depth $(1/K_d)$ is larger than sinking velocity of phytoplankton,
- i.e., $\mu_m/K_d w \ge 0$, there exists a non-negative value of parameter σ , which
- increases with increasing $\frac{K_{v2}(\mu_m/K_d-w)}{\varepsilon(\mu_m-\varepsilon)}$.
- Using dimensional analysis, Klausmeier and Litchman (2001) found that the degree
- of turbulence determines the thickness of SCML. Our analytical result shows that the
- 326 thickness of SCML increases with increasing vertical diffusivity below the surface
- mixed layer (K_{v2}) . In addition, the SCML thickness decreases with increasing sinking
- velocity of phytoplankton (w) and light attenuation coefficient (K_d).
- The right hand term in Eq. (23), $\frac{K_{v2}(\mu_m/K_d-w)}{\varepsilon(\mu_m-\varepsilon)}$, can be rearranged as
- 330 $\frac{K_{v2}(\mu_m/K_d-w)}{-(\varepsilon-\mu_m/2)^2+\mu_m^2/4}$. Thus, the effect of loss rate (ε) on parameter σ depends on $\mu_m/2$.

Note that $\mu_{\rm m} f(I_0) > \varepsilon$ once the SCM occurs. When the surface light intensity I_0 is smaller than or equals to the half-saturation constant for light K_I , i.e., $f(I_0) \le 0.5$, then $0 < \varepsilon < \mu_{\rm m} f(I_0) \le \mu_{\rm m}/2$, thus, σ decreases with increasing ε . Conversely, when $f(I_0) > 0.5$, for $\varepsilon \ge \mu_{\rm m}/2$, σ increases with increasing ε ; for $\varepsilon < \mu_{\rm m}/2$, σ decreases with increasing ε . In summary, for smaller loss rates ($\varepsilon < \mu_{\rm m}/2$), decreased ε leads to a thicker SCML, while for larger loss rates ($\varepsilon \ge \mu_{\rm m}/2$), decreased ε leads to a thinner SCML.

Equation (16) can be rewritten as:

$$z_{m} = \frac{1}{K_{d}} \ln \left(AI_{0} \right), \tag{24}$$

where $A = \frac{1}{K_{-}} \left(\frac{\mu_m}{\varepsilon + K_{-} \cdot / \sigma^2} - 1 \right)$. Clearly, from Eq. (18) we know A does not depend on 340 surface light intensity (I_0) , thus we infer that the depth of SCML increases 341 logarithmically with increasing I_0 . In other words, the SCML gets deeper due to the 342 seasonal increase of I_0 , and remains almost unchanged when the surface light 343 344 intensity increases to a certain degree. Observations at the HOT (Hawaii Ocean Time-series) site in the eastern Pacific and the SEATS (South East Asia Time-series 345 Station) station in the South China Sea showed a significant seasonal variation of 346 347 SCML depth (Chen et al., 2006; Hense and Beckmann, 2008). Hense and Beckmann (2008) explained the deepening of SCML depth in spring at HOT site by the seasonal 348 increase of the light intensity. Modeling sensitivity analyses also showed that an 349 increase in the surface light intensity yields a deeper SCML (Jamart et al., 1979; 350 Varela et al., 1994; Beckmann and Hense, 2007). 351 352 Determining the effect of vertical diffusivity below the surface mixed layer $(K_{\nu 2})$ on the steady-state SCML intensity is more difficult. Increased $K_{\nu 2}$ increases parameter 353 σ (Eq. 23) and the diffusive flux of nutrients from below (Eq. 22), however, this 354 parameter has opposite effects on P_{max} (Eq. 22). Rearranged Eq. (23) we obtain 355

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$$\frac{K_{v2}}{\sigma} = \frac{(\mu_m - \varepsilon)\varepsilon}{(\mu_m/K_d - w)/\sigma^2 + \varepsilon/\sigma} + \frac{(\mu_m - \varepsilon)w}{(\mu_m/K_d - w)/\sigma + \varepsilon} - \frac{\mu_m w/K_d}{\mu_m/K_d - w + \varepsilon\sigma}.$$
 (25)

Clearly, all the three terms in the right hand of this equality increase due to the increasing σ by a higher $K_{\nu 2}$. Therefore, it can be inferred that increased vertical

- diffusivity below the surface mixed layer $(K_{\nu 2})$ leads to a stronger SCML intensity
- 360 (P_{max}) .
- The influences of various parameters on SCM characteristics determined by Eqs.
- 362 (16)-(18), (21) and (22) are summarized in Table 1. For example, increased light
- levels (increasing surface light intensity I_0 , decreasing attenuation coefficient K_d) or
- increased light competitive ability (decreasing half-saturation constant for light K_l)
- moves the SCML deeper; increased nutrients supply (increasing vertical diffusivity
- below the surface mixed layer $K_{\nu 2}$ and loss rate of phytoplankton ε) moves the layer
- 367 toward the surface. The shape of SCML (thickness and intensity) is mainly
- influenced by nutrients supply ($K_{\nu 2}$ and ε). The intensity of SCML becomes weaker
- as a result of expanding the SCML by a lower sinking velocity of phytoplankton (w)
- and a smaller light attenuation coefficient (K_d).

4 Discussion

- Considering the two compartment system (nutrients and Chl a) in steady state and a
- 373 general Gaussian function for vertical Chl a concentration, we derived the analytical
- 374 solution for the fundamental relationships between SCM characteristics and various
- parameters. Three special scenarios, limitation and implications of this study were
- 376 discussed below.
- 377 4.1 Three special scenarios
- Equation (18) indicates that the parameter σ is affected by changes in the vertical
- diffusivity below the surface mixed layer $(K_{\nu 2})$, the sinking velocity of phytoplankton
- 380 (w) and the light attenuation coefficient (K_d) , which inversely affects depth and
- intensity of SCML (Eqs. 16, 17, and 22). Thus, three special situations of the
- theoretical solutions for SCM characteristics are discussed below.
- Firstly, the term K_{v2}/σ^2 in the right hand of Eq. (18) is neglected. This special
- situation occurs either when the vertical diffusivity below the surface mixed layer is
- too small to be considered $(K_{\nu 2} \rightarrow 0)$, or when $K_{\nu 2}/\sigma^2$ is much smaller than $\mu_m \varepsilon$,
- i.e., the mixing time scale $(T = \sigma^2/K_{v2})$ below the surface mixed layer is much longer
- than the time taken by net growth of phytoplankton, $(\mu_m \varepsilon)^{-1}$. Indeed, in the
- seasonal thermocline, vertical turbulent diffusive time scales can vary from weeks to
- months for phytoplankton displacements as small as several meters (Denman and

Gargett, 1983). The value of $(\mu_m - \varepsilon)^{-1}$ used in many studies is usually from 0.1 to 5

days (Gabric and Parslow, 1989; Klausmeier and Litchman, 2001; Huisman et al.,

392 2006).

In this situation, from Eq. (14), the growth rate at SCML depth can be expressed as:

$$\mu_m f(I)|_{z=z_m} = \varepsilon. \tag{26}$$

In regions with a low vertical diffusivity, Fennel and Boss (2003) derived that, at the

396 SCML depth, the growth rate of phytoplankton is equal to the loss rate and the

divergence of phytoplankton due to changes in the sinking velocity. Clearly, Eq. (26)

is identical to that of Fennel and Boss (2003) for constant sinking velocity of

399 phytoplankton.

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In this situation, the depth of SCML can be derived from Eq. (16), i.e.,

$$z_{m} = \frac{1}{K_{d}} \ln \frac{\left(\mu_{m} - \varepsilon\right) I_{0}}{\varepsilon K_{I}}.$$
 (27)

402 It indicates the SCML depth is directly proportional to the light penetration depth

403 $(1/K_d)$. Observations by four Bio-Argo floats in the North Pacific Subtropical Gyre,

404 the South Pacific Subtropical Gyre, the Levantine Sea, and in the northwestern

Mediterranean Sea showed a significant positive linear relationship between the two

variables (Mignot et al. 2014). Beckmann and Hense (2007) also found a similar

result by statistical analysis of numerical modeling.

The right hand term of Eq. (27) can be rewritten as $\frac{1}{K_d} \ln \frac{I_0}{I^*}$ by letting $I^* = \frac{\varepsilon K_I}{\mu_m - \varepsilon}$,

where $\mu_m f(I^*) = \varepsilon$. Under the assumption of infinitely thin SCML $(\sigma \rightarrow 0)$,

Klausmeier and Litchman (2001) also have derived Eq. (27) by setting the vertical

diffusivity for phytoplankton as zero, i.e., $K_y = 0$, in poorly mixed waters. Here, we

go further to obtain the approximate expression of the thickness of SCML from Eq.

413 (23), that is,

$$2\sigma = \frac{w}{\varepsilon} + \sqrt{\left(\frac{w}{\varepsilon}\right)^2 + \frac{w}{K_d \left(\varepsilon - \varepsilon^2 / \mu_{\rm m}\right)}}.$$
 (28)

Obviously, the thickness of SCML increases with an increase in the sinking velocity of phytoplankton (w), and with a decrease in the maximal growth rate (μ_m) and the light attenuation coefficient (K_d) .

The second special situation occurs when the term w/σ in the left hand of Eq. (18) is neglected. This special case occurs in regions where phytoplankton sinking velocity is very low $(w\rightarrow 0)$, or when w/σ is much smaller than $\mu_m - \varepsilon$, i.e., the time taken by phytoplankton sinking through half-length of SCML, $(w/\sigma)^{-1}$, is much longer than the time taken by net growth of phytoplankton, $(\mu_m - \varepsilon)^{-1}$. Phytoplankton sinking velocities exhibit a range of values depending on physical and physiological phenomena (e.g., size and shape of the cell). In the environment, estimates of sinking velocity vary from 0 to 9 m per day (Gabric and Parslow, 1989; Huisman and Sommeijer, 2002). Thus, the latter special scenarios (i.e., $w/\sigma \ll \mu_m - \varepsilon$) can indeed occur.

In this situation, according to Eq. (15), the net growth rate at the lower boundary of SCML can be expressed as

$$\mu_m f(I)|_{z=z_m+\sigma} -\varepsilon = 0. \tag{29}$$

- That is, the lower boundary of SCML, $z_m + \sigma$, is located at the compensation depth.
- In this situation, the depth of SCML can be derived from Eq. (17), i.e.,

$$z_{m} = \frac{1}{K_{d}} \ln \frac{(\mu_{m} - \varepsilon) I_{0}}{\varepsilon K_{I}} - \sigma.$$
 (30)

Compared with Eq. (27), we know that the depth of SCML is shallower in this special case than that in the case of neglecting the influence of vertical diffusivity below the surface mixed layer on SCM. This result implies that the displacement (σ) of SCML depth is the result of combined influences of vertical diffusivity and sinking velocity of phytoplankton.

In this situation, from Eq. (23), we have

$$\sigma \left(\sigma + \sqrt{\frac{K_{v2}}{\mu_{m} - \varepsilon}} \right) \left(\sigma - \sqrt{\frac{K_{v2}}{\mu_{m} - \varepsilon}} \right) = \frac{\mu_{m} K_{v2}}{(\mu_{m} - \varepsilon) \varepsilon K_{d}}. \tag{31}$$

The SCML thickens with a larger vertical diffusivity below the surface mixed layer (K_{v2}) , a smaller growth rate (μ_m) or a lower light attenuation coefficient (K_d) .

Especially, when $K_{v2} = 0$, we have $\sigma = 0$. In other words, for non-sinking phytoplankton $(w \rightarrow 0)$, when the vertical diffusivity below the surface mixed layer is very small $(K_{v2} \rightarrow 0)$, the SCML disappears. This indicates that there must be a vertical diffusion window sustaining non-sinking phytoplankton species in deep waters.

The third special situation occurs when $K_d\sigma$ (i.e., $\sigma/(K_d)^{-1}$) is too small to be considered in Eq. (18). This may occur in clear waters where the light attenuation coefficient is very small $(K_d\rightarrow 0)$, or in regions where the light penetration depth $(1/K_d)$ is much larger than a half-width of SCML (σ) . Very narrow (from several to tens of centimeters) SCML has been observed in clear, stratified lakes where the light penetration depths were from several to tens of meters (Fee, 1976; Camacho, 2006).

In this situation, Eq. (18) can be modified to

$$w\sigma + K_{v2} = 0. {32}$$

Clearly, when $K_{v2} = 0$, w=0, this equation has infinitely many solutions. This means in stable, clear waters with a predominance of small cells, the deep SCML can occur with different thicknesses. For example, in the basin of South China Sea, <3 μ m phytoplankton (such as *Prochlorococcus*, *Synechococcus*, picoeukaryotes, etc.) are the dominant species in SCMLs (Takahashi and Hori, 1984; Liu et al., 2007) with variable thicknesses (Lee Chen, 2005; Chen et al., 2006).

4.2 Limitation and potential application

To make the complex problem (SCM phenomenon) tractable, the ecosystem dynamical equations adopted in this study are judiciously simplified. For example, a constant eddy diffusivity is assumed in the surface mixed layer and below this layer, respectively. Many processes (turbulence, internal waves, storms, slant-wise and vertical convection) in upper ocean dynamics are not captured in the model system. The assumption of steady state will be broken during episodic events of strong physical forcing, nutrient injection, or blooms (Fennel and Boss, 2003). Similarly the biological representation is also extremely limited. We neglect food-web and microbial loop dynamics (detritus, dissolved organic matter, and zooplankton are not included explicitly), and assume all loss processes, except sinking, to be linearly proportional to phytoplankton. The sinking velocity of phytoplankton is assumed to

474 be constant with depths, excluding the effects of temperature and density gradients. Our model also neglects some feedback mechanisms, like the effect of phytoplankton 475 on light attenuation. Although these are important aspects, their addition is unlikely to 476 change our conclusions qualitatively under the boundary conditions chosen in this 477 study (Fennel and Boss, 2003). 478 Without considering nutrient input directly to the surface mixed layer, Chl a 479 concentration within it is assumed to be nearly zero. The SCML is assumed to occur 480 significantly deeper than the base of surface mixed layer, and the vertical gradient of 481 Chl a is assumed to be identically zero at the transition between the two layers. This 482 vertical profile of Chl a (Fig. 1) is assumed to be well fitted by a general Gaussian 483 function (Eq. 7), in which Chl a within the surface mixed layer is an approximation 484 for the tail of Gaussian function. The Gaussian assumption leads to the results that 485 the vertical diffusivity within the surface mixed layer has no roles on the SCM. 486 487 However, the assumption of a general Gaussian profile can be broken in several ways. If nutrient input directly to the mixed layer due to riverine inputs, surface runoff, or 488 atmospheric deposition, a positive Chl a concentration within the surface mixed layer 489 will be sustained (Mellard et al., 2011). In this case, the general Gaussian function 490 can be modified with a superimposition of a background which is either a constant 491 492 (Platt et al., 1988; Richardson et al., 2003) or a function of water depth (Uitz et al., 2006; Mignot et al., 2011). If the depth of surface mixed layer z_s is large, this allows 493 another way for the surface Chl a concentration being positive by extracting some of 494 the Chl a from the SCML (Beckman and Hense, 2007), then the vertical gradient of 495 Chl a may not be identically zero at the transition between the two layers. 496 Under the assumption of a constant loss rate, the lower compensation depth we got 497 from Eq. (11), the location where the growth rate of phytoplankton limited by light 498 equals the loss rate, is similar to the popular definition of compensation depth given 499 by Sverdrup (1953), below which no net growth occurs. This assumption is in the 500 heart of the Sverdrup's critical depth model and we now understand that it has 501 502 significant limitations (Behrenfeld and Boss, 2014). Particularly, the treatment of grazing loss, is, in the least, an oversimplification, though many numerical models 503 used a similar one (e.g., Klausmeier and Litchman, 2001; Fennel and Boss, 2003; 504 Huisman et al., 2006). Grazing loss depends strongly on phytoplankton and 505

zooplankton concentrations (it is an encounter based process) and, given that

- zooplankton can move, or, in the least, grow faster where more food is available, is
- unlikely to have a constant concentration distribution (Behrenfeld and Boss, 2014).
- Our model suggests that the condition for the existence of a SCM is that the growth
- rate under the limitation of light intensity, $\mu_m f(I_0)$, is larger than the loss rate, ε , in
- stratified water columns. Fennel and Boss (2003) found a similar result and pointed
- out that this condition for a SCM is general. Many numerical studies have reproduced
- 513 the SCM phenomenon, of which the condition of SCM occurrence met with variable
- values of the sinking velocity of phytoplankton and the mixing diffusivity
- 515 (Klausmeier and Litchman, 2001; Huisman et al., 2006; Mellard et al., 2011).
- Our two compartment system model reproduces some of the results of the more
- 517 complex model with three compartments (phytoplankton, nutrients, and detritus,
- Beckmann and Hense, 2007). For example, our model predicts that with fully
- recycling of the dead phytoplankton, the total Chl a concentration in water columns
- depends on the sinking velocity of phytoplankton and the vertical diffusivity, but
- independents on the growth rate and the loss rate of phytoplankton. Beckmann and
- Hense (2007) found similar results. Here, we go further to point out an interesting
- finding that the derivations of the total Chl a are irrespective of the form of the
- growth limiting function. Since growth functional forms in phytoplankton models are
- still debated in the literature (Haney, 1996; Ayata et al., 2013), this will be most
- helpful to estimate the vertical integrated Chl a and primary production.
- The relationships (in previous sections and in Appendices A and B) we derived can
- be used to compute missing model parameters (such as maximum growth rate μ_m ,
- loss rate ε , recycling rate α) which are difficult to obtain by on-site observation, if
- estimates of others are available. For example, Eq. (B4) allows us to obtain an
- estimate of the sinking velocity of phytoplankton from the measurement of SCM
- thickness and intensity, the nutrient concentration at water column depth, and the
- vertical diffusivity below the surface mixed layer.
- Our analytic solutions can in principle be tested through a comparison with
- observations: for example, the shape of profiles (the SCML thickness, depth, and
- intensity), expressed by the characteristic relationships (Eqs. 16-18, 22 and B4), the
- vertical integral of total subsurface Chl a concentration (Eqs. 21 and B3), the
- consistency of independent field estimates for sinking velocity, vertical diffusivity,

recycling rate and loss rate (Eqs. 21-22 and B3-B4).

We retrieve the three SCM characteristics from Eqs. (16-18, and 22) by combining remote sensing data (annual averaged values of surface light intensity I_0 and light attenuation coefficient K_d) and some parameters from published field and numerical studies (e.g., sinking velocity of phytoplankton w, vertical diffusivity below the surface mixed layer K_{v2} , loss rate ε , maximum growth rate μ_m). Table 2 lists the values of model parameters at three time-series stations in different ocean regions, i.e., the SEATS station, the HOT station, and the BATS (Bermuda Atlantic Time-Series Study) site in the Sargasso Sea, and the corresponding references. The estimated results and the observed values of the SCML thickness, depth and intensity at the three stations are shown in Fig. 2.

The estimated depths and thicknesses of the SCML agree reasonably well with the observations at all three stations. However, the intensities of the SCML are poorly estimated, implying other mechanisms (e.g., wind-driven nutrient pulse) supplying nutrients for the SCML, except upward diffusivity, for phytoplankton growth (Williams et al., 2013). This is the first try to estimate the depth, thickness and intensity of the SCML using parameters from satellite data and field studies. It should be noted that the estimation is sensitive to the used values of these environmental parameters. The values used in estimations above are representative for the averages over a large spatial or temporal scale, but they may not reflect the real values in a specific station. Even though disagreements could be associated with uncertainties from several sources, this type of try would give some idea of how real-world data could be incorporated into the model and thus be applied to the field (Pitarch et al. 2014).

5 Summary

- A general Gaussian function is assumed to represent a bell-shape vertical distribution of Chl a in stratified water columns. The function is incorporated into the ecosystem dynamical equations to determine three steady-state SCM characteristics and examine their dependence on environmental parameters such as vertical diffusivity, sinking
- velocity of phytoplankton, light attenuation coefficient.
- The maximum Chl a concentration occurs at or below the location of the maximum growth rates of phytoplankton determined by the vertical mixing time scale and the

571 time taken by a phytoplankton sinking through the length scale.

The depth of the SCML in steady state deepens logarithmically with an increase in surface light intensity, but shoals with increasing light attenuation coefficient, increasing vertical diffusivity below the surface mixed layer, increasing loss rate of

575 phytoplankton, and with decreasing sinking velocity of phytoplankton.

The thickness and intensity of the SCML are mainly influenced by nutrients supply, but independent of sea surface light intensity. The SCML gets thicker and stronger with a higher vertical diffusivity below the surface mixed layer. The intensity of SCML in steady state weakens as a result of expanding the SCML by a smaller sinking velocity of phytoplankton and a lower light attenuation coefficient.

In regions with a low vertical diffusivity, the SCML depth is inversely proportional to light attenuation coefficient, and is deeper than that in regions dominated by non-sinking phytoplankton. In clear and stable waters with a predominance of small cells, deeper SCMLs can occur with different thicknesses.

Upon potential risk of climate change, it is critical to accurately estimate the global and regional SCML-related primary production. However, the SCM characteristics cannot be detected by remote sensing satellites, which will restrict the application of satellite data in estimating primary production in a large temporal and spatial scale. The Argo float equipped with optical sensor has been developed to measure the distribution of particles and chlorophyll in the world's ocean (Mignot et al., 2014), but the data are still limited. The relationships we derived might help to estimate depth-integrated primary production using available data from satellite observations (incident light and light attenuation coefficient) when appropriate vertical estimates of growth rate and loss rate of phytoplankton, sinking velocity of phytoplankton and vertical diffusivity were adopted based on observations or model results. Again, the solutions could also help to compute environmental parameters that are difficult to obtain from on-site observation.

599 Appendix A

- In steady state, the net nutrient flux at any given depth (z) is equals to the net
- nutrients consumption by phytoplankton, then from steady-state of Eq. (2) we obtain
- Eq. (A1) below the surface mixed layer:

$$\int \left(\mu_{m} \min\left(f(I), g(N)\right) - \alpha \varepsilon\right) P(z) dz \approx K_{v2} \frac{dN(z)}{dz} \Big|_{z} \tag{A1}$$

- If $\mu_m \min(f(I), g(N)) \varepsilon > 0$, then $\mu_m \min(f(I), g(N)) \alpha \varepsilon > 0$ for $0 < \alpha \le 1$, we will
- have $\frac{dN}{dz} > 0$. That is, N(z) will increase with depth below the surface mixed layer.
- From the properties of the quadratic function in the right hand of Eq. (10), we have
- 607 $\mu_m \min(f(I), g(N)) \varepsilon > 0$ on the interval (z_{c1}, z_{c2}) . Hence, we have
- 608 $\mu_m \min(f(I), g(N)) \alpha \varepsilon > 0$ for $0 < \alpha \le 1$, then dN/dz > 0. In other words, N(z)
- 609 increases with depth on the interval (z_{c1}, z_{c2}) .
- According to Eq. (4), we know that g(N) is a monotonic increasing function on
- interval (z_{c1}, z_{c2}) , and f(I) is a monotonic decreasing function on interval (z_{c1}, z_{c2}) .
- It is well known that the stable SCML occurs in stratified water column only when
- the growth of phytoplankton in the surface mixed layer is nutrient-limited (Mellard et
- al., 2011; Ryabov et al., 2010). In other words, the limitation by nutrients g(N) is less
- than the limitation by light f(I) within the surface mixed layer, i.e., g(N) < f(I) for
- 616 $0 \le z \le z_s$.
- Because there is only one maximum in the growth rates of phytoplankton which
- occurs at the depth $z_0 = z_m \frac{w\sigma^2}{2K_{v2}}$, and $z_{c1} < z_0 < z_{c2}$ (Eq. 11), we arrive at

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$$\min(f(I), g(N)) = \begin{cases} g(N) & z_{c1} \le z \le z_0 \\ f(I) & z_0 \le z \le z_{c2} \end{cases}$$
 (A2)

620 and

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$$\max \left(\mu_m \min \left(f(I), g(N) \right) \right) = \left. \mu_m f(I) \right|_{z=z_0}. \tag{A3}$$

That is, the maximum growth rate occurs at the depth z_0 where is the transition

- 623 from nutrients limitation to light limitation, and the growth of phytoplankton is
- 624 light-limited below the depth z_0 .

625 Appendix B

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The dead phytoplankton is entirely recycled ($\alpha = 1$), and thus the system is closed. In

this case, at steady state Eq. (19) reduces to

$$w\frac{dP}{dz} = \frac{d^2}{dz^2} \left(K_{\nu} \left(P + N \right) \right) \tag{B1}$$

Integrating this equation twice from the surface to bottom boundary (z_b) gives

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$$w \int_0^{z_b} P(z) dz = K_{v1} (P+N) \Big|_0^{z_s} + K_{v2} (P+N) \Big|_{z_s+0}^{z_b}$$
 (B2)

Note that we have known that the SCML occurs only when the growth of

phytoplankton within the surface mixed layer is nutrient-limited, then we further

assume the surface nutrients value is negligible. Using the assumption of small Chl a

at the top and the bottom boundaries of the model domain, we obtain

$$h = \frac{K_{v2}}{w} N(z_b)$$
 (B3)

and the intensity of SCML is

$$P_{\text{max}} = \frac{K_{v2}}{\sqrt{2\pi}\sigma w} N(z_b)$$
 (B4)

where $N(z_b)$ is the nutrients concentration at depth z_b . Therefore, with $\alpha = 1$, the

intensity of SCML is affected by the ambient nutrients concentration below the

surface mixed layer. The total Chl a in the water column depends on the sinking

velocity of phytoplankton and the diffusivity, but it is independent on the growth rate

and loss rate of phytoplankton. Analogous results have been obtained by Liccardo et

al. (2013). Beckmann and Hense (2007) also found similar result by introducing an

explicit compartment for the detritus in their models.

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- Anderson, G. C.: Subsurface chlorophyll maximum in the northeast Pacific Ocean, Limnol. Oceanogr.,
- 654 14, 386-391, 1969.
- 655 Ardyna, M., Babin, M., Gosselin, M., Devred, E., Bélanger, S., Matsuoka, A., Tremblay, J. E.:
- Parameterization of vertical chlorophyll a in the Arctic Ocean: impact of the subsurface chlorophyll
- 657 maximum on regional, seasonal and annual primary production estimates, Biogeosciences, 10,
- 658 4383-4404, doi: 10.5194/bg-10-4383-2013, 2013.
- 659 Ayata, S., Lévy, M., Aumont, O., Sciandra, A., Sainte-Marie, J., Tagliabue, A., Bernard, O.:
- 660 Phytoplankton growth formulation in marine ecosystem models: should we take into account
- photo-acclimation and variable stoichiometry in oligotrophic areas? J. Marine Syst., 125, 29-40, 2013.
- Beckmann, A., Hense, I.: Beneath the surface: Characteristics of oceanic ecosystems under weak
- mixing conditions-A theoretical investigation, Prog. Oceanogr., 75, 771-796, 2007.
- Behrenfeld, M. J., Boss, E. S.: Resurrecting the ecological underpinnings of ocean plankton blooms,
- 665 Annu. Rev. Mar. Sci., 6, 167-194, 2014.
- 666 Bienfang, P. K., Harrison, P. J.: Sinking-rate response of natural assemblages of temperate and
- subtropical phytoplankton to nutrient depletion, Mar. Biol., 83, 293-300, 1984.
- Bowdon, K. F.: Physical oceanography of coastal waters. Limnol. Oceanogr., 30, 449-450, 1985.
- 669 Cai Y. M., Ning X. R., Liu C. G.: Synechococcus and Prochlorococcus growth and mortality rates in
- the northern China Sea: range of variations and correlation with environmental factors. Acta Ecol. Sin.
- 671 7, 2237-2246, 2006.
- 672 Camacho, A.: On the occurrence and ecological features of deep chlorophyll maxima (DCM) in
- 673 Spanish stratified lakes, Limnetica, 25, 453-478, 2006.
- 674 Chen, C. C., Shiah, F. K., Chung, S. W., Liu, K. K.: Winter phytoplankton blooms in the shallow
- 675 mixed layer of the South China Sea enhanced by upwelling, J. Marine Syst., 59, 97-110, 2006.
- 676 Cianca, A., Godoy, J. M., Martin, J. M., Perez Marrero, J., Rueda, M. J., Llinás, O., Neuer, S.:
- 677 Interannual variability of chlorophyll and the influence of low frequency climate models in the North
- Atlantic subtropical gyre, Global Biogeochem. Cy., 26, 2012.
- 679 Cianca, A., Helmke, P., Mouriño, B., Rueda, M. J., Llinás, O., Neuer, S.: Decadal analysis of
- 680 hydrography and in situ nutrient budgets in the western and eastern North Atlantic subtropical gyre, J.
- 681 Geophys. Res., 112, C07025, 1-18, 2007.
- 682 Cloern, J. E., Grenz, C., Vidergar-Lucas, L.: An empirical model of the phytoplankton chlorophyll:
- 683 carbon ratio-the conservation factor between productivity and growth rate, Limnol. Oceanogr., 40,
- 684 1313-1321, 1995.
- 685 Cullen, J. J., Eppley, R. W.: Chlorophyll maximum layers of the Southern California Bight and
- possible mechanisms of their formation and maintenance, Oceanol. Acta, 1, 23-32, 1981.
- 687 Cullen, J. J.: The deep chlorophyll maximum: comparing vertical profiles of chlorophyll a, Can. J. Fish.
- 688 Aquat. Sci., 39, 791-803, 1982.
- Dekshenieks, M. M., Donaghay, P. L., Sullivan, J. M., Rines, J. E., Osborn, T. R., Twardowski, M. S.:
- 690 Temporal and spatial occurrence of thin phytoplankton layers in relation to physical processes, Mar.
- 691 Ecol-Prog. Ser., 223, 61-71, 2001.
- 692 Denman, K. L., Gargett, A. E.: Time and space scales of vertical mixing and advection of
- 693 phytoplankton in the upper ocean, Limnol. Oceanogr., 28, 801-815, 1983.
- Fee, E. J.: The vertical and seasonal distribution of chlorophyll in lakes of the Experimental Lakes Area,
- 695 northwestern Ontario: Implications for primary production estimates. Limnol. Oceanogr., 26, 767-783,
- 696 1976.
- Fennel, K., Boss, E.: Subsurface maxima of phytoplankton and chlorophyll: Steady-state solutions
- from a simple model, Limnol. Oceanogr., 48, 1521-1534, 2003.

- 699 Gabric, A. J., Parslow, J.: Effect of physical factors on the vertical distribution of phytoplankton in
- eutrophic coastal waters, Aust. J. Mar. Fishwater Res., 40, 559-569, 1989.
- 701 Ghosal, S., Mandre, S.: A simple model illustrating the role of turbulence on phytoplankton blooms, J.
- 702 Math. Biol., 46, 333-346, 2003.
- Haney, J. D.: Modeling phytoplankton growth rates, J. Plankton Res., 18, 63-85, 1996.
- 704 Hanson, C. E., Pesant, S., Waite, A. M., Pattiaratchi, C. B.: Assessing the magnitude and significance
- of deep chlorophyll maxima of the coastal eastern Indian Ocean, Deep-Sea Res. Pt. II, 54, 884-901,
- 706 2007.
- Hense, I., Beckmann, A.: Revisiting subsurface chlorophyll and phytoplankton distributions, Deep-Sea
- 708 Res. Pt. I, 55, 1193-1199, 2008.
- 709 Hodges, B. A., Rudnick, D. L.: Simple models of steady deep maxima in chlorophyll and biomass,
- 710 Deep-Sea Res. Pt. I, 51, 999-1015, 2004.
- 711 Hood, R. R., Bates, N. R., Capone, D. G., Olson, D. B.: Modeling the effect of nitrogen fixation on
- carbon and nitrogen fluxes at BATS, Deep-Sea Res. Pt. II, 48, 1609-1648, 2001.
- Huisman, J., Sommeijer, B.: Maximal sustainable sinking velocity of phytoplankton, Mar. Ecol-Prog.
- 714 Ser., 244, 39-48, 2002.
- Huisman, J., Thi, N., Karl, D. M., Sommeijer, B.: Reduced mixing generates oscillations and chaos in
- the oceanic deep chlorophyll maximum, Nature, 439, 322-325, 2006.
- 717 Jamart, B. M., Winter, D. F., Banse, K.: Sensitivity analysis of a mathematical model of phytoplankton
- 718 growth and nutrient distribution in the Pacific Ocean off the northwestern US coast, J. Plankton Res., 1,
- **719** 267-290, 1979.
- 720 Klausmeier, C. A., Litchman, E.: Algal games: The vertical distribution of phytoplankton in poorly
- 721 mixed water columns, Limnol. Oceanogr., 8, 1998-2007, 2001.
- 722 Lee Chen, Y.: Spatial and seasonal variations of nitrate-based new production and primary production
- 723 in the South China Sea, Deep-Sea Res. Pt. I, 52, 319-340, 2005.
- 724 Lewis, M. R., Cullen, J. J., Platt, T.: Phytoplankton and thermal structure in the upper ocean:
- consequences of nonuniformity in chlorophyll profile, J. Geophys. Res., 88, 2565-2570, 1983.
- Lewis, M. R., Harrison, W. G., Oakey, N. S., Hebert, D., Platt, T.: Vertical nitrate fluxes in the
- 727 oligotrophic ocean, Science, 234, 870-873, 1986.
- 728 Liccardo, A., Fierro, A., Iudicone, D., Bouruet-Aubertot, P., Dubroca, L.: Response of the Deep
- 729 Chlorophyll Maximum to fluctuations in vertical mixing intensity, Prog. Oceanogr., 2013, 33-46, 2013.
- 730 Liu, H., Chang, J., Tseng, C. M., Wen, L. S., Liu, K. K.: Seasonal variability of picoplankton in the
- Northern South China Sea at the SEATS station, Deep-Sea Res. Pt. II, 54, 1602-1616, 2007.
- 732 Lu, Z., Gan, J., Dai, M., Cheung, A.: The influence of coastal upwelling and a river plume on the
- subsurface chlorophyll maximum over the shelf of the northeastern South China Sea, J. Marine Syst.,
- 734 82, 35-46, 2010.
- Mara On, E., Holligan, P. M.: Photosynthetic parameters of phytoplankton from 50°N to 50°S in the
- 736 Atlantic Ocean, Mar. Ecol.-Prog. Ser., 176, 191-203, 1999.
- 737 Mellard, J. P., Yoshiyama, K., Litchman, E., Klausmeier, C. A.: The vertical distribution of
- phytoplankton in stratified water columns, J. Theor. Biol., 269, 16-30, 2011.
- 739 Mignot, A., Claustre, H., D'Ortenzio, F., Xing, X., Poteau, A., Ras, J.: From the shape of the vertical
- profile of in vivo fluorescence to Chlorophyll-a concentration, Biogeosciences, 8, 2391-2406,
- 741 doi:10.5194/bg-8-2391-2011, 2011.
- 742 Mignot, A., Claustre, H., Uitz, J., Poteau, A., D'Ortenzio, F., Xing, X.: Understanding the seasonal
- dynamics of phytoplankton biomass and the deep chlorophyll maximum in oligotrophic environments:
- A Bio-Argo float investigation, Global Biogeochem. Cy., 28, 2013G-4781G, 2014.
- 745 Morel, A.: Optical modeling of the upper ocean in relation to its biogenous matter content (case I
- 746 waters), J. Geophys. Res., 93, 749-768, 1988.

- Ono, S., Ennyu, A., Najjar, R. G., Bates, N. R.: Shallow remineralization in the Sargasso Sea estimated
- 748 from seasonal variations in oxygen, dissolved inorganic carbon and nitrate, Deep-Sea Res. Pt. II, 48,
- 749 1567-1582, 2001.
- Oschlies, A.: Model-derived estimates of new production: New results point towards lower values,
- 751 Deep-Sea Res. Pt. II, 48, 2173-2197, 2001.
- Pedrós-Alió, C., Calderón-Paz, J. I., Guixa-Boixereu, N., Estrada, M., Gasol, J. M.: Bacterioplankton
- 753 and phytoplankton biomass and production during summer stratification in the northwestern
- 754 Mediterranean Sea, Deep-Sea Res. Pt. I, 46, 985-1019, 1999.
- Pérez, V., Fernández, E., Mara Ón, E., Morán, X., Zubkov, M. V.: Vertical distribution of
- 756 phytoplankton biomass, production and growth in the Atlantic subtropical gyres, Deep-Sea Res. Pt. I,
- 757 53, 1616-1634, 2006.
- 758 Pitarch, J., Odermatt, D., Kawka, M., Wüest, A.: Retrieval of vertical particle concentration profiles by
- optical remote sensing: a model study, Opt. Express, 22, A947-A959, 2014.
- Platt, T., Sathyendranath, S., Caverhill, C. M., Lewis, M. R.: Ocean primary production and available
- light: further algorithms for remote sensing, Deep-Sea Res. Pt. I, 35, 855-879, 1988.
- Prairie, J. C., Franks, P. J. S., Jaffe, J. S., Doubell, M. J., Yamazaki, H.: Physical and biological
- controls of vertical gradients in phytoplankton, Limnol. Oceanogr., 1, 75-90, 2011.
- Raven, J. A., Richardson K.: Photosynthesis in marine environments. In: Topics in Photosynthesis,
- 765 Elsevier, Edited by N. R. Baker and Long S., 7, 337-399, 1986.
- 766 Richardson, A. J., Silulwane, N. F., Mitchell-Innes, B. A., Shillington, F. A.: A dynamic quantitative
- approach for predicting the shape of phytoplankton profiles in the ocean, Prog. Oceanogr., 59, 301-319,
- 768 2003.
- Riley, G. A., Stommel, H., Bumpus, D. F.: Quantitative ecology of the plankton of the western North
- 770 Atlantic, Bulletin Bingham Oceanogra. Collect., 12, 1-69, 1949.
- 771 Ryabov, A. B., Rudolf, L., Blasius, B.: Vertical distribution and composition of phytoplankton under
- the influence of an upper mixed layer, J. Theor. Biol., 263, 120-133, 2010.
- 773 Salihoglu, B., Garçon, V., Oschlies, A., Lomas, M. W.: Influence of nutrient utilization and
- 774 remineralization stoichiometry on phytoplankton species and carbon export: A modeling study at
- 775 BATS, Deep-Sea Res. Pt. I, 55, 73-107, 2008.
- 576 Sharples, J., Moore, C. M., Rippeth, T. P., Holligan, P. M., Hydes, D. J., Fisher, N. R., Simpson, J. H.:
- 777 Phytoplankton distribution and survival in the thermocline, Limnol. Oceanogr., 46, 486-496, 2001.
- 778 Siswanto, E., Ishizaka, J., Yokouchi, K.: Estimating chlorophyll-a vertical profiles from satellite data
- and the implication for primary production in the Kuroshio front of the East China Sea, J. Oceanogr.,
- **780** 61, 575-589, 2005.
- 781 Sullivan, J. M., Donaghay, P. L., Rines, J. E.: Coastal thin layer dynamics: consequences to biology
- 782 and optics, Cont. Shelf Res., 30, 50-65, 2010.
- Sverdrup, H. U.: On conditions for the vernal blooming of phytoplankton, J. Cons. int. Explor. Mer., 18,
- 784 287-295, 1953.
- 785 Takahashi, M., Hori, T.: Abundance of picophytoplankton in the subsurface chlorophyll maximum
- layer in subtropical and tropical waters, Mar. Biol., 79, 177-186, 1984.
- 787 Tjiputra, J. F., Polzin, D., Winguth, A. M.: Assimilation of seasonal chlorophyll and nutrient data into
- an adjoint three dimensional ocean carbon cycle model: Sensitivity analysis and ecosystem parameter
- optimization, Global Biogeochem. Cv., 21, GB1001, 1-13, 2007.
- 790 Uitz, J., Claustre, H., Morel, A., Hooker, S. B.: Vertical distribution of phytoplankton communities in
- 791 open ocean: An assessment based on surface chlorophyll, J. Geophys. Res, 111, C8005, 2006.
- 792 Varela, R. A., Cruzado, A., Tintoré, J.: A simulation analysis of various biological and physical factors
- influencing the deep-chlorophyll maximum structure in oligotrophic areas, J. Marine Syst., 5, 143-157,
- **794** 1994.

- 795 Weston, K., Fernand, L., Mills, D. K., Delahunty, R., Brown, J.: Primary production in the deep
- chlorophyll maximum of the central North Sea, J. Plankton Res., 27, 909-922, 2005.
- 797 Williams, C., Sharples, J., Mahaffey, C., Rippeth, T.: Wind driven nutrient pulses to the subsurface
- 798 chlorophyll maximum in seasonally stratified shelf seas, Geophys. Res. Lett., 2013.
- 799 Wu Y. P., Gao K. S.: Photosynthetic response of surface water phytoplankton assemblages to different
- wavebands of UV radiation in the South China Sea, Acta Oceanol. Sin., 5, 146-151, 2011.
- Yoshiyama, K., Mellard, J. P., Litchman, E., Klausmeier, C. A.: Phytoplankton competition for
- nutrients and light in a stratified water column, Am. Nat., 174, 190-203, 2009.

804 Figure 1

Growth limitation by light and nutrients

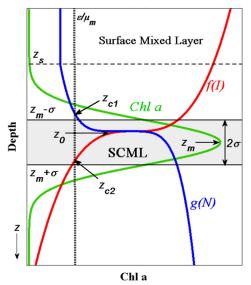


Fig. 1 Schematic picture of Chl a distribution under the limitation by light and nutrient in stratified water columns (green solid line is Chl a concentration as a function of depth; red solid line is the growth limiting term with respect to light, f(I); blue solid line is the growth limiting term with respect to nutrients, g(N); horizontal dashed line represents the depth of surface mixed layer, z_s ; horizontal solid lines indicate the locations of the upper- and lower-SCML, z_m - σ , z_m + σ , respectively; vertical dotted line is the ratio of loss rate to maximum growth rate, ε/μ_m ; z_{c1} and z_{c2} refer to the two compensation depths where $\mu_m g(N) = \varepsilon$ and $\mu_m f(I) = \varepsilon$, respectively; z_0 and z_m indicate the depths of maximum in growth rates and in Chl a concentrations, respectively; double arrow represents the thickness of the SCML, z_0

815 Figure 2

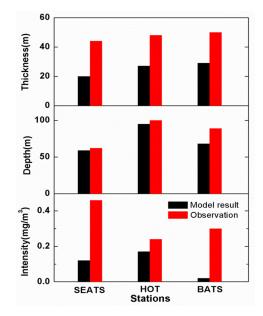


Fig. 2 Comparisons of the model results and observations (in terms of thickness, depth, and intensity of SCML) at SEATS, HOT, and BATS (black columns represent the model results, red columns are the observations at the three stations which were fitted by Gaussian function using annually averaged data obtained from http://www.odb.ntu.edu.tw/, http://bats.bios.edu/,

respectively)

Model parameters (↑)	2σ	z_m	P_{max}	h
(Surface light intensity)	-	↑	-	-
K_I (Half-saturation constant of light limited growth)	-	\downarrow	-	-
K_{v2} (Vertical diffusivity below surface mixed layer)	↑	\downarrow	↑	↑
w (Sinking velocity of phytoplankton)	\downarrow	\downarrow	↑	-
K_d (Light attenuation coefficient)	\downarrow	\downarrow	↑	-
	↓*	\downarrow	/	\downarrow
ε (Loss rate of phytoplankton)	^**	\downarrow	\downarrow	\downarrow
α (Nutrient recycling coefficient)	-	-	↑	↑
$\frac{dN}{dz}\Big _{z=z_b}$ Nutrient gradient at the lower boundary of SCML	-	-	↑	↑
K_N (Half-saturation constant of nutrient limited growth)	-	-	-	-
K_{vI} (Vertical diffusivity in surface mixed layer)	-	-	-	-
μ_{max} (Maximum growth rate of phytoplankton)	/	/	/	/

 $[\]uparrow$ indicates increase, \downarrow indicates decrease, - indicates no effect, / indicates no straightforward result, * indicates a result when $\varepsilon < \mu_{max}/2$, and ** indicates a result when $\varepsilon > \mu_{max}/2$.

Table 2 Parameter values at SEATS, HOT, and BATS

Parameters	Units	Values at Stations			
		SEATS	НОТ	BATS	
I_0	μ mol photos m^{-2} s^{-1}	700 (1, 2)	550 (1, 3)	448 (1, 4)	
K_d	m^{-1}	0.052 (1, 5)	0.04 (1, 3)	0.042 (1, 4)	
K_{v2}	$m^2 s^{-1}$	5*10 ^{-5 (6)}	5*10 ^{-5 (3)}	1*10 ^{-4 (7, 8)}	
μ_{max}	d^{-I}	1.2 (9, 10)	0.96 (3)	1 (11)	
K_I	μ mol photos m^{-2} s^{-1}	40 (12)	20 (3)	20 (3, 12, 13)	
3	d^{I}	0.5 (9, 10)	0.24 (3)	0.5 (14)	
α	-	0.3 (10)	0.5 (3)	0.16 (8)	
W	$m d^{-1}$	1 (15)	1 (3, 15)	2 (8)	
dN/dz at depth of z_b	mmol N m ⁻⁴	0.1 (16)	0.05 (17, 18)	0.02 (19, 20)	
z_b	m	200	200	200	

Superscripts refer to the references that provide the source for the parameter value and the citations are as follows: ⁽¹⁾http://oceandata.sci.gsfc.nasa.gov/SeaWiFS/Mapped/Annual/9km/; ⁽²⁾Wu and Gao, 2011; ⁽³⁾Huisman et al., 2006; ⁽⁴⁾Varela et al., 1994; ⁽⁵⁾Lee Chen et al., 2005; ⁽⁶⁾Lu et al., 2010; ⁽⁷⁾Hood et al., 2001; ⁽⁸⁾Salihoglu et al., 2008; ⁽⁹⁾Cai et al., 2006; ⁽¹⁰⁾Liu et al., 2007; ⁽¹¹⁾Ayata et al., 2013; ⁽¹²⁾Raven and Richardson, 1986; ⁽¹³⁾Mara On and Holligan, 1999; ⁽¹⁴⁾Tjiputra et al., 2007; ⁽¹⁵⁾Bienfang and Harrison, 1984; ⁽¹⁶⁾Chen et al., 2006; ⁽¹⁷⁾Fennel and Boss, 2003; ⁽¹⁸⁾Hense and Beckmann, 2008; ⁽¹⁹⁾Cianca et al., 2007; ⁽²⁰⁾Cianca et al., 2012.