Seasonal photoacclimation in the North Pacific Transition Zone

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Abstract

The Transition Zone Chlorophyll Front (TZCF) is a dynamic region of elevated chlorophyll concentrations in the Northeast Pacific that migrates from a southern winter (February) extent of approximately 30°N to a northern summer (August) extent of approximately 40°N. The transition zone has been highlighted as important habitat for marine animals and fisheries. We re-examine the physical and biological drivers of seasonal TZCF variability using a variety of remote sensing, reanalysis, and in-situ datasets. Satellite-based remote sensing estimates of chlorophyll and carbon concentrations show that seasonal TZCF migration primarily reflects a seasonal increase in the chlorophyll to carbon ratio, rather than changes in phytoplankton carbon. We use our data compilation to demonstrate how the seasonality of light and nutrient fluxes decouple chlorophyll and carbon seasonality at the transition zone latitudes. Seasonal mixed-layer-averaged light availability is positively correlated with carbon and negatively correlated with chlorophyll through the transition zone, while climatological nitrate profiles show that chlorophyll to carbon ratios are facilitated by wintertime nitrate entrainment. These empirical results are consistent with physiological data and models describing elevated chlorophyll to carbon ratios in low light, nutrient-replete environments, demonstrating the importance of latitudinal structure in interpreting seasonal chlorophyll dynamics at the basin scale.

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8 Key Points:

- Seasonal variations in North Pacific chlorophyll show a distinct latitudinal structure in
 phase and magnitude
- Chlorophyll concentrations are negatively correlated with carbon concentrations in the transition zone (30-40°N)
- Latitudinally varying nutrient and light supply drive chlorophyll and carbon covariation
 via photoacclimation

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17 concentrations in the Northeast Pacific that migrates from a southern winter (February) extent of

approximately 30°N to a northern summer (August) extent of approximately 40°N. The

19 transition zone has been highlighted as important habitat for marine animals and fisheries. We

20 re-examine the physical and biological drivers of seasonal TZCF variability using a variety of

21 remote sensing, reanalysis, and *in-situ* datasets. Satellite-based remote sensing estimates of

chlorophyll and carbon concentrations show that seasonal TZCF migration primarily reflects a
 seasonal increase in the chlorophyll to carbon ratio, rather than changes in phytoplankton carbon.

We use our data compilation to demonstrate how the seasonality of light and nutrient fluxes

decouple chlorophyll and carbon seasonality at the transition zone latitudes. Seasonal mixed-

26 layer-averaged light availability is positively correlated with carbon and negatively correlated

27 with chlorophyll through the transition zone, while climatological nitrate profiles show that

chlorophyll to carbon ratios are facilitated by wintertime nitrate entrainment. These empirical

results are consistent with physiological data and models describing elevated chlorophyll to

30 carbon ratios in low light, nutrient-replete environments, demonstrating the importance of

31 latitudinal structure in interpreting seasonal chlorophyll dynamics at the basin scale.

32

33 Plain Language Summary

34 Satellite-observed marine chlorophyll concentrations are regularly interpreted as phytoplankton

35 carbon. However, the chlorophyll content of cells can vary due to several environmental factors,

thus complicating the interpretation of satellite-observed chlorophyll variability. In this study, we

37 examine the relationship between chlorophyll and phytoplankton carbon in the Northeast Pacific

- a region that has been highlighted as important habitat for marine animals. We find that

39 satellite-observed chlorophyll seasonality is strongly correlated with light and nutrient

40 availability but relatively uncorrelated with phytoplankton carbon due to changes in the

41 chlorophyll to carbon ratio. Deep winter mixed layers are the primary physical factor driving the

seasonal cycle in light and nutrient availability. These results provide a new perspective on
 marine ecosystem productivity in the Northeast Pacific and demonstrate how latitudinal

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 differences in the seasonality of light and nutrient fluxes connect chlorophyll and carbon

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46

47 **1 Introduction**

48 The North Pacific Transition Zone Chlorophyll Front (TZCF) is a basin-scale chlorophyll feature

49 noted in early satellite observations of ocean colour (Glover et al., 1994). The front exhibits

50 marked seasonality, moving from a summertime northern position at approximately 40°N, to a

51 wintertime southern position at approximately 30°N (Figure 1; Follett et al., 2021). The

52 transition zone (30-40°N) has been repeatedly highlighted as a congregation area for marine

animals and fisheries (Block et al., 2011; Hazen et al., 2013; Kappes et al., 2010; Polovina et al.,

54 2017; Xu et al., 2017). Tagging data, where animals are fitted with geo-locators, have shown

55 migratory and feeding behaviors throughout the transition zone, including commercially

- ⁵⁶ important tuna (Polovina et al., 2017; Xu et al., 2017), seabirds (Block et al., 2011; Kappes et al.,
- 57 2010), and a variety of other marine animals (Block et al., 2011; Hazen et al., 2013).
- 58 Since early satellite observations, several studies have investigated the physical, chemical, and
- 59 biological drivers of TZCF variability, with authors noting the uniqueness of the apparent
- 60 wintertime productivity maximum at mid-latitudes (Ayers & Lozier, 2010; Bograd et al., 2004;
- 61 Chai et al., 2003; Glover et al., 1994; Le et al., 2019; Polovina et al., 2001, 2017). Glover &
- 62 Mcclain (1994) and Chai et al. (2003) suggested that deep winter mixed layers entrain nitrate and
- drive the apparent wintertime productivity maximum, while Ayers & Lozier (2010) and Le et al.
- 64 (2019) suggested a stronger role for southward Ekman transport of nitrate.
- 65 Implicit in these earlier studies of TZCF dynamics is the assumption that the observed
- 66 chlorophyll variability corresponds to changes in phytoplankton carbon, while not considering
- 67 potential seasonal variations in chlorophyll to carbon ratios. Like other mid- and high-latitude
- environments, the transition zone is characterized by large seasonal cycles in surface irradiance
- and mixed layer depth which drives low wintertime light availability and elevated nutrient fluxes
- to the mixed layer. The seasonal mixed layer cycle acts reduces light supply due to the
- exponential extinction of light with depth while entraining nutrients if deep mixed layer
- 72 penetrate a strongly sloping nutricline (Evans et al., 1985). Physiological data and models show
- 73 that nutrient-replete, low light environments will increase the ratio of chlorophyll to carbon via
- 74 photoacclimation (Behrenfeld et al., 2016; Geider et al., 1996, 1998; Inomura et al., 2020; Laws
- 75 & Bannister, 1980; Westberry et al., 2008).
- 76 Behrenfeld et al., (2005) and others (e.g. Graff et al., 2016) leveraged correlations between
- satellite-observed backscatter and phytoplankton carbon to demonstrate physiologically
- interpretable decoupling of phytoplankton carbon and chlorophyll in response to light and
- 79 nutrient availability. From these relationships, we expect the North Pacific TZCF chlorophyll
- signal to, in part, reflect photoacclimation responses to seasonal variations in light and nutrient
- availability rather than variations in the underlying phytoplankton carbon. We test this
- 82 expectation by studying the latitudinal structure of surface chlorophyll and carbon variations
- using a variety of remote sensing, reanalysis, and *in-situ* datasets. We disentangle the role of photoacclimation in the observed TZCF chlorophyll dynamics by quantifying relationships
- between chlorophyll concentrations, carbon concentrations, net primary productivity, light
- availability, and mixed layer depths over the seasonal cycle. We examine these relationships
- across latitudes to identify the unique chlorophyll cycle in the transition zone. These results will
- help better understand ecosystem productivity in the North Pacific and improve our
- interpretation of the mid-latitude satellite chlorophyll record.
- 90

91 **2 Methods**

- 92 We assembled a variety of remote sensing, reanalysis, and *in-situ* datasets to examine the
- 93 latitudinal structure of chlorophyll and carbon variations and the dynamics of photoacclimation
- at the TZCF. Satellite remote sensing estimates of surface chlorophyll concentrations, carbon
- 95 concentrations, net primary productivity, surface photosynthetically active radiation (PAR), and
- 96 the diffuse attenuation coefficient (k_d) were obtained from the Oregon State Ocean Productivity
- 97 database (http://sites.science.oregonstate.edu/ocean.productivity/). These estimates have been
- gap-filled for missing observations due to cloud cover according to the algorithm described at
- 99 http://orca.science.oregonstate.edu/gap_fill.php. Chlorophyll concentrations are based on the

- 100 Generalized Inherent Optical Property Algorithm (GIOP) using MODIS observations (Werdell et
- al., 2013). Carbon concentrations are based on backscatter coefficients estimated via the GIOP
- algorithm, modeled via the functions given in Behrenfeld et al., (2005) and Westberry et al.,
- 103 (2008). We also repeated analyses using chlorophyll and carbon concentrations derived from the
- 104 alternative Garver-Siegel-Maritorena (GSM) ocean color inversion algorithm (Maritorena et al.,
- 105 2002) to ensure the consistently of results across inversion products. Net primary productivity
- 106 was taken from four models that differ in their parameterization of phytoplankton growth: the
- vertically generalized productivity model (VGPM; Behrenfeld & Falkowski, 1997), the carbon based productivity model (CBPM; Westberry et al., 2008), the Eppley vertically generalized
- 109 productivity model (EPVGPM;
- 110 http://sites.science.oregonstate.edu/ocean.productivity/eppley.model.php), and the Carbon,
- 111 Absorption, and Fluorescence Euphotic-resolving model (CAFÉ; Silsbe et al., 2016). Surface
- 112 PAR and k_d are estimated from MODIS observations (Son & Wang, 2015).
- 113 Reanalysis datasets of mixed layer depth were obtained from the simple ocean data assimilation
- 114 product (SODA; version SODA3 12.2; available at https://www.soda.umd.edu/) and the Hybrid
- 115 Coordinate Ocean Model (HyCOM; hindcast version GLBu0.08; available at
- 116 http://sites.science.oregonstate.edu/ocean.productivity). SODA mixed layers were linearly
- 117 interpolated from the original five-day interval to an eight-day interval to match the satellite and
- 118 HyCOM datasets obtained from the Oregon State Productivity Database. We obtained mixed
- layer estimates from Argo based on the methods of Holte et al., (2017)
- 120 (http://mixedlayer.ucsd.edu/). In the supplementary material we demonstrate a high correlation
- between SODA, HyCOM, and Argo mixed layer estimates (Supplementary Figure S1),
- 122 indicating robustness across mixed layer depth products. All satellite and reanalysis datasets
- 123 were bilinearly interpolated onto a common 0.5° grid. Climatological nitrate observations were
- taken from the World Ocean Atlas, version WOA2018 (Garcia et al., 2019). Mixed layer-
- 125 averaged irradiance was calculated according to

126
$$\bar{I} = \frac{1}{z_{ml}} \int_{z_{ml}}^{0} I(z) dz = \frac{1}{k_d z_{ml}} I_0 e^{-k_d z_{ml}} (1 - e^{-k_d z_{ml}}),$$

where I(z) is the scaler irradiance at depth z, k_d is the diffuse attenuation coefficient, z_{ml} is the depth of the mixed layer, and I_0 is the PAR incident at the sea surface. Entrainment nitrate fluxes

129 into the mixed layer during mixed layer deepening were calculated according to

$$F_N = \frac{dz_{ml}}{dt}(N_0 - N_{ml}),$$

131 where $\frac{dz_{ml}}{dt}$ is the entrainment velocity, N_0 is the nitrate concentration one meter below the mixed 132 layer according a linear interpolation of World Ocean Atlas nitrate profiles, and N_{ml} is the mixed

- 133 layer-averaged nitrate concentration.
- 134 To investigate the latitudinal structure of chlorophyll and carbon dynamics, we binned
- observations according to six latitude bands across the Northeast Pacific, defined by longitudinal
- bounds of -180°W to -115°W. Latitudinal bands were defined by the intervals 0-10°N, 10-20°N,
- 137 20-30°N, 30-40°N, 40-50°N, 50-60°N. The western bound of -180°W was adopted to avoid
- 138 influence of the western boundary current which imparts stochastic variability on the physical
- 139 and chemical properties compared to the more stable latitudinal structure observed in the central
- 140 and eastern reaches of the basin.

141 **3 Results**

142 *3.1 Covariation between chlorophyll and carbon*

The seasonal TZCF migration is readily observed by comparing chlorophyll maps for the months of August and February (**Figure 1a,d**). August corresponds to the month of the maximum northward extent at approximately 40°N, while February corresponds to the month of maximum southward extent at approximately 30°N. Transition zone chlorophyll concentrations are elevated three- to five-fold during the February chlorophyll maximum, relative to August. While February is the month of maximum transition zone chlorophyll, concentrations are elevated over the months of January-February-March which we hereafter refer to as transition zone winter.

150 Climatological maps demonstrate that February transition zone chlorophyll does not spatially

151 correlate to carbon concentrations across the basin (**Figure 1**). The transition zone shows no

appreciable increase in February carbon concentrations relative to August. February carbon

153 concentrations are also depressed to the north and south of the transition zone, in contrast to

chlorophyll which remains elevated to the north. The seasonal decoupling between chlorophyll

and carbon becomes clear in the climatological distribution of the chlorophyll to carbon ratio

156 (Figure 1c,f), where ratios are elevated roughly five-fold in winter from the transition zone

northward. Primary productivity models also disagree on the magnitude and spatial pattern of
 wintertime transition zone productivity with no consistent spatial correlation with chlorophyll

(Supplementary Figure S2). Notably, the two productivity models that include

160 photoacclimation processes (CBPM and CAFÉ) estimate productivity south of the transition

161 zone to be more than double that of productivity models that do not include photoacclimation

162 (VGPM and Epply VGPM), again reflecting the decoupling of chlorophyll and carbon, here with

163 consequences for satellite-based productivity estimates.

164 The temporal dynamics of chlorophyll and carbon concentration time series clearly demonstrate

165 the latitudinal dependence of covariation between the two variables (**Figure 2**). Over the

seasonal cycle, chlorophyll concentrations in the transition zone are negatively correlated with

167 carbon concentrations (r=-0.40, **Figure 2c**). Negative correlations between chlorophyll and

168 carbon extend from 10-40°N, with the strongest negative correlation at 20-30°N (r=-0.77; Figure

169 **2d**). However, mean transition zone chlorophyll concentrations are several-fold higher than more 170 southern latitudes with a seasonal chlorophyll range of approximately 0.2 mgChl/m^3 ,

highlighting the strength of the TZCF signal. Across latitudes, the seasonal correlation between

chlorophyll and productivity is also negative in the transition zone when averaging over four

primary productivity models (r=-0.11), with positive correlations between chlorophyll and

productivity at more northern and southern latitudes (**Supplementary Figure S3**). Across years,

negative correlations between chlorophyll and carbon are driven by a consistent offset in the

timing of their respective seasonal maxima, reflecting the chlorophyll peak in photoacclimated

177 low light conditions (**Figure 3**). All latitudes show an average lag between chlorophyll and

carbon maxima, with chlorophyll consistently peaking prior to carbon. The average lag across
 latitudes is 90 days with a standard deviation of 95 days. The most consistent lag was found in

the transition zone, with a mean offset of 93 days and a standard deviation of 25 days across

years (**Figure 3**). Taken together, the negative correlations between transition zone chlorophyll

and carbon concentrations demonstrate the unique latitudinal structure of the chlorophyll cycle

across the North Pacific. The transition zone exhibits a high-amplitude chlorophyll cycle that is
 largely independent of variations in phytoplankton carbon.

185 *3.2 Relationships with mixed layer depth and light availability*

The seasonal covariation of chlorophyll and carbon across latitude can be interpreted in terms of 186 latitude-specific responses to seasonal mixed layer depth variability and the resulting impact on 187 nutrient and light availability (Figure 4). Across latitudes, we find a consistently negative 188 relationship between mixed layer depth and carbon, with the strength of the relationship 189 decreasing southward (Figure 4a-f). In contrast, we find both positive and negative correlations 190 191 between mixed layer depth and chlorophyll across latitudes, with the largest positive slope in the transition zone (Figure 4i). The consistent positive slope of the transition zone chlorophyll 192 193 relationship combines with a large range of seasonal mixed layer depths (20-145m) to a yield a 194 transition zone chlorophyll cycle that is well predicted from a linear relationship with mixed

- 195 layer depth.
- 196 The correlation of mixed layer and light availability is shown in **Figure 5**. Large seasonal cycles
- in surface irradiance and mixed layer depth are apparent at mid- and high-latitudes (Figure 5a-
- **b**). The mean surface irradiance (around which the seasonal cycle oscillates) decreases
- northward, with the summer maximum surface irradiance at 50-60°N roughly equal to
- 200 wintertime light availability in the tropics. Combining surface irradiance, mixed layer depth, and
- satellite estimates of the light attenuation coefficient k_d , we find that the largest seasonal cycle in
- mixed layer-averaged irradiance occurs in the transition zone, with a climatological seasonal amplitude of approximately 170μ Mol guanta/m²/s and a wintertime light availability less than
- 10μ Mol guanta/m²/s (**Figure 5c**). The minimal light availability in the transition zone winter
- thus approaches the minimum compensation irradiance of photosynthesis (Geider et al., 1986;
- 206 Venables & Moore, 2010), highlighting the severe light limitation experienced by the wintertime
- 207 transition zone phytoplankton community.
- 208 The distinct responses of transition zone carbon and chlorophyll to mixed layer depth and light
- availability are shown spatially by mapping the seasonal correlation coefficients across the basin
- (Figure 6). The correlation of mixed layer depth vs. carbon concentration is negative across most
- of the basin, with a band of weaker correlations in the transition zone (**Figure 6a**). Conversely,
- correlations of mixed layer depth vs. chlorophyll are weak across much of the basin, except for
- the band of strong negative correlations across the transition zone (**Figure 6b**). These patterns
- further demonstrate that deepening winter mixed layers drive increases in chlorophyll
- concentrations in the transition zone that are uncorrelated with variations in carbon. In terms of
- mixed layer-averaged irradiance, we see weak positive correlations with carbon north of 20°N (r
- ~ 0.20) and weak negative correlations southward (r ~ -0.10; Figure 6c). In contrast to carbon and consistent with photoacclimation, chlorophyll strongly correlates with mixed layer-averaged
- 219 irradiance throughout the transition zone, with correlations between -0.5 and -0.7 (**Figure 6d**).
- 220 Zonally-averaged correlations reiterate this picture, showing diverging responses of carbon and
- chlorophyll with respect to mixed layer depth and mixed layer-averaged irradiance in the
- transition zone (**Figure 7**). We note that the correlation analysis was repeated using the
- alternative mixed layer reanalysis from SODA (Supplementary Figures S4 and S5). We found

- no appreciable difference in the magnitudes or spatial structure of the calculated correlations,
- indicating robustness in these relationships across mixed layer depth estimates.

226 Re-expressing relationships between light, carbon, and chlorophyll in terms of the chlorophyll to

227 carbon ratio (Chl:C) reveals a strong nonlinear association between Chl:C and mixed layer-

- 228 averaged irradiance, driven by nonlinear increases in Chl:C at the lowest light availability
- (Figure 8). The relationship between Chl:C and mixed layer-averaged irradiance is well-
- described by a negative exponential with light supply, like those described previously based on
- physiological principles (Behrenfeld et al., 2005; Graff et al., 2016; Jackson et al., 2017;
- 232 Sathyendranath et al., 2020). The nonlinear response of Chl:C occurs in the light-limited regime
- encountered during transition zone winter.

234 *3.2 Mixed layer driven nutrient availability*

235 Deepening mixed layers can also act to entrain nutrients as deepening mixed layers penetrate

- layers of elevated nutrient concentrations. The resulting nutrient supply can then further enhancechlorophyll to carbon ratios due to the nitrogen requirement of chlorophyll and light harvesting
- 237 emotophyn to carbon ratios due to the introgen requirement of emotophyn and light harvesting 238 protein synthesis, as faster growth rates require more chlorophyll for a given light level (Geider
- et al., 1998; Inomura et al., 2020). The entrainment nitrate flux is composed of the entrainment
- 240 velocity from deepening mixed layers $\frac{dz_{ml}}{dt}$ and the nitrate gradient at the base of the mixed layer.
- 241 We find significant seasonality in entrainment velocity from 20°N and northward, with a
- 242 maximum velocity that increases with latitude (Figure 9a-b). Similar latitudinal patterns are
- found in the mean nitrate gradient at the base of the mixed layer, with the mean gradient
- increasing northward beyond the equatorial latitudes (**Figure 9c**). The seasonal cycle in the
- subsurface nitrate gradient shows a more distinct latitudinal pattern. Nitrate uptake in the spring and summar drives the gradient from 20, 40° N near zero in summar. This pattern contracts with
- and summer drives the gradient from 20-40°N near zero in summer. This pattern contrasts with
 latitudes to the north and south of 20-40°N where spring and summer nitrate depletion is less
- severe. The entrainment velocity and nitrate gradient combine to yield a seasonal cycle of
- entrainment nitrate flux with an amplitude that increases with latitude (**Figure 9d**). Wintertime
- nitrate flux exceeds 30 mmol/m²/month north of 40°N and exceeds 10 mmol/m²/month in the
- transition zone, while remaining several fold lower southward. These results demonstrate that the
- transition zone latitudes of 30-40°N is the most southern latitude to receive a significant
- wintertime nitrate entrainment flux, consistent with arguments of Glover et al., (1994).
- To ensure robustness of our results, we repeated all chlorophyll and carbon analyses above using
- estimates from the alternative GSM ocean colour inversion algorithm (Maritorena et al., 2002).
- Although low latitude GSM chlorophyll concentration magnitudes were slightly higher than
- 257 GIOP estimates and low latitudes carbon slightly higher, the seasonality in the two inversions
- was highly consistent, reproducing the correlations reported above. All results using GSM
- estimates are reported in the Supplemental Information (Figures S6-10).

260 **4 Discussion**

261 *4.1 Photoacclimation as a primary driver of wintertime transition zone chlorophyll*

The observed negative covariation of chlorophyll and carbon in the transition zone, and its relationships with light and nutrient availability, demonstrate photoacclimation as a primary driver of chlorophyll variability in the TZCF. Consistent with earlier satellite-based (Behrenfeld

et al., 2005) and physiological (Behrenfeld et al., 2016; Geider et al., 1996, 1998; Inomura et al.,

266 2020; Laws & Bannister, 1980; Talmy et al., 2013) analyses, the wintertime increase in transition

267 zone chlorophyll appears to be due to an increase in the chlorophyll to carbon ratio in response to

light-limited, nutrient-replete growth conditions. Extending this earlier work, we explicitly

resolve the latitudinal structure of photoacclimation dynamics at the basin scale and link the

- decoupling of chlorophyll and carbon seasonality to the observed seasonality of latitudinal light
- and nutrient supply.

272 This interpretation of TZCF chlorophyll dynamics extends previous studies investigating

transition zone seasonality (Ayers & Lozier, 2010; Bograd et al., 2004; Chai et al., 2003; Glover

et al., 1994; Le et al., 2019). These studies interpreted the southern extent of the wintertime chlorophyll front as a phytoplankton carbon signal and sought the necessary environmental

- chlorophyll front as a phytoplankton carbon signal and sought the necessary environmental
 drivers to explain elevated wintertime productivity. Our results, while supporting the existence of
- a significant vertical wintertime nitrate flux (Chai et al., 2003; Glover et al., 1994), suggest that
- wintertime nutrient supply has a limited impact on phytoplankton carbon and productivity.
- 279 Instead, wintertime nitrate supply relaxes nutrient stress, thus enriching the growth environment
- to provide nutrients to fully acclimatize to light-limited conditions. Laboratory observations
- show that nutrient stress decreases the chlorophyll to carbon ratio in light-lighted conditions by
- diverting nitrogen from pigments and light-harvesting proteins (Geider et al., 1996; Inomura et
- al., 2020; Laws & Bannister, 1980), supporting the suggestion that wintertime nutrient supplies
- help maximize the seasonal photoacclimation response at transition zone latitudes. Established empirical models also show a positive relationship between surface nitrate concentrations and the
- N:C of particulate biomass (Galbraith & Martiny, 2015) which further suggest that
- phytoplankton are able to build nitrogen rich pigments and light-harvesting proteins at high
- 288 nitrate supply.

289 This new perspective on seasonal chlorophyll dynamics in in the transition zone may motivate

290 re-analysis of higher trophic animal usage patterns in the region. Niche models for marine

291 mammals and seabirds have used satellite chlorophyll to represent bottom up drivers (Abrahms

et al., 2018; Block et al., 2011; Hazen et al., 2013). However, our results suggest that chlorophyll serves as a poor proxy for transition zone phytoplankton carbon and that high chlorophyll levels

- 295 serves as a poor proxy for transition zone phytopiankton carbon and that high enforophyll levels 294 indicate a relatively low carbon content of cells, either though seasonal succession of low-light
- adapted species or intra-specific acclimation. Because chlorophyll correlates strongly with sea
- surface temperature (Bograd et al., 2004), periods of decorrelation between chlorophyll and
- phytoplankton carbon, as characterized here, may provide the opportunity to isolate the roles of
- temperature and food availability on animal habitat utilization.

299 *4.2 General implications for satellite observing of marine ecosystems*

- 300 This study reiterates the need to account for photoacclimation when interpreting the satellite
- 301 chlorophyll record, echoing calls from previous authors (Behrenfeld et al., 2005, 2016; Fox et al.,
- 2020; Graff et al., 2016; Omta et al., 2009). Although the dynamics and interannual trends in
- chlorophyll have been insightful (Ayers & Lozier, 2010; Boyce et al., 2010, 2017; Glover et al.,
- 1994; Hammond et al., 2020), many studies continue to use chlorophyll as a proxy for
- 305 phytoplankton carbon and productivity. Complex relationships between chlorophyll, carbon, and

productivity, including the latitudinal dependence demonstrated here, complicate this 306

- interpretation. 307
- Beyond the North Pacific, our results suggest further work to elucidate the latitudinal structure of 308
- light and nutrient supplies globally to better understand their role in coupling chlorophyll and 309
- carbon dynamics. In the North Pacific, winter conditions conspire at the 30-40°N band to exhibit 310
- deep winter mixing that drives a large seasonal cycle in light and nutrient supply which 311
- maximizes seasonal photoacclimation; however, it is not clear whether the seasonal cycles of 312
- light and nutrient supply in other basins will show similar latitudinal structure. 313
- The relationship between mixed layer-averaged irradiance and Chl:C also has important 314
- implications for observing climate impacts on marine ecosystems (Behrenfeld et al., 2016), 315
- including multiple studies that have examined multidecadal trends in chlorophyll (Boyce et al., 316
- 2010; Hammond et al., 2020; Henson et al., 2010). Warming of the upper water column has 317
- increased stratification and reduced mixed layer depths in the North Pacific (Freeland, 2013) and 318
- across the global ocean (Li et al., 2018), with trends expected to increase into the future (Fu et 319
- al., 2016). These changes are often associated with reduced nutrient supply and productivity 320
- (Behrenfeld et al., 2006; Fu et al., 2016), which are invoked to explain chlorophyll declines. 321
- However, photoacclimation dynamics suggest that chlorophyll declines may also occur with 322
- reduced mixed layers due to photoacclimation (higher light availability with mixed layer 323 shoaling) and may be uncorrelated or negatively correlated with changes in phytoplankton
- 324
- carbon and productivity (Behrenfeld et al., 2016). 325
- Going forward, we highlight the need to further integrate eco-physiology with satellite observing 326
- of marine ecosystems to better understand the physiological growth conditions of phytoplankton 327
- at large spatial scales. Continued progress has been made with carbon-based estimates of ocean 328
- productivity that explicitly account for photoacclimation processes (Behrenfeld et al., 2005, 329
- 2016; Fox et al., 2020; Westberry et al., 2008). More recent work has further integrated resource-330
- 331 allocation strategies into remote sensing models of phytoplankton growth (Tanioka et al., 2020)
- which provide additional constraints on phytoplankton physiology by resolving cellular 332
- 333 stoichiometry as a function of light, nutrient supply, and temperature (Geider et al., 1998;
- 334 Inomura et al., 2020; Laws & Bannister, 1980).

335 **5** Conclusions

- 336 In conclusion, our results suggest chlorophyll variability in the transition zone is primarily driven
- by photoacclimation to nutrient replete, light-limited growth conditions found in the transition 337
- zone winter. The latitudinal structure of seasonal light and nutrient supply creates conditions that 338
- maximize seasonal photoacclimation and govern the TZCF chlorophyll signal. Further synthesis 339
- of phytoplankton physiology and carbon-based remote-sensing of marine ecosystems will 340
- improve our understanding of the phytoplankton growth environment and reduce uncertainties in 341
- 342 detecting climate change impacts on ocean ecosystems from space.

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- 350 available publicly through http://sites.science.oregonstate.edu/ocean.productivity/,
- 351 https://www.soda.umd.edu, http://mixedlayer.ucsd.edu/ and through the World Ocean Atlas
- 352 (Garcia et al., 2019). Details are given in the main text. All code to reproduce the analyses are
- 353 available publicly at https://github.com/gregbritten/transition_zone_chlorophyll

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487 Figure 1. Satellite-estimated climatology for surface chlorophyll concentrations, carbon

488 concentrations, and their ratio for the months of August and February in the Northeast
 489 Pacific. Panels a-c give the climatological chlorophyll, carbon, and chlorophyll:carbon ratio

490 distributions for August, respectively. Panels d-f give the same respective fields for February.









503 Figure 3. Means and standard deviations of the lag between seasonal chlorophyll and

carbon maxima. Maxima are taken within individual years according to latitudinal time series in
 Figure 2. Mean and standard deviations are calculated across years. Negative values indicate that

506 chlorophyll peaks before carbon in the seasonal cycle.



508 Figure 4. Latitudinal relationships between carbon and chlorophyll concentrations with

respect to mixed layer depth. Left column (panels a-f) gives the relationships between carbon and mixed layer depth. Right column (panels g-l) gives the relationships with chlorophyll. Rows

are latitude bands as in Figure 2. Latitude range is given in the top left of left column panels.

512 Grey lines give the ordinary least squares regression line.



513 514



516 (b), and day-averaged mixed layer-averaged irradiance (c) for ten-degree latitude bands in

517 the Northeast Pacific. Color bar gives the latitude bands for each line. Dashed line indicates the 518 transition zone (30-40°N).



522 Figure 6. Time series correlation coefficients between mixed layer depth (MLD) vs. carbon 523 concentration (a), MLD vs. chlorophyll concentration (b), MLD-averaged irradiance vs.

- concentration (a), MLD vs. chlorophyll concentration (b), MLD-averaged irradiance vs.
 carbon concentration (c), and MLD-averaged irradiance vs. chlorophyll concentration (d).
- 525 Correlations are calculated at each grid cell using the 20-year satellite time series record.



Latitude



dashed light green lines, respectively. Carbon vs. mixed layer depth (MLD) and carbon vs.

531 MLD-averaged irradiance are given in solid and dashed dark green lines, respectively. Black

532 horizontal dashed line gives the zero-correlation line.





534 Figure 8. Relationships between mixed layer-averaged irradiance and the satellite-

535 estimated chlorophyll:carbon ratio. Each box represents a ten-degree latitude band with the

- 536 limits given in the upper left of each panel. Color represents month.
- 537



Figure 9. Climatological seasonal nitrate entrainment flux across latitudes. Colors represent ten-degree latitude bands as in Figure 5. Panel a gives the climatological seasonal cycle in mixed layer depth. Panel b gives the entrainment velocity (solid black gives the zero line). Panel c gives the nitrate gradient evaluated between the mixed layer and one meter below. Panel d gives the calculated entrainment flux. Dashed lines represent the transition zone (30-40°N).

SUPPLEMENTAL INFORMATION

Seasonal photoacclimation in the North Pacific Transition Zone Gregory L. Britten¹, Christine Padalino^{1,2}, Gaël Forget¹, Michael J. Follows¹

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Figure S1. Time series of the three mixed layer estimates considered in this study. Each panel corresponds to a ten-degree latitude band (labels on the right of each panel). Argo observations are given in red, HyCOM reanalysis is in green, SODA reanalysis is in blue. We note small deviations in the amplitude across mixed layer estimates but no appreciable deviation in phasing which controls the correlations analyzed in the paper.



Figure S2. Comparison of August (left column) and February (right column) net primary productivity according to four different satellite based net primary productivity models. Model acronym is given in left column panel labels. See main text for details.



Figure S3. Time series and interannual correlations between four net primary productivity models and chlorophyll. Each pair of panels gives time series for a ten-degree latitude band (labeled in the top left). Net primary productivity models are colored. Carbon-based productivity model (CBPM) is green; vertically generalized productivity model (VGPM) is orange; carbon, assimilation, and fluorescence-resolving model (CAFÉ) is blue; vertically generalized productivity model with the Eppley temperature dependence (VGPM_EP) is in red. The chlorophyll time series corresponding to each latitude band is given in black under each net primary productivity panel. Colored numbers give the correlation coefficient between each net primary productivity model and chlorophyll.



Figure S4. As in Figure 6 of the main text but using mixed layer depth estimates from the Simple Ocean Data Assimilation (SODA) reanalysis product. See main text for details.



Figure S5. As in Figure 7 of the main text but using mixed layer depth estimates from the Simple Ocean Data Assimilation (SODA) mixed layer depth reanalysis product. See main text for details.



Figure S6. As in Figure 1 of the main text but using chlorophyll concentrations and backscatter coefficients estimated from the Garver-Siegel-Maritorena (GSM) ocean color inversion algorithm (Maritorena et al., 2002). Backscatter coefficients are converted to phytoplankton carbon estimates using the equations given in Behrenfeld et al., (2005) and Westberry et al., (2008).



Figure S7. As in Figure 2 of the main text but using chlorophyll concentrations and backscatter coefficients estimated from the Garver-Siegel-Maritorena (GSM) ocean color inversion algorithm (Maritorena et al., 2002).



Figure S8. As in Figure 3 of the main text but using chlorophyll concentrations and backscatter coefficients estimated from the Garver-Siegel-Maritorena (GSM) ocean color inversion algorithm (Maritorena et al., 2002).



Figure S9. As in Figure 4 of the main text but using chlorophyll concentrations and backscatter coefficients estimated from the Garver-Siegel-Maritorena (GSM) ocean color inversion algorithm (Maritorena et al., 2002).



Figure S10. As in Figure 8 of the main text but using chlorophyll concentrations and backscatter coefficients estimated from the Garver-Siegel-Maritorena (GSM) ocean color inversion algorithm (Maritorena et al., 2002).