# DETAILED RESPONSE TO THE REVIEWER'S COMMEMTS

We would like to thank the editor and the reviewer for re-reviewing the paper and providing the final comments. This manuscript has been improved significantly during the review process.

We have made a number of changes in response to the latest rounds of comments:

1. In response to the request for more discussion about how the models might be improved, two passages on how to improve the models have been added to the paper

# Lines 416-424 now read

In BLING and miniBLING, by contrast, the biomass responds almost instantaneously to changes in growth conditions. This means that if the growth rate increases from 0.05 day-1 to 0.1 day-1 over the course of a month, the biomass associated with large phytoplankton will increase eightfold, even though the additional growth should only be enough to give an increase of a factor of 30 days×0.05 day-1=1.5. Possibilities for addressing this effect include replacing the DOP tracer with a biomass tracer, which could then partitioned between the different phytoplankton boxes based on the temporally smoothed growth rate, or increasing the timescale over which the growth rate is smoothed when biomass is calculated.

At lines 429-430 we state that the details of light limitation representation will be addressed in future research.

Lines 638-643

Overflows are difficult to simulate in level-coordinate models because they are prone to excessive entrainment of the dense plume (Winton et al., 1998). While significant effort has gone into simulating the Denmark Straits overflow at coarse resolution (Legg et al., 2009), our results show that smaller overflows such as the Persian Gulf may be regionally significant. This may provide further impetus for using isopycnal models in high resolution simulations, as such models can potentially simulate such overflows with greater fidelity.

2. All required technical corrections have been made.

3. The manuscript has been re-reviewed thoroughly by the authors and minor changes and spelling corrections have been made. For more clarity on the corrections, a corrected version of the manuscript that shows the corrections is accompanying this letter.

4. The quality of almost all figures has been enhance via using a post-script format.

Sincerely

Safoura Seddigh Marvasti and Anand Gnanadesikan

# **Response to reviewer 2**

I would like to express our appreciation to the Referee #2 for the comments.

Below, I respond to the comments, providing more elaboration or changing the manuscript if applicable.

1) The explanation given in the manuscript that "this is probably associated with the models mixing to excessive depth during the wintertime" (page 14, line 408/409) is very dissatisfactory. It may not be feasible to include a deeper analysis in this manuscript, but the authors could at least provide a discussion how the origin of the asymmetry in the observed nutrient dynamics may be investigated in further studies.

Response : The origins and also possibilities to address the issue are added to the manuscript as follows,

On page 14:

• • • •

The equilibrium assumption, which means that biomass in both BLING and miniBLING is not directly simulated. In TOPAZ, the growth of plankton during the spring is limited by the biomass of phytoplankton, whereas in the fall TOPAZ continues to have higher heterotrophic biomass (diagnosed from growth rates over previous months) that then grazes the plankton. In BLING and miniBLING, by contrast, the biomass responds almost instantaneously to changes in growth conditions. This means that if the growth rate increases from 0.05 day<sup>-1</sup> to 0.1 day<sup>-1</sup> over the course of a month, the biomass associated with large phytoplankton will increase eightfold, even though the additional growth should only be enough to give an increase of a factor of 30 days×0.05 day-1=1.5. Possibilities for addressing this effect include replacing the DOP tracer with a biomass tracer, which could then partitioned between the different phytoplankton boxes based on the temporally smoothed growth rate, or increasing the timescale over which the growth rate is smoothed when biomass is calculated.

And also on page 21

Accordingly, there is a potential to improve the numerical models by better simulating the Persian Gulf Outflow to produce a sharper thermocline, allowing more realistic nutrient supply. Overflows are difficult to simulate in level-coordinate models because they are prone to excessive entrainment of the dense plume (Winton et al., 1998). While significant effort has gone into simulating the Denmark Straits overflow at coarse resolution (Legg et al., 2009), our results show that smaller overflows such as the Persian Gulf may be regionally significant. This may provide further impetus for using isopycnal models in high resolution simulations, as such models can potentially simulate such overflows with greater fidelity.

2) Technical corrections:

p. 4, l. 114: Resplandy et al. (2011) indicated that the spatial  $\rightarrow$  Resplandy et al. (2011) indicated that the spatial  $\rightarrow$  Done (please see the manuscript).

p. 4, l. 122: with the specification four -> with the specification of four $\rightarrow$ Done (please see the manuscript).

p. 10, l. 286: macronutrient tracer Behrenfeld (2010). -> macronutrient tracer (Behrenfeld, 2010). → Done (please see the manuscript).

p. 10, l. 307: Within the smalle region the values are  $\rightarrow$  Within the small region, the values are  $\rightarrow$  Done (please see the manuscript).

p. 18, l. 530: much water from Persian Gulf  $\rightarrow$  much water from the Persian Gulf  $\rightarrow$  Done (please see the manuscript).

p. 20, l. 590: blooms in during certain times  $\rightarrow$  blooms during certain times  $\rightarrow$  Done (please see the manuscript).

p. 20, l. 617/618: However, for a few cases eddies with a bloom -> However, for a few cases, eddies with a bloom  $\rightarrow$  Done (please see the manuscript).

p. 20, l. 618/619: In this region consistency is observed -> In this region, consistency is observed  $\rightarrow$  Done (please see the manuscript).

- 1 Challenges in modelling spatiotemporally varying phytoplankton blooms in the
- 2 Northwestern Arabian Sea and Gulf of Oman
- 3

# 4 S. Sedigh Marvasti<sup>1</sup>, A. Gnanadesikan<sup>2</sup>, A.A. Bidokhti<sup>3</sup>, J.P. Dunne<sup>4</sup>, S. Ghader<sup>5</sup>

5 [1] Department of Marine Sciences, Science and Research Branch, Islamic Azad University,

6 Tehran, Iran. (<u>safoora.seddigh@gmail.com</u>)

- [2] Department of Earth and Planetary Sciences, Johns Hopkins University, Olin Hall, 3400
  N. Charles St., Baltimore, MD 21218, USA. (gnanades@jhu.edu)
- 9 [3] Institute of Geophysics, University of Tehran, Tehran, P. O. Box 14155-6466, Iran.
  10 (bidokhti@ut.ac.ir)
- [4] National Oceanic and Atmospheric Administration/Geophysical Fluid Dynamics
   Laboratory, 201 Forrestal Rd., Princeton, NJ 08540-6649. (john.dunne@noaa.gov)
- 13 [5] Institute of Geophysics, University of Tehran, Tehran, P. O. Box 14155-6466, Iran.
- 14 (sghader@ut.ac.ir)
- 15 Correspondence to: S. Sedigh Marvasti (safoora.seddigh@gmail.com)
- 16

#### 17 Abstract

18 Recent years have shown an increase in harmful algal blooms in the Northwest Arabian Sea 19 and Gulf of Oman, raising the question of whether climate change will accelerate this trend. 20 This has led us to examine whether the Earth System Models used to simulate phytoplankton 21 productivity accurately capture bloom dynamics in this region- both in terms of the annual 22 cycle and interannual variability. Satellite data (SeaWIFS ocean color) shows two 23 climatological blooms in this region, a wintertime bloom peaking in February and a 24 summertime bloom peaking in September. On a regional scale, interannual variability of the 25 wintertime bloom is dominated by cyclonic eddies which vary in location from one year to 26 another. Two coarse (1°) models with the relatively complex biogeochemistry (TOPAZ) 27 capture the annual cycle but neither eddies nor the interannual variability. An eddy-resolving 28 model (GFDL CM2.6) with a simpler biogeochemistry (miniBLING) displays larger 29 interannual variability, but overestimates the wintertime bloom and captures eddy-bloom

30 coupling in the south but not in the north. The models fail to capture both the magnitude of 31 the wintertime bloom and its modulation by eddies in part because of their failure to capture the observed sharp thermocline/nutricline in this region. When CM2.6 is able to capture such 32 33 features in the Southern part of the basin, eddies modulate diffusive nutrient supply to the surface (a mechanism not previously emphasized in the literature). For the model to simulate 34 35 the observed wintertime blooms within cyclones, it will be necessary to represent this 36 relatively unusual nutrient structure as well as the cyclonic eddies. This is a challenge in the 37 Northern Arabian Sea as it requires capturing the details of the outflow from the Persian Gulf-38 something that is poorly done in global models.

39

#### 40 **1** Introduction

41 The region of northwestern Arabian Sea and the Gulf of Oman  $(15^{\circ}-26^{\circ} \text{ N}, 56^{\circ}-66^{\circ}\text{E})$ 42 is a highly productive region (Madhupratap et al., 1996; Tang et al., 2002), with satellite 43 estimates of carbon export of 137 gC/m<sup>2</sup>/yr, much higher than the ~80 gC/m<sup>2</sup>/yr found in the 44 Subpolar North Atlantic and Pacific (Dunne et al., 2007). Peak chlorophyll-a concentrations 45 exceed 0.7 mg/m<sup>3</sup> in this region (Fig. 1a).

This region may be changing in important ways. In both the Persian Gulf and the Gulf 46 47 of Oman, there is evidence that harmful algal bloom (HABs) and their impacts are increasing 48 (Richlen et al., 2010). HAB occurrences have been more frequently reported in the Gulf of 49 Oman than in the Persian Gulf. A total of 66 red tide events (mostly dominated by Noctiluca 50 scintillans) have been recorded between 1976 and 2004 including 25 blooms resulting in mass mortality of fish and marine organisms. Reasons for the increase in blooms include 51 52 aquaculture, industrial and sewage inputs, natural dispersal and human-aided transport, long-53 term increases in nutrient loading and global expansion of species (Richlen et al., 2010) as 54 well as global climate change (Goes et al., 2005). The latter paper suggested that increasing 55 blooms were driven by an increase in the strength of the Asian monsoon.

Evaluating such a possibility and extending it into the future requires the use of Earth
System Models. However, such projections will only be as good as the models on which they
are based. In this paper we examine several models run at the Geophysical Fluid Dynamics
Laboratory in the Arabian Sea. We consider numerical results from five, different 3D global
Earth system models, which we denote CORE-TOPAZ, Coupled-TOPAZ, Coupled-BLING,
Coupled-miniBLING, and the Geophysical Fluid Dynamics Laboratory Climate Model

Anand 1/7/2016 10:04 AM

Deleted: rather

Anand 1/7/2016 10:05 AM Deleted: e activities Anand 1/7/2016 10:05 AM Deleted: and

Anand 1/7/2016 10:06 AM Deleted: our

66 version 2.6 (CM2.6 miniBLING). The first two of these models use the relatively complex

67 TOPAZ biogeochemistry, but have low resolution and do not resolve eddies, the third has a

simplified biogeochemistry (BLING, Galbraith et al., 2010) which does not carry
phytoplankton biomass as a separate variable while the last two models have an even, simpler
biogeochemistry that does not directly simulate dissolved organic matter or iron cycling
(miniBLING, Galbraith et al., 2015b). Only the final model resolves eddies.

72 The seasonal cycle is an important metric for models to be able to simulate. The Arabian Sea is influenced by a reversing monsoonal cycle (Wang and Zhao, 2008), an 73 74 evaporative fresh-water flux over most of the basin, and an annual mean heat gain (Banse and McClain, 1986; Fischer et al., 2002). In summer (June-September), the southwest Monsoon 75 76 (SWM) blows strongly across the northwestern Arabian Sea (Al-Azri et al., 2010). Driven by 77 a land-sea pressure gradient, the SWM is a large-scale feature of the atmospheric circulation of the tropics, extending from a surface pressure high near  $30^{\circ}$  S in the southern hemisphere 78 79 northward to the surface low over Asia (Anderson and Prell, 1993). During the SWM, winds 80 are steered by the East African highlands to form a strong low level atmospheric jet, referred 81 to as the Findlater Jet (Bartolacci and Luther, 1999; Honjo et al., 2000), which crosses the 82 Equator over the Indian Ocean and blows over the Arabian Sea parallel to the Omani 83 coastline in a northeast direction (Honjo et al., 2000). The orientation of the Findlater Jet parallel to Omani coast leads to coastal upwelling along the coast and downwelling on the 84 eastern side of the Jet in the middle of Arabian Sea. This upwelling provides nutrients to the 85 surface layer (Fig. 1b) (Al-Azri et al., 2013; Kawamiya and Oschlies, 2003; Madhupratap et 86 87 al., 1996; Murtugudde et al., 2007; Veldhuis et al., 1997; Wang and Zhao, 2008). The SWM does not destabilize the surface layers, which are fairly stable in northern summer (Fig. 1c). 88

89 The Northeast Monsoon (NEM), which happens from December through February, is 90 not as strong as the SWM (Dickey et al., 1998; Shalapyonok et al., 2001; Veldhuis et al., 1997). Ocean surface wind stress is lower (0.032 N/m<sup>2</sup> in NEM compared to 0.127 N/m<sup>2</sup> in 91 92 SWM), and does not lead to upwelling like the SWM along the Omani coast. However, 93 negative heat flux results in a destabilizing buoyancy flux, subsequent convective overturning (Barimalala et al., 2013; Kawamiya and Oschlies, 2003), and deepening and cooling to a 94 depth of ~60m (Fig. 1c, 1d). This brings up nutrients and fuels a wintertime bloom. In 95 addition, as shown in Fig. 1d in wintertime bloom the mixed layer depth (MLD) is deeper 96 97 than summer.

Anand 1/7/2016 10:06 AM Deleted: ) ( Anand 1/7/2016 10:07 AM Deleted: as very high resolution, but also a 1/10/2016 10:44 AM Shahab Deleted: ( Shahabeddin Tora..., 1/10/2016 10:45 AM Deleted: ) 1/10/2016 10:44 AM Shahabeddin Tora. Deleted: Galbraith et al., 2015 Anand 1/7/2016 10:07 AM Deleted: Biological cycling in CM2.6 is simulated using a modified version of the Biogeochemistry with Light Iron Nutrients and Gasses (BLING) model (Galbraith et al., 2010) called mini-BLING.



108 A second metric of the bloom dynamics is the relationship between the blooms and 109 mesoscale eddies (Al-Azri et al., 2013; Dickey et al., 1998; Hamzehei and Bidokhti, 2013; Shalapyonok et al., 2001; Gomes et al., 2005). The confluence of the Persian Gulf outflow 110 111 current and the East Arabian Sea Current parallel to Omani and Yemeni coastlines in Arabian 112 Sea leads to formation of a frontal zone and formation of persistent eddies in the region. 113 Because the size of eddies is comparable to the width of the Gulf of Oman, they can affect 114 mixing and transport of biota on a basin scale (Fischer et al., 2002; Piontkovski et al., 2012). 115 Piontkovski et al. (2012) suggested that the increased amplitude of the seasonal cycle of 116 chlorophyll-a might be associated with the increased variability of mesoscale eddy kinetic 117 energy (EKE) per unit mass in the Gulf of Oman or in the western Arabian Sea. Gomes et al. (2008) noted potential anticorrelation between sea surface height and chlorophyll, but did not 118 119 find a consistent relationship over time.

120 Gaube et al. (2014) provide a global overview of how eddies influence chlorophyll 121 blooms. They find that the effect of mesoscale eddies on the chlorophyll bloom varies both 122 temporally and spatially. They identify four particular mechanisms that can be distinguished 123 by linking sea surface anomalies to chlorophyll, namely eddy stirring, trapping, eddy 124 intensification, and Ekman pumping. Although Gaube et al. (2014) find a negative correlation 125 between chlorophyll and SSH in the Arabian Sea, they do not analyse which of these mechanisms is involved in this region, nor do they quantify the extent to which this 126 127 correlation varies over the course of the season.

Resplandy et al. (2011) indicated that the spatial variability associated with mesoscale eddies in the Arabian Sea produces spatial variability in the bloom and that another source of variability is found to be restratification at these structures. Advection from coastal region is identified as the mechanism providing nutrients in summer, while vertical velocities associated with mesoscale structure are found to increase the overall nutrient supply. However, this work does not make clear how the spatial distribution of the eddy\_nutrient supply is related to the eddies, not whether this relationship is the same in all seasons.

The structure of this paper is as follows: all datasets including ocean color data and altimeter data are explained in section 2 of the paper along with the specification of five, different 3D global Earth system models. In section 3, the remote sensing results are used to study the spatiotemporal variability of chlorophyll-a in mesoscale structures in the study region. We find a seasonal relationship between SSHA and chlorophyll such that cyclonic

Anand 1/7/2016 10:12 AM Deleted: our

141 eddies are associated with blooms, but only during the winter. This means that interannual 142 variability in blooms will be shaped by mesoscale eddy activity and may not be predictable. 143 Results of the 3D global Earth system models are discussed in section 4. Annual cycles of 144 variation of chlorophyll-a and nutrients for all GFDL models within the whole region are 145 compared against the corresponding satellite results and field measurements. The models tend 146 to overestimate wintertime productivity, in large part due to excessive mixing. They also fail 147 to explain the bloom-SSHA relationship except in a few special cases. We argue that the 148 eddies act to modulate turbulent mixing of nutrients to the surface- a mechanism not 149 emphasized in previous literature. However, this can only occur if there is a strong and 150 relatively shallow nutricline. Since the model only simulates such a feature in the Southern 151 Arabian Sea, it does not capture the observed relationship between SSH and biology. Both the 152 overestimation of the wintertime bloom and the failure to predict its modulation by eddies can 153 thus be traced to difficulties in modeling the stratification of the Northwest Arabian Sea, most 154 likely as a result of a failure to properly simulate overflows.

155

# 156 2 Description of data and models

#### 157 2.1 Satellite products

We examine the relationship of blooms and eddies using the GSM5 Maritorena et al. (2002) product based on the SeaWIFS (Sea-viewing Wide Field-of-view Sensor) ocean color data and Sea Surface Height Anomaly (SSHA), based on altimeter data acquired from the Archiving, Validation and Interpretation of Satellite Oceanographic (AVISO) Data Center (http://www.aviso.oceanobs.com). The SSH anomaly is calculated relative to the annual cycle.

164 The GSM algorithm represents the normalized water leaving radiance  $L_{wN}(\lambda)$  at multiple 165 wavelengths as a nonlinear function, as following (Maritorena et al., 2002),

$$\hat{L}_{wN}(\lambda) = \frac{iF_0(\lambda)}{n_w^2} \sum_{i=1}^2 g_i \left\{ \frac{b_{bw}(\lambda) + b_{bp}(\lambda_0)(\lambda/\lambda_0)^{-\eta}}{b_{bw}(\lambda) + b_{bp}(\lambda_0)(\lambda/\lambda_0)^{-\eta} + a_w(\lambda) + Chl a_{ph}^*(\lambda) + a_{cdm}(\lambda_0) exp[-S(\lambda - \lambda_0)]} \right\}^{l}$$
(1)

where *t* is the sea-air transmission factor,  $F_{\theta}(\lambda)$  is the extraterrestrial solar irradiance,  $n_w$  is the index of refraction of the water, seawater backscatter  $b_{bw}(\lambda)$ , absorption  $a_w(\lambda)$ ,  $a_{ph}^*$  is the chlorophyll-a (*chl*) specific absorption coefficient, *S* is the spectral decay constant for

170 absorption by chromophoric dissolved organic materials (CDOM),  $\eta$  is the power-law 171 exponent for the particulate backscattering coefficient, and  $\lambda_0$  is a scaling wavelength (443 nm). The cdm absorption coefficient  $[a_{cdm}(\lambda_0)]$ , and slope factor S then determine the 172 173 absorption across a range of wavelengths while the particulate backscatter coefficient  $[b_{bp}(\lambda_0)]$ 174 and coefficient  $\eta$  constrain the scattering. Letting  $\lambda_{\theta}$  be 443 nm assuming that all terms other 175 than chl,  $[a_{cdm}(\lambda_0)]$  and  $b_{bp}(443nm)$  are constant, one can then use the <u>normalized</u> water 176 leaving radiance to invert for chl, a<sub>cdm</sub>, and backscatter <u>b<sub>bp</sub></u>. One limitation of this approach is 177 that if the inherent optical properties vary with time or space, this variation will introduce 178 errors into the estimate. Following Behrenfeld et al. (2005), we convert the backscatter 179 coefficient into units of particulate carbon biomass using the relationship  $p_{carb}=13000(b_{bp}-$ 180 0.00035).

181 Satellite-based remote sensing is the only observational method suitable for measuring 182 physical and biological properties over large regions of the ocean. However, satellite ocean 183 color and SST are limited to surface distributions and provide no information about the 184 vertical structure within the ocean (McGillicuddy et al., 2001). Additionally acquiring data 185 requires cloud-free viewing of the ocean surface, which as we will see is a problem in this 186 region at certain times of the year. This lack of information motivates our examination of 187 numerical models, which ideally could be used to provide estimates of the ocean state when 188 observations are sparse as well as to extrapolate both vertically and into the future.

189

#### 190 2.2 Numerical models

191 Numerical results are presented in this paper based on the output of five different 3D 192 global Earth system models, which we denote CORE-TOPAZ, Coupled-TOPAZ, Coupled-193 BLING/miniBLING and GFDL CM2.6 (miniBLING). The first two of these models use the 194 relatively complex TOPAZ biogeochemistry, but have low resolution and do not resolve 195 eddies. The third and fourth use, two simplified biogeochemistry codes (BLING and 196 miniBLING) which do not carry phytoplankton biomass as a separate variable while the last 197 model has very high resolution and uses the miniBLING simplified biogeochemistry. Below, 198 we describe the different physical models, followed by a summary of the biogeochemical 199 codes run within these models.

200

Anand 1/7/2016 10:22 AM Deleted: our

Anand 1/7/2016 10:23 AM Deleted: s

### 203 <u>2.2.1 Physical model description</u>

204 The ocean-ice model used in the CORE-TOPAZ model follows the corresponding 205 components of the GFDL CM2.1 global coupled climate model (Delworth et al., 2006). The 206 vertical resolution ranges from 10 m over the top 200 m to a maximum thickness of 250 m at 207 5500 m depth with 50 layers in all. The meridional resolution is 1°, whereas the zonal resolution varies between  $1^{\circ}$  in mid-latitudes and  $\frac{1}{3}^{\circ}$  at the equator. North of 65°, a tripolar 208 209 grid is employed to avoid singularity arising from convergence of meridians at the North 210 Pole. Up-to-date parameterizations of mixed-layer dynamics, isopycnal mixing, advection by 211 subgridscale eddies, bottom topography, bottom flows, and lateral viscosity are included- for 212 more detail see Griffies et al. (2005) and Gnanadesikan et al. (2006). Both the dynamics and 213 thermodynamics sea ice are simulated with of five thickness classes of sea ice being resolved.

214 In the CORE-TOPAZ model, surface forcing is set using the Coordinated Ocean-ice 215 Reference Experiment (CORE) protocol (Griffies et al., 2009), where the inputs for 216 calculating surface fluxes are taken from an atmospheric analysis dataset adjusted to agree 217 better with in situ measurements. Sensible and latent heat fluxes are then calculated using 218 bulk formulae. Freshwater forcing is given by a combination of applied precipitation, 219 evaporation computed using bulk fluxes, and a correction diagnosed to restore surface 220 salinities in the top 10 m to climatological monthly values over 60 d. Hence, the fluxes 221 forcing the CORE runs could be thought of as "best guess" observationally based estimates. 222 Such a prescription omits important feedbacks whereby the atmosphere ensures that rainfall 223 and evaporation are consistent with each other, although the restoring correction is a crude 224 representation of these feedbacks. We use the version of the model described in Gnanadesikan 225 et al. (2011), which analyzed different modes of interannual variability in biological cycling 226 across the North Pacific Ocean.

227 The Coupled-TOPAZ model corresponds to the control simulation of the GFDL 228 ESM2M submitted as part of the IPCC AR5 process (Dunne et al., 2012). In this model the 229 ocean is coupled to the atmosphere, land, and sea ice components. Gnanadesikan et al. (2014) 230 discuss the behavior of this model in the North Atlantic, but its behavior in the Arabian Sea 231 has not been previously analyzed. Two additional versions of this model, referred to here as 232 Coupled-BLING/miniBLING, were run using the BLING and mini-BLING biogeochemical 233 models described below, but with the light field given by the TOPAZ code. The differences 234 between the 1-degree models highlight differences due to biological formulation.

Anand 1/7/2016 10:25 AM Deleted: as Anand 1/7/2016 10:25 AM Deleted: emphasizes

237 The ocean component of ESM2M employs the MOM4p1 code of Griffies et al. (2009) 238 which largely mimics the CM2.1 ocean (identical horizontal and vertical resolution and parameterization of mixing). However, ESM2M ocean uses a rescaled geopotential vertical 239 240 coordinate ( $z^*$ ; Adcroft et al., 2004; Stacey et al., 1995) for a more robust treatment of free 241 surface undulations. The ESM2M implementation includes updates to the K-profile 242 parameterization (Large et al., 1994) based on Danabasoglu et al. (2006), as well as model-243 predicted chlorophyll modulation of short-wave radiation penetration through the water 244 column. ESM2M also includes completely novel parameterizations relative to CM2.1, such as 245 parameterization of submesoscale eddy-induced mixed layer restratification (Fox-Kemper et al., 2008). Instead of prescribed vertical diffusivity for interior mixing (Bryan and Lewis, 246 247 1979), ESM2M employs the Simmons et al. (2004) scheme along with a background diffusivity of  $1.0 \times 10^{-5}$  m<sup>2</sup> s<sup>-1</sup> in the tropics and  $1.5 \times 10^{-5}$  m<sup>2</sup> s<sup>-1</sup> poleward of 30° latitude 248 249 following a tanh curve.

The Geophysical Fluid Dynamics Laboratory Climate Model version 2.6 (CM2.6) is a high-resolution eddy-resolving model. This model has the same atmosphere model and ocean Physics as CM2.5 (Delworth et al., 2012). CM2.6's ocean component has higher horizontal resolution than CM2.5, with grid spacing, which is changeable from 11 km at the equator to less than 4 km at very high latitudes. This means that the model is capable of resolving eddy features in the tropics, as we will see below.

256

### 257 2.2.2 Biogeochemical Cycling codes

258 The TOPAZ code (Tracers of Ocean Productivity with Allometric Zooplankton code 259 of Dunne et al., 2010), keeps track of five inorganic nutrients used by phytoplankton: nitrate and ammonia, inorganic phosphate, silicate, and dissolved iron. Additionally, the model 260 261 carries three other dissolved inorganic tracers: dissolved inorganic carbon, alkalinity and 262 dissolved oxygen. Based on the work of Dunne et al. (2007), the model also keeps track of 263 fine lithogenic material, which plays a role in ballasting organic material and delivering it to 264 the sediment (Armstrong et al., 2002; Klaas and Archer, 2002). The five inorganic nutrients 265 are taken up in different ways by three classes of phytoplankton: small, large and diazotrophic. A comprehensive description of TOPAZ v2 can be found in the supplemental 266 267 material of Dunne et al. (2013).

Shahabeddin Tora..., 1/10/2016 11:02 AM Formatted: Font:(Default) Times New Roman

Shahabeddin Tora..., 1/10/2016 11:02 AM Formatted: Font:(Default) Times New Roman

Shahabeddin Tora..., 1/10/2016 11:02 A Formatted: Font:(Default) Times New Roman

268 TOPAZ is unusual among comprehensive Earth System Models in that it uses a highly 269 parameterized version of grazing. Instead of grazers being explicitly simulated, grazing rates are simply taken as a function of phytoplankton biomass, with different power-law 270 271 dependence for small and large phytoplankton. The grazing formulation was fit to about 40 272 field sites to produce a size structure that transitions realistically between being dominated by 273 small phytoplankton and low particle export ratio at low levels of growth and large 274 phytoplankton and high particle export ratio in nutrient and light-replete conditions. At 275 equilibrium, the resulting parameterization produces biomass that is a function of growth rate 276 (linear for small plankton, cubic for large). A similar scaling in particle size spectrum was 277 seen across ecosystems by Kostadinov et al. (2009). In contrast to models that explicitly 278 simulate zooplankton, TOPAZ, does not depend on poorly known zooplankton behavioral 279 parameters (such as handling efficiency or grazing half-saturation) or, on the details of how 280 different trophic levels interact.

281 Even though it does not simulate zooplankton explicitly, TOPAZ still carries over two 282 dozen tracers, making it extremely expensive to run in high-resolution simulations. For this reason Galbraith et al. (2010) developed the Biogeochemistry with Light Iron Nutrients and 283 284 Gasses (BLING) model, which parameterizes the entire ecosystem. The original version of 285 BLING has only five explicit tracers: dissolved inorganic phosphorus ( $PO_4$ ), dissolved organic phosphorus (DOP), dissolved Iron (Fe), dissolved inorganic carbon (DIC), and 286 287 oxygen (O<sub>2</sub>). It includes the impacts of macronutrient and micronutrient limitation and light 288 limitation on phytoplankton by using these to calculate a growth rate. Using the same 289 machinery as TOPAZ, it then uses this growth rate and implicit treatment of community 290 structure to estimate phytoplankton biomass, and uses this biomass to calculate the rate at 291 which nutrient is taken up by plankton and cycled through the ecosystem.

292 The miniBLING code (Galbraith et al., 2015) represents a further simplification. In 293 this model the iron field is taken from a lower-resolution version of the model (an 294 approximation which has limited impact in the Arabian Sea, where phytoplankton are 295 generally not iron-limited) and so Fe is not treated prognostically. Additionally the DOP pool 296 is eliminated. Simulations using the ESM2M physical model show that control simulations of oxygen and surface nutrients produced by the miniBLING and BLING models are very 297 298 similar to those produced in the same model with TOPAZ (Galbraith et al., 2015). Galbraith 299 et al. (2015), also show, that BLING and miniBLING simulate very similar patterns of oxygen Anand 1/7/2016 10:27 AM **Deleted:** it Anand 1/7/2016 10:27 AM **Deleted:** n to

Anand 1/7/2016 10:28 AM Deleted: (Galbraith et al., 2010) Shahabeddin Torab..., 1/10/2016 9:57 AM Formatted: Subscript Anand 1/7/2016 10:29 AM Deleted: DIC (

Anand 1/7/2016 10:30 AM Deleted: This manuscript Anand 1/7/2016 10:30 AM Deleted: s

306 change and anthropogenic uptake in a simulation where CO2 is increased by 1% per year307 until it is twice the preindustrial concentration,

308 It should be noted that simplified BLING and miniBLING codes neglect some 309 processes that may be important. Only nonliving components are advected and mixed by the 310 ocean circulation, which could result in inaccurate distribution of biology in frontal regions at 311 high resolution. Additionally, as will be discussed below, the lack of a biomass variable may 312 lead to overestimating how rapidly plankton inventories can grow. Also, the rich behavior of 313 the nitrogen cycle with its interaction with iron, phosphorus and oxygen cannot be simulated 314 with one macronutrient tracer (Behrenfeld 2010). Specifying iron limitation, as done in 315 miniBLING, may also have some impacts in our region. As extensively discussed by Naqvi et 316 al., (2010) there is a possibility of iron limitation over the southern parts of the Omani shelf 317 and in the offshore region during the latter part of the Southwest Mansoon, which can result 318 in high nitrate-low chlorophyll conditions. The western equatorial and southern tropical 319 region of the Indian Ocean are iron-limited and the Arabian Sea (southern parts) may become 320 iron-limited under strong upwelling conditions (Wiggert et al., 2006).

321

## 322 3 Remote sensing results

#### 323 **3.1** Annual cycle and interannual variability

We begin by using the GSM5 satellite data to examine the annual cycle and interannual variability in two different regions, the whole NW Arabian Sea (56°-66° E, 15°-26° N) and a smaller region including the Gulf of Oman, (60°-62° E, 22°-26° N). As shown in Fig. 2 (a) to (c) for whole region, clear annual cycles of chlorophyll-a, backscattering and CDOM are observed. Even larger annual cycles of variation of chlorophyll-a, backscattering and CDOM are seen in the smaller region, as shown in Fig. 2 (d) to (f). More pronounced interannual variability is observed in the smaller region as opposed to the larger region.

The annual variations of all parameters are broadly consistent with each other. The maximum values associated with the summer bloom are generally seen in September, with values of 1.0 mg/m<sup>3</sup>, 50 mgC/m<sup>3</sup>, and 0.1 m<sup>-1</sup> for chlorophyll, particulate carbon and CDOM, respectively, within the whole region. Within the small<u>er</u> region, the values are 1.25 mg/m<sup>3</sup>, 65 mgC/m<sup>3</sup>, and 0.125 m<sup>-1</sup> for chlorophyll, particulate carbon, and CDOM, respectively. For two years of 2001 and 2002, the particulate carbon values (~90 mgC/m<sup>3</sup>) are much higher that the average

#### Anand 1/7/2016 10:30 AM Deleted: doubles

Shahabeddin Torabian 1/6/2016 8:59 PM Deleted: (

Anand 1/7/2016 10:33 AM Deleted: I

Anand 1/7/2016 10:34 AM **Deleted:** e

- 341 of the other months over both regions, but the chlorophyll does not show pronounced peaks.
- 342 A winter bloom is also pronounced in February as a second maximum in a yearly cycle,
- 343 where the magnitudes are about 0.07 mg/m<sup>3</sup>, 40 mgC/m<sup>3</sup>, and 0.07 m<sup>-1</sup> for chlorophyll,
- 344 particulate carbon and CDOM, respectively, within the whole region, and about 0.09~1.5
- $345 \mid mg/m^3$ ,  $55 \sim 80 mgC/m^3$ , and  $0.11 \sim 0.14 m^{-1}$  for chlorophyll, particulate carbon and CDOM,

346 respectively, within the smaller region. That the summer bloom in the both regions is stronger

than the winter bloom has been discussed by Al-Azri et al. (2010), and Levy et al. (2007).

#### 348 3.2 Variability of Chlorophyll-a in Mesoscale Structures

349 Mesoscale structures can be seen in the Northwest Arabian Sea in both the SeaWiFS 350 chlorophyll-a distribution and AVISO sea surface height anomaly. Over the course of 2001 351 (Fig. 3), both a summer bloom (which most likely starts in August and ends in ~October) and 352 a winter bloom (which starts in January and goes away in April) can be seen in chlorophyll-a. 353 In March, the last month of the winter bloom, chlorophyll-a concentrations are high over the 354 entire region in both the anticyclones (warm eddies with positive SSHA) and the cyclones 355 (cold eddies with negative SSHA). The observed bloom in March terminates abruptly in 356 April, although the observations show that eddies are still active in the region. In June, July 357 and August, the satellite ocean color data is not available due to excessive cloudiness. In 358 September, the last month of the summer bloom, most of the region including cyclones and 359 anticyclones and coastal regions had high chlorophyll-a concentration. However in the 360 following months the bloom persists only within the cold eddies and disappears over the 361 warm eddies (a phenomenon also seen in Sargasso Sea by McGillicuddy et al., 2001). The 362 relationship between sea surface chlorophyll-a and eddies for the other years between 1998 363 and 2005 during the month of November, is shown in Fig. 4. The relationship between blooms and SSHA is clear and striking. Note particularly the difference between 1998 and 364 365 2001, when the location of high and low chlorophyll regions switches relative to the Ras al 366 Hadd. This difference in bloom location is perfectly reflected in the different locations of the 367 eddies.

## 368 3.3 Chlorophyll-Sea Surface Height Anomaly (SSHA) cross-correlation

The seasonal relationship between chlorophyll and SSHA can be seen in <u>the</u> monthly variation of the spatial cross-correlation between the two variables over the entire northwest

371 Arabian Sea. chlorophyll-SSHA cross-correlations between 1998 and 2005 in the satellite

Anand 1/7/2016 10:35 AM Deleted: mgC/m<sup>3</sup>

Anand 1/7/2016 10:36 AM **Deleted:** mgC/m<sup>3</sup>

Anand 1/7/2016 10:37 AM **Deleted:** patterns

375 data are shown in Fig. 5a. To check that the chlorophyll results are not an artifact of the 376 remote sensing inversion, two other related parameters, the backscattering coefficient (BBP) and chromophoric dissolved organic matter (CDOM) are also cross-correlated with SSHA, as 377 378 depicted in Figs. 5b and c. The results show consistent annual cycles of variation in the cross-379 correlation of all three variables. This suggests a repeatable yearly phenomenon in the region 380 as discussed in the previous sections. The cross-correlation results over the 8 years of study 381 show, that there are several months (i.e. November-December) with relatively high anti-382 correlation for most of the years and also several other months (i.e. April-May) with no or 383

even low positive correlation.

384 The averaged climatological monthly cross-correlation with SSHA and climatological 385 monthly values between 1998 and 2005 are shown in Fig. 6 for all parameters. Two blooms 386 ending in March (winter) and September (summer) are seen. At the peak of the blooms the 387 average cross-correlation values are very low due to the existence of blooms in both cyclones 388 and anticyclones. The months after the winter and summer blooms show a clear difference in 389 the correlation. After the winter bloom (typically April and May), the cross-correlation is 390 positive or very small, which suggests no relation between the mesoscale eddies and the 391 blooms. As discussed in Kumar et al. (2001), low primary production is observed after 392 termination of winter cooling during Spring Inter-Mansoon (SIM) (see also Gomes et al., 393 2008). This result would be also consistent with SIM producing weak atmospheric forcing in 394 the region.

395 In contrast, after the summer bloom (typically October-December) as the average values of 396 chlorophyll-a decrease, chlorophyll and SSHA become relatively highly anti-correlated. The 397 reason for the anti-correlation is the persistence of chlorophyll at the regions with negative 398 SSHA that are typically considered to be cyclonic (cold) eddies and disappearance of 399 chlorophyll-a in positive SSHA that are assumed to be anti-cyclonic (warm) eddies. Particle 400 backscatter also provides almost same cross-correlation, suggesting that the chlorophyll-a 401 signal does not result purely from photo-adaptation. Moreover, the CDOM-SSHA cross-402 correlation shows the same annual cycle although with smaller peak values,

403 The spatial relationship between blooms and eddies seen in the Northern Arabian Sea can be 404 compared with the patterns noted by Gaube et al. (2014). Their eddy stirring mechanism 405 involves advection of high and low chlorophyll signals around an eddy, resulting in a low 406 which is offset from the center of an anticyclone and a high which is offset from the center of

Anand 1/7/2016 10:38 AM Deleted: s

Shahabeddin Torab..., 1/12/2016 8:17 PM Deleted:

Anand 1/7/2016 10:40 AM

12

Deleted: and average value results

Anand 1/7/2016 10:41 AM Deleted: patterns for bloom tim Anand 1/7/2016 10:41 AM Deleted: es but typically shows Anand 1/7/2016 10:42 AM Deleted: lower Anand 1/7/2016 10:42 AM Deleted: correlation

- 414 a cyclone. Ekman pumping would be expected to produce negative anomalies in cyclones 415 with a positive "halo" and positive anomalies in anticyclones with a negative "halo" (Gaube 416 et al. (2014), Fig. 2). Trapping of chlorophyll involves eddies retaining the properties that 417 they had when shed from a boundary current, which would generally imply low values in 418 anticyclones and high values in cyclones. Eddy intensification would be expected to produce 419 the same picture, as cyclones would see rising nutriclines in the center but anticyclones would 420 see deepening nutriclines. The basic picture seen in the Arabian Sea is inconsistent with the 421 first two mechanisms but is potentially consistent with the second two. However, without in-422 situ data it is impossible to validate either of these mechanisms.
- 423

#### 424 4 Numerical modelling results

#### 425 4.1 Temporal variability

Time series of chlorophyll-a, phosphate and nitrate for all GFDL models are shown in Fig. 7a 426 427 to c within the whole region and compared against the corresponding GSM5 satellite results or WOA09. Note that the eight years of the model output, selected as the last eight years of 428 429 the run, would not be expected to correspond to the eight actual years in the satellite data. The 430 annual cycles of chlorophyll-a and biomass are quite similar to each other in all GFDL 431 models, insofar as they show two distinct blooms in yearly cycle. The maximum values that 432 can be considered as a winter bloom in the whole region are mostly seen around February (Piontkovski et al., 2011), with values of 0.32–0.38, 0.48–0.62, 1.0–2.0, 1.5-2.2, 0.8-1.6, and 433 0.6-0.75 mgm<sup>-3</sup> for chlorophyll in CORE-TOPAZ, Coupled-TOPAZ, CM2.6 (miniBLING), 434 miniBLING (Low resolution), BLING and satellite data, respectively. A summer bloom is 435 also pronounced in September as a second maximum in the yearly cycle over the whole 436 region, with peak magnitudes of about 0.25-0.52, 0.65-0.7, 0.65-1.15, 0.8-1.15, 0.5-0.75, and 437 0.75-1.3 mgm<sup>-3</sup> for chlorophyll across the different datasets. 438 439 Notice the results from the BLING model run in the coarser resolution ESM2M code (purple

lines). The differences between BLING and miniBLING (light blue lines) in this code are just
due to having fixed iron<u>and no dissolved phosphorus</u> in miniBLING. The light field in these
ESM2M runs is computed from using TOPAZ-derived chlorophyll, so that all three models
see identical physical conditions. Both BLING and miniBLING in ESM2M produce an
asymmetry in chlorophyll between February and September that is similar to that produced in

Anand 1/7/2016 10:43 AM Deleted: Annual cycles of variation

CM2.6 miniBLING. This asymmetry is not seen in TOPAZ. Analysis of what drives this asymmetry shows that it is not straightforward. All of the model runs show an asymmetry in the nutrient concentrations that is in the opposite direction as the observations, with higher nutrients in February than in September, as shown in Fig 7b. As we will show later in the manuscript, this is probably associated with the models mixing to excessive depth during the wintertime. However, in TOPAZ this does not produce an asymmetry in chlorophyll, while in BLING and miniBLING it does. There are two possible reasons for this:

453 1) The equilibrium assumption, which means that biomass in both BLING and miniBLING is 454 not directly simulated. In TOPAZ, the growth of plankton during the spring is limited by the 455 biomass of phytoplankton, whereas in the fall TOPAZ continues to have higher heterotrophic 456 biomass (diagnosed from growth rates over previous months) that then grazes the plankton. In 457 BLING and miniBLING, by contrast, the biomass responds almost instantaneously to changes in growth conditions. This means that if the growth rate increases from 0.05  $day_{1}^{-1}$  to 0.1  $day_{2}^{-1}$ 458 459 over the course of a month, the biomass associated with large phytoplankton will increase 460 eightfold, even though the additional growth should only be enough to give an increase of a factor of 30 days  $\times 0.05$  day<sup>-1</sup>=1.5. Possibilities for addressing this effect include replacing the 461 DOP tracer with a biomass tracer, which could then partitioned between the different 462 463 phytoplankton boxes based on the temporally smoothed growth rate, or increasing the 464 timescale over which the growth rate is smoothed when biomass is calculated. 465 2) Different handling of light limitation. In TOPAZ light limitation is calculated using the

instantaneous local light, whereas in BLING it is calculated using the mixed layer average light. Preliminary results with a very coarse resolution model using BLING show that this reduces the summer-winter asymmetry slightly, but is not sufficient to make the February bloom smaller than the September bloom. This effect will also be addressed in future research.

471 It is likely that all three of these factors- too deep winter mixed layers leading to too high 472 nutrients, too little light limitation and instantaneous response to changes in growth 473 conditions, are all responsible for the overly strong blooms in boreal winter in the Arabian 474 Sea.

To get a better sense of the mechanisms driving the blooms in the model, the biomass (mol P  $kg^{-1}$ ) of the miniBLING CM2.6 model is compared with the light intensity in the mixed layer and the light-saturated photosynthesis rate (carbon specific) (s<sup>-1</sup>) in Figs. 8a and b for January

Anand 1/7/2016 10:48 AM Formatted: Superscript Anand 1/7/2016 10:48 AM Formatted: Superscript

478 of year 195. The two terms in Fig. 8 are the two terms in the model that affect growth rate.

479 Because biomass in the miniBLING model is a function of growth rate only, <u>understanding</u>

480 the variation in two terms is sufficient to understand what drives the variation of biomass, in

the model. The biomass production and mixed layer light intensity (Fig. 8a) are not meaningfully correlated parameters. On the other hand, the biomass and the light-saturated carbon specific growth rate (Fig. 8b; indicating the degree of nutrient limitation) are positively correlated. From this, it can be concluded that the blooms in this region are more driven by nutrient rather than light, consistent with, for example, Gomes et al. (2008). This suggests in turn that it is likely biases in nutrient supply that drive biases in productivity.

487 We can get more insight into nutrient biases by examining the individual tendency terms 488 associated with advection, vertical diffusion and subgridscale eddy fluxes and time rate of 489 change of nutrients. For simplicity, in this paper we combine the vertical diffusive flux associated with small-scale mixing from the background diffusion with that due to the mixed 490 491 layer parameterization. Fig. 9 shows PO<sub>4</sub> advection, diffusion and time tendency flux terms 492 for the whole region (56°-66°E, 15°-26°N) over a typical year. We calculate these by 493 integrating the time tendency terms for phosphate over the top 50m. The results show that the 494 dominant source in whole region during the winter bloom is diffusion, suggesting the model 495 predicts excessively strong mixing during the wintertime. By contrast, the advection dominates diffusion during the summer bloom, supplying the majority of nutrients during the 496 497 months of July and August, The fact that the summertime bloom is close to observations 498 suggests that the model correctly simulates this wind-driven upwelling.

499 In addition to having annual cycles that are different from observations, the models also differ 500 from data in terms of interannual variability. As shown in Fig. 10, low-resolution models 501 (CORE- and coupled-TOPAZ) provide an almost uniform seasonal coefficient of variation 502 (mean C.o.Vs are 0.15 and 0.18, respectively), while both data and eddy resolving CM2.6 503 models show higher interannual variability and seasonal changes (mean C.o.Vs are 0.35 and 504 0.5, respectively). The C.o.Vs are particularly higher during the winter and summer blooms in 505 the observations, while the low-resolution models do not see these signals. In other words, the low-resolution models fail to get enough variability, while the high-resolution models produce 506 507 too much interannual variability. Together with the Fig. 4, this statistical analysis suggests 508 that eddies are necessary to explain the variability in the data as opposed to the low-resolution

Anand 1/7/2016 10:55 AM

**Deleted:** it is in fact sufficient to understand what drives the growth

Shahabeddin Tora..., 1/10/2016 10:04 AM Deleted: 10 Anand 1/7/2016 10:58 AM Deleted: , which is Anand 1/7/2016 10:58 AM Deleted: averaged flux values over the upper 50 meter calculated for phosphate (PO<sub>4</sub>).

Anand 1/7/2016 10:59 AM Deleted: in Anand 1/7/2016 10:59 AM Deleted: particularly Anand 1/7/2016 10:59 AM Deleted: and Anand 1/7/2016 10:59 AM Deleted: t

519 models, but that the high-resolution model does not properly capture this variability. Below,

520 we examine the relationship of eddies and blooms in the high-resolution models.

521

# 522 4.2 Blooms and sea surface height in CM2.6

## 523 4.2.1 Large-scale correlation

524 The relationship between SSHA and chlorophyll is quite different in the model as compared 525 to the satellite. Monthly variation in the cross-correlation of chlorophyll and SSHA for eight 526 consequent years in CM2.6 is shown in Fig. 11. As in the remote sensing, the model shows 527 annual cycles of variation in the cross-correlation, suggesting a repeatable yearly phenomenon 528 in the region. However the structure of this annual cycle is not consistent with the satellite 529 data. The model predicts several months (i.e. March-August) with anti-correlation for most of 530 the years, but with values less than 0.5, smaller than the peak anti-correlation values in 531 satellite results. The model also predicts that several other months (i.e. October-February) 532 should have no or even positive correlation, while the satellite shows strong negative 533 correlations during these months.

534

## 535 4.2.2 Blooms in Mesoscale Structures

536 Why does the GFDL CM2.6 model not produce the same relationship between SSHA and 537 chlorophyll as the satellite? We can gain some insight by examining snapshots of the two fields. In Figs. 12a and b, sea surface chlorophyll-a concentration and sea surface height 538 anomaly (SSHA) are shown at two snapshots of time. November 9th and December 28th for 539 model year 195. Comparing the figures with the corresponding satellite results in Fig. 3 for 540 541 the months of November and December, we see that the southern part of the GFDL model is 542 more similar to the satellite data, with high concentrations of chlorophyll-a tending to be located at the center of cyclones. In contrast, in the northern part of the region, the GFDL 543 544 model predicts high chlorophyll at the edges of the cyclones as well as in the center of 545 anticyclones. The eddy structures have smaller diameters in GFDL results than the field 546 observations, though it is not clear whether this represents smoothing in the AVISO product 547 or some physical weakness of the model.

548 We now focus on the few examples in our model output where chlorophyll blooms are found in the center of cyclonic eddies. These are denoted as E1 and E2 in Figs. 12a and b. To track 549 the movement of the selected eddies, E1 and E2, over the time from November 9<sup>th</sup> to 550 December 28th, modeled chlorophyll and SSHA are shown in Figs. 12c and d along two 551 552 different latitudes, 16°N (for E1) and 19°N (for E2). Fig. 12c shows that E1 moved westward 553 during this period of time, and that the chlorophyll concentration was kept high within the 554 central part of the eddy. E1 appears to be created by the passage of a cyclone, similar to the 555 eddy observed by Wang and Zhao (2008) in the aftermath of Cyclone Gonu. Similarly, as 556 shown in Figs. 12d, E2 was a persistent eddy with both central and edge blooms during the 557 month of November that started to move towards the west during the December along 19°N. However, at other latitudes, the largest blooms offshore are found along gradients in SSH 558 559 rather than being associated with maxima or minima. This suggests a different mechanism for 560 producing blooms in the model. Following Gaube et al. (2014). it appears that the eddy stirring mechanism is dominant. Satellite data (i.e. see Fig. 3 for the month of May) provide 561 562 some hints of high-chlorophyll plumes being advected away from coastal regions. As shown 563 in Figs. 12a and b, high velocities in the marginal region between adjacent cyclonic and 564 anticyclonic eddies can cause such plumes in the GFDL models as well. 565 Why is the model only able to simulate the relationship between SSH and chlorophyll in the southern part of the domain? We hypothesize this is due to differences in stratification 566

between the two regions. The average water temperature (colors) and the macronutrient (PO<sub>4</sub>) 567 568 concentrations (contours) for model year 197 are compared to the corresponding measured 569 values in World Ocean Atlas (WOA09) within the upper 200m in the northern (60°-66°E and 19°-23°N) and southern (60°-66°E and 15°-17°N) part of the region are shown in Fig. 13. In 570 571 the northern part of the region (see Fig 13a and b), the GFDL model provides a reasonably 572 good estimation of the mean temperature field near the surface, but subsurface temperatures 573 are not as consistent as there is far too little stratification. This is also associated with a very 574 weak nutricline in CM2.6. Variations in isopycnal depth will therefore not lead to big differences in nutrient supply. Figs. 13c and d show the same fields for the southern part of 575 576 the region. Unlike the northern part of the domain, the temperature gradient over these depths 577 is well estimated by CM2.6. While the nutricline is still too weak there is some gradient in 578 nutrients between 80 and 120m.

579 As seen in Fig. 1d both the ARGO and WOA09 wintertime mixed layer depth is considerably

Anand 1/7/2016 11:03 AM Deleted: for the existence Anand 1/7/2016 11:03 AM Deleted: advective

#### Anand 1/7/2016 11:03 AM Deleted: S Shahabeddin Torab..., 1/10/2016 9:57 AM Formatted: Subscript

Shahabeddin Torab..., 1/12/2016 8:22 PM Deleted: results

Anand 1/7/2016 11:04 AM Deleted: ( Anand 1/7/2016 11:04 AM Deleted: )

586 deeper than the summertime mixed layer depth, reaching a maximum of 65m. However, in

587

the northern regions of the model the MLD seems to be too deep in winter, reaching values of

588 130-150 m. This suggests that the overly deep mixed layer in the northern part of the region 589 may explain both the tendency towards an overly strong winter bloom and the failure of 590 mesoscale eddies in modulating chlorophyll blooms. If we look during the time period where 591 we have eddies E1 and E2 (Nov-Dec. year 197, Fig 6c,d) we see shallower mixed layers 592 associated with both eddies.

593 Both the temperature and mixed layer biases in the northern part of the Arabian Sea may 594 result from having too much water from the Persian Gulf in this region. This can be seen in 595 the yearly averaged subsurface salinity-density distribution over the region, shown in Figs 13e 596 and f for both WOA09 data and CM2.6 (model year 197), respectively. Fig. 13e shows two 597 separate tongues of salty water, one near the surface and one at the depth of ~300m. These 598 salty water signals are consistent with the seasonal cycle of Persian Gulf outflow as discussed 599 in Ezam et al. (2010). On the other hand, CM2.6 shows one subsurface salty water signal 600 from the northern part, which is deep and strong enough to result in weak stratification in the 601 north to a depth of 250m, as shown in Fig. 13f. These results suggest that a sharp thermocline 602 and nutricline is necessary for eddy activity to modulate the mixing of nutrients to the surface.

603 We test the idea that a sharper thermocline could modulate mixing of nutrients to the surface 604 by looking at the sources of nutrient in the southern part of Arabian Sea where eddy-bloom 605 relationships similar to observations are occasionally seen. Accordingly, the region containing 606 eddy E1 in Fig. 12 is analyzed to determine the physical mechanisms by which nutrient is 607 transported into the surface layer. Fig. 14 contrasts chlorophyll concentration, advection, and 608 diffusion terms for the region from 63°-66°E, 15°-18°N over the December of two consecutive 609 CM2.6 years of 197 and 198. In December of year 197 we see an eddy associated with a 610 bloom while there is no eddy in December of year 198, and the chlorophyll concentrations are 611 much lower. In both years the diffusive flux of nutrient to the top 50m mirrors the 612 chlorophyll. But in 197 it is larger and positive (~10 mol/m<sup>2</sup>/month) in the eddy while the 613 advective flux is actually negative in this region. By contrast in Year 198, there is no cyclonic 614 eddy and the diffusive fluxes are much smaller.

The bloom associated with eddies E1 and E2 do not fit with any of the mechanisms highlighted in Gaube et al. (2014). We first consider the mechanism of trapping. Eddy E1 is

617 generated in the ocean interior, not as a result of coastal upwelling. As shown in Fig. 15, the

Anand 1/7/2016 11:04 AM Deleted: at of

Shahabeddin Torab..., 1/10/2016 1:18 PM Deleted: Anand 1/7/2016 11:07 AM Deleted: Y Anand 1/7/2016 11:07 AM Deleted: at the same time

622 nutrient supply rate ranges between 5 and 8 mmol/m<sup>2</sup>/month in the eddy. The concentrations

623 in this eddy are only 0.01  $\mu$ M (5 mmol/m<sup>2</sup>) over the top 50 m. It cannot be the case that the

nutrients in the eddy can last for several months as a result of "trapping", there must be a

625 continuous supply. Moreover although eddy E2 shows a horizontal advection signal in

626 November (with a positive ring around the edge in Fig. 12a), the signal in December has the

627 opposite sign. Eddy intensification is also an unlikely mechanism for explaining the blooms, 628 as dSSH/dt is relatively small (particularly if we track the minimum SSH associated with E1

629 in Fig. 12c or E2 in Fig. 12d). Finally, Ekman pumping signatures in Gaube et al. (2014) have

630 the opposite sign as what is seen in E1 and E2.

631 Our results also contrast with those in Resplandy et al. (2011). The focus in Resplandy et al. 632 (2011) is on the productivity driven by horizontal and vertical advection in summer and 633 mostly vertical advection in winter. This contradicts our finding of a primary diffusive source 634 of nutrient in winter although it is consistent the finding of advective source of nutrients in 635 summer. We point out that in our model, the only two eddies that actually look like what we 636 see in the satellite observations involve enhanced mixing from below. This is a different result 637 from Levy et al. (2014) and Resplandy et al. (2011). Moreover it is not clear whether these 638 papers get the seasonal correlation with SSH or not. Resplandy et al. (2011) do not focus on 639 structures at the eddy scale as they are more concerned with the net impact of eddies.

640

641 To summarize, we hypothesize that

642 1. The reason that blooms are found in cyclones in the Arabian Sea during the NEM is that
643 the dominant source of nutrients to the surface, i.e. mixing (Barimalala et al., 2013;
644 Kawamiya and Oschlies, 2003) is concentrated there.

645 2. Interannual variability in wintertime blooms in the Northwest Arabian Sea is controlled by646 the combined presence of these eddies and strength of wintertime cooling.

647 3. Excessive mixing (resulting in too weak a thermocline) prevents mixing from being
648 modulated by eddies in the model except occasionally in the southern part of our region. In
649 the real world the modulation of mixing seen in Fig. 14 extends into the Northwest Arabian
650 Sea and the Gulf of Oman.

651

### 652 5 Conclusions

653 Our analysis of bloom variability in the northwestern Arabian Sea and Gulf of Oman has 654 illustrated both similar and dissimilar descriptive features between satellites and a suite of 655 models and explored the various mechanisms involved. Satellite analyses demonstrate the 656 existence of two blooms, the stronger one associated with the Southwest Monsoon and the 657 weaker one associated with the Northeast Monsoon as also shown by Madhupratap et al. 658 (1996), Kawamiya and Oschlies (2003), Murtugudde et al. (2007), and Al-Azri et al. (2010). 659 We demonstrate a pronounced anti-correlation between SSHA and chlorophyll blooms during 660 certain times in northern winter but a much weaker relationship in other months (typically northern summer) with the relationship disappearing as the blooms vanish in the months of 661 April and May (northern spring). While the depth of thermocline and nutricline and also the 662 stratification affect the convection during the Northeast Monsoon (Dickey et al., 1998; Kumar 663 et al., 2001; Wiggert et al., 2002), we show that a thin nutricline/thermocline and a strong 664 stratification are also required to enable cold eddies to bring nutrients to euphotic zone and 665 develop phytoplankton blooms. During the wintertime monsoon, while both cooling in the 666 winter and eddies control the blooms, variability in bloom location will arise from variability 667 in the location of eddies, and so may not be predictable. In contrast, during the Southwest 668 669 Monsoon the dominant upwelling associated with the intense environmental forcing 670 supersedes the effect of eddies and the activity of the cold eddies is not pronounced.

671 Understanding of this phenomenon has been sought using five different 3D ocean-atmosphere 672 models, including a CORE-forced ocean with the TOPAZ biogeochemistry, a coupled model 673 with the TOPAZ biogeochemistry and CM2.6. Because the coarse models with TOPAZ are not able to capture eddies and the interannual variability, CM2.6 (miniBLING), a eddy-674 675 resolving high resolution model, was also considered for simulating the spatial and temporal 676 changes of the bloom in the region. This model simulates the two blooms seen in the data and shows that the nutrients driving the northern summer bloom are supplied by advection while 677 678 those driving the wintertime bloom are supplied by vertical diffusion. However, this model is 679 unable to simulate the seasonal relationship observed in the satellite products between blooms 680 and sea surface height. Although there is some anti-correlation, it tends to be associated with 681 larger spatial scales and not really related to eddies. Instead, eddies in the model usually wrap 682 the chlorophyll around themselves, producing high chlorophyll concentrations around their 683 edges and not at their centers. Comparing the model results to field measurements (WOA09) Anand 1/7/2016 11:09 AM Deleted: a set of Anand 1/7/2016 11:09 AM Deleted: d among Anand 1/7/2016 11:09 AM Deleted: verified Anand 1/7/2016 11:10 AM Deleted: m Anand 1/7/2016 11:10 AM Deleted: M Shahabeddin Torabian 1/6/2016 9:08 PM Deleted: in

#### Anand 1/7/2016 11:10 AM Deleted: are Anand 1/7/2016 11:10 AM Deleted: ing Anand 1/7/2016 11:10 AM Deleted: o

Shahabeddin Torab..., 1/12/2016 8:23 PM Deleted: three

showed that the model does not account for the strong thermocline and nutricline in the

northern part of the region. In the wintertime, this leads to excessive convective supply of

nutrients and too strong of a bloom. However, for a few cases, eddies with blooms at the 696 697 center are tracked in the southern part of the domain. In this region, consistency is observed 698 between the model results and the field data. Analysis of the term balances in mixed layer 699 show that eddies in this region modulate the diffusive supply of nutrients. We suggest that 700 what happens in the model in the Southern Arabian Sea actually describes the Arabian Sea as 701 a whole according to the observations and the field data. The model misses the eddy signal in 702 the north because it lacks a thin nutricline, motions of which will lead to differences in 703 nutrient supply. In the real world, eddies modulate the diffusive supply of nutrients during the 704 wintertime and there is more mixing in the eddy centers along with the diffusive supply 705 provided by the cooling in the wintertime.

706 Accordingly, there is a potential to improve the numerical models by better simulating the 707 Persian Gulf Outflow to produce a sharper thermocline, allowing more realistic nutrient 708 supply. Overflows are difficult to simulate in level-coordinate models because they are prone 709 to excessive entrainment of the dense plume (Winton et al., 1998). While significant effort 710 has gone into simulating the Denmark Straits overflow at coarse resolution (Legg et al., 2009), our results show that smaller overflows such as the Persian Gulf may be regionally 711 712 significant. This may provide further impetus for using isopycnal models in high resolution 713 simulations, as such models can potentially simulate such overflows with greater fidelity.

714 It is worth noting that regional models, (such as Resplandy et al. (2011)) do have the potential 715 to better simulate the hydrography of the Northern Arabian Sea. Because such models are 716 very tightly constrained through "sponges" that restore hydrography at the boundaries, they 717 may not have the problems that global models do at representing the effects of overflows that 718 they do not properly simulate. However, such models cannot by themselves simulate the 719 effects of changing climate, which in turn changes the boundary conditions. For this reason, 720 global models must still be used for projection, making it important to identify the reasons 721 that they are not going to work.

722

723

724

Deleted: a bloom

Shahabeddin Toral

1/12/2016 8:11 PM

# 726 Acknowledgements

- 727 The authors thank Eric Galbraith, Shahabeddin Torabian, Grace Kim, Carlos del Castillo, and
- 728 Jeremy Wardell for useful discussions. We also thank Rick Slater and Whit Anderson for
- 729 their support of the model simulations. Argo climatology data were collected and made freely
- 730 available by the International Argo Program and the national programs that contribute to it.
- 731 (http://www.argo.ucsd.edu, http://argo.jcommops.org). The Argo Program is part of the
- 732 <u>Global Ocean Observing System.</u>
- 733
- 734

#### 735 References

- Adcroft, A., Campin, J.-M., Hill, C. and Marshall, J.: Implementation of an Atmosphere-736 737 Ocean General Circulation Model on the Expanded Spherical Cube, Mon. Weather Rev., 132(12), 2845-2863, doi:10.1175/MWR2823.1, 2004. 738
- 739 Al-Azri, A. R., Piontkovski, S. A., Al-Hashmi, K. A., Goes, J. I., Gomes, H. D. R. and 740 Glibert, P. M.: Mesoscale and Nutrient Conditions Associated with the Massive 2008 741 Cochlodinium polykrikoides Bloom in the Sea of Oman/Arabian Gulf, Estuaries and Coasts, 742 doi:10.1007/s12237-013-9693-1, 2013.
- Al-Azri, A. R., Piontkovski, S. A., Al-Hashmi, K. A., Goes, J. I. and Gomes, H. R.: 743 744 Chlorophyll a as a measure of seasonal coupling between phytoplankton and the monsoon 745 periods in the Gulf of Oman, Aquat. Ecol., 44(2), 449-461, doi:10.1007/s10452-009-9303-2, 746 2010.
- 747 Anderson, D. M. and Prell, W. L.: A 300 KYR Record of Upwelling Off Oman During the 748 Late Quaternary: Evidence of the Asian Southwest Mansoon, Paleoceanography, 8(2), 193-749 208, 1993.
- 750 ARGO: Climatology, [online] Available from: http://www.argo.ucsd.edu, 2015.
- 751 Armstrong, R. A., Lee, C., Hedges, J. I., Honjo, S. and Wakeham, S. G.: A new, mechanistic model for organic carbon fluxes in the ocean based on the quantitative association of POC 752 753 with ballast minerals, Deep. Res. Part II Top. Stud. Oceanogr., 49(1-3), 219-236,
- 754 doi:10.1016/S0967-0645(01)00101-1, 2002.
- 755 Banse, K. and McClain, C. R.: Winter blooms of phytoplankton in the Arabian Sea as observed by the Coastal Zone Color Scanner, Mar Ecol Prog Ser, 34, 201-211, 1986. 756
- 757 Barimalala, R., Bracco, A., Kucharski, F., McCreary, J. P. and Crise, A.: Arabian Sea 758 ecosystem responses to the South Tropical Atlantic teleconnection, J. Mar. Syst., 117-118, 759 14-30, doi:10.1016/j.jmarsys.2013.03.002, 2013.
- 760 Bartolacci, D. M. and Luther, M. E.: Patterns of co-variability between physical and biological parameters in the Arabian Sea, Deep Sea Res. Part II Top. Stud. Oceanogr., 46(8-761 9), 1933-1964, doi:10.1016/S0967-0645(99)00049-1, 1999. 762
- 763 Behrenfeld, M. J.: Abandoning Sverdrup's Critical Depth Hypothesis on phytoplankton 764 blooms, Ecology, 91(4), 977-989, doi:10.1890/09-1207.1, 2010.
- 765 Behrenfeld, M. J., Boss, E., Siegel, D. A. and Shea, D. M.: Carbon-based ocean productivity and phytoplankton physiology from space, Global Biogeochem. Cycles, 19(1), 1-14, 766 doi:10.1029/2004GB002299, 2005. 767
- 768 Bryan, K. and Lewis, L. J.: A Water Mass Moel of the World Ocean, 84(8), 2503, 1979.
- 769 Danabasoglu, G., Large, W. G., Tribbia, J. J., Gent, P. R., Briegleb, B. P. and McWilliams, J. C.: Diurnal coupling in the tropical oceans of CCSM3, J. Clim., 19(11), 2347-2365, 770 771 doi:10.1175/JCLI3739.1, 2006.
- 772 Delworth, T. L., Broccoli, A. J., Rosati, A., Stouffer, R. J., Balaji, V., Beesley, J. A., Cooke,
- 773 W. F., Dixon, K. W., Dunne, J., Dunne, K. A., Durachta, J. W., Findell, K. L., Ginoux, P.,
- Gnanadesikan, A., Gordon, C. T., Griffies, S. M., Gudgel, R., Harrison, M. J., Held, I. M., Hemler, R. S., Horowitz, L. W., Klein, S. A., Knutson, T. R., Kushner, P. J., Langenhorst, A. 774
- 775
- 776 R., Lee, H. C., Lin, S. J., Lu, J., Malyshev, S. L., Milly, P. C. D., Ramaswamy, V., Russell, J.,

23

Shahabeddin Torab..., 1/12/2016 8:06 PM

Deleted:

- 778 Schwarzkopf, M. D., Shevliakova, E., Sirutis, J. J., Spelman, M. J., Stern, W. F., Winton, M.,
- Wittenberg, A. T., Wyman, B., Zeng, F. and Zhang, R.: GFDL's CM2 global coupled climate
  models. Part I: Formulation and simulation characteristics, J. Clim., 19(5), 643–674,
- 781 doi:10.1175/JCLI3629.1, 2006.
- Delworth, T. L., Rosati, A., Anderson, W., Adcroft, A. J., Balaji, V., Benson, R., Dixon, K.,
  Griffies, S. M., Lee, H.-C., Pacanowski, R. C., Vecchi, G. A., Wittenberg, A. T., Zeng, F. and
- 784 Zhang, R.: Simulated Climate and Climate Change in the GFDL CM2.5 High-Resolution
- 785 Coupled Climate Model, J. Clim., 25(8), 2755–2781, doi:10.1175/JCLI-D-11-00316.1, 2012.
- Dickey, T., Marra, J., Sigurdson, D. E., Weller, R. A., Kinkade, C. S., Zedler, S. E., Wiggert,
  J. D. and Langdon, C.: Seasonal variability of bio-optical and physical properties in the
- Arabian Sea : October 1994 October 1995, Deep. Res. II, 45(October 1994), 2001–2025, 1998.
- Dunne, J., Gnanadesikan, A., Sarmiento, J. L. and Slater, R. D.: Technical description of the
  prototype version (v0) of Tracers Of Phytoplankton with Allometric Zooplankton (TOPAZ)
  ocean biogeochemical model as used in the Princeton IFMIP\* model, Biogeosciences Suppl.,
  7(1), 3593, doi:10.5194/bg-7-3593-2010, 2010.
- Dunne, J. P., John, J. G., Adcroft, A. J., Griffies, S. M. and Hallberg, R. W.: GFDL's ESM2
  Global Coupled Climate-Carbon Earth System Models. Part I: Physical Formulation and
  Baseline Simulation Characteristics, J. Clim., 25, 6646–6665,
- 797 doi:http://dx.doi.org/10.1175/JCLI-D-11-00560.1, 2012.
- Dunne, J. P., John, J. G., Shevliakova, S., Stouffer, R. J., Krasting, J. P., Malyshev, S. L.,
  Milly, P. C. D., Sentman, L. T., Adcroft, A. J., Cooke, W., Dunne, K. A., Griffies, S. M.,
  Hallberg, R. W., Harrison, M. J., Levy, H., Wittenberg, A. T., Phillips, P. J. and Zadeh, N.:
  GFDL's ESM2 global coupled climate-carbon earth system models. Part II: Carbon system
  formulation and baseline simulation characteristics, J. Clim., 26(7), 2247–2267,
  doi:10.1175/JCLI-D-12-00150.1, 2013.
- Bunne, J. P., Sarmiento, J. L. and Gnanadesikan, A.: A synthesis of global particle export
  from the surface ocean and cycling through the ocean interior and on the seafloor, Global
  Biogeochem. Cycles, 21(4), 1–16, doi:10.1029/2006GB002907, 2007.
- Ezam, M., Bidokhti, A. A. and Javid, <u>A</u> H.: Numerical simulations of spreading of the Persian
  Gulf outflow into the Oman Sea, Ocean Sci., 6(4), 887–900, doi:10.5194/os-6-887-2010,
  2010.
- Fischer, A. S., Weller, R. A., Rudnick, D. L., Eriksen, C. C., Lee, C. M., Brink, K. H., Fox, C.
  A. and Leben, R. R.: Mesoscale eddies, coastal upwelling, and the upper-ocean heat budget in
  the Arabian Sea, Deep Sea Res. Part II Top. Stud. Oceanogr., 49(12), 2231–2264,
  doi:10.1016/S0967-0645(02)00036-X, 2002.
- Fox-Kemper, B., Ferrari, R. and Hallberg, R.: Parameterization of Mixed Layer Eddies. Part
  I: Theory and Diagnosis, J. Phys. Oceanogr., 38(6), 1145–1165, doi:10.1175/2007JPO3792.1,
- 816 2008.
- 817 Galbraith, E. D., Dunne, J. P., Gnanadesikan, A., Slater, R. D., Sarmiento, J. L., Dufour, C.
- 818 O., de Souza, G. F., Bianchi, D., Claret, M., Rodgers, K. B. and Sedigh Marvasti, S.: 819 Complex functionality with minimal computation: Promise and pitfalls of reduced-tracer
- 820 ocean biogeochemistry models, J. Adv. Model. Earth Syst., 1–17,
- 821 doi:10.1002/2015MS000463, 2015.

- Gaube, P., McGillicuddy, D., Chelton, D., Behrenfeld, M. J. and Strutton, P.: Regional
   variations in the influence ofmesoscale eddies on near-surface chlorophyll Peter, J. Geophys.
- 824 Res. Ocean., 119, 8195–8220, doi:10.1002/2014JC0101111.Received, 2014.
- 825 Gnanadesikan, A., Dixon, K. W., Griffies, S. M., Balaji, V., Barreiro, M., Beesley, J. A.,
- Cooke, W. F., Delworth, T. L., Gerdes, R., Harrison, M. J., Held, I. M., Hurlin, W. J., Lee, H.
  C., Liang, Z., Nong, G., Pacanowski, R. C., Rosati, A., Russell, J., Samuels, B. L., Song, Q.,
- Spelman, M. J., Stouffer, R. J., Sweeney, C. O., Vecchi, G., Winton, M., Wittenberg, A. T.,
- Zeng, F., Zhang, R. and Dunne, J. P.: GFDL's CM2 global coupled climate models. Part II:
- 830 The baseline ocean simulation, J. Clim., 19(5), 675–697, doi:10.1175/JCLI3630.1, 2006.
- Gnanadesikan, A., Dunne, J. P. and John, J.: What ocean biogeochemical models can tell us
  about bottom-up control of ecosystem variability, ICES J. Mar. Sci., 68(6), 1030–1044,
  doi:10.1093/icesjms/fsr068, 2011.
- Gnanadesikan, A., Dunne, J. P. and Msadek, R.: Connecting Atlantic temperature variability
  and biological cycling in two earth system models, J. Mar. Syst., 133, 39–54,
  doi:10.1016/j.jmarsys.2013.10.003, 2014.
- Goes, J. I., Thoppil, P. G., Gomes, H. D. R. and Fasullo, J. T.: Warming of the Eurasian
  landmass is making the Arabian Sea more productive., Science, 308(5721), 545–547,
  doi:10.1126/science.1106610, 2005.
- 840 Gomes, R., Goes, J. I., Matondkar, S. G. P., Parab, S. G., Al-azri, A. R. N. and Thoppil, P. G.:
- Beep-Sea Research I Blooms of Noctiluca miliaris in the Arabian Sea An in situ and
  satellite study, Deep Sea Res. Part I Oceanogr. Res. Pap., 55, 751–765,
  doi:10.1016/j.dsr.2008.03.003, 2008.
- Griffies, S. M., Biastoch, A., Böning, C., Bryan, F., Danabasoglu, G., Chassignet, E. P.,
  England, M. H., Gerdes, R., Haak, H., Hallberg, R. W., Hazeleger, W., Jungclaus, J., Large,
  W. G., Madec, G., Pirani, A., Samuels, B. L., Scheinert, M., Gupta, A. Sen, Severijns, C. A.,
  Simmons, H. L., Treguier, A. M., Winton, M., Yeager, S. and Yin, J.: Coordinated Ocean-ice
- 848
   Reference
   Experiments
   (COREs),
   Ocean
   Model.,
   26(1-2),
   1-46,

   849
   doi:10.1016/j.ocemod.2008.08.007, 2009.

   <t
- B50 Griffies, S. M., Gnanadesikan, <u>A.</u>, Dixon, K. W., Dunne, J. P., Gerdes, R., Harrison, M. J.,
  B71 Rosati, <u>A.</u>, Russell, J. L., Samuels, B. L., Spelman, M. J., Winton, M. and Zhang, R.:
  B72 Formulation of an ocean model for global climate simulations, Ocean Sci. Discuss., 2(3),
- 853 165–246, doi:10.5194/osd-2-165-2005, 2005.
- Hamzehei, S. and Bidokhti, A.: Red tide monitoring in the Persian Gulf and Gulf of Oman
  using MODIS sensor data., Tech. J. ..., 1100–1107, 2013.
- Honjo, S., Dymond, J., Prell, W. and Ittekkot, V.: Monsoon-controlled export fluxes to the
  interior of the Arabian Sea, Deep. Res. II, 46(1999), 1859–1902, 2000.
- Kawamiya, M. and Oschlies, A.: An eddy-permitting , coupled ecosystem-circulation model
  of the Arabian Sea : comparison with observations, J. Mar. Syst., 38, 221–257, 2003.
- Klaas, C. and Archer, D. E.: Association of sinking organic matter with various types of
  mineral ballast in the deep sea: Implications for the rain ratio, Global Biogeochem. Cycles,
  16(4), 1116, doi:10.1029/2001GB001765, 2002.
- Kostadinov, T. S., Siegel, D. A. and Maritorena, S.: Retrieval of the particle size distribution
   from satellite ocean color observations, J. Geophys. Res., 114(C9), C09015,

- 865 doi:10.1029/2009JC005303, 2009.
- Kumar, S. P., Ramaiah, N., Gauns, M., Sarma, V. V. S. S., Muraleedharan, P. M.,
  Raghukumar, S., Kumar, M. D. and Madhupratap, M.: Physical forcing of biological
  productivity in the Northern Arabian Sea during the Northeast Monsoon, Deep. Res. II, 48,
- 869 1115–1126, 2001.
- Large, W. G., McWilliams, J. C. and Doney, S. C.: Oceanic vertical mixing: A review and a
  model with a nonlocal boundary layer parameterization, Rev. Geophys., 32(4), 363,
  doi:10.1029/94RG01872, 1994.
- 873 Legg, S., Briegleb, B., Chang, Y., Chassignet, E. P., Danabasoglu, G., Ezer, T., Gordon, A.
- L., Griffies, S., Hallberg, R., Jackson, L., Large, W., Özgükmen, T. M., Peters, H., Price, J.,
  Riemenschneider, U., Wu, W., Xu, X. and Yang, J.: Improving oceanic overflow
- 875 Riemenschneider, U., Wu, W., Xu, X. and Yang, J.: Improving oceanic overflow 876 representation in climate models: The Gravity Current Entrainment Climate Process Team,
- 877 Bull. Am. Meteorol. Soc., 90(5), 657–670, doi:10.1175/2008BAMS2667.1, 2009.
- Levy, M., Resplandy, L. and Lengaigne, M.: Oceanicmesoscale turbulence drives large
  biogeochemical interannual variability atmiddle and high latitudes, Geophys. Res. Lett.,
  41(7), 2467–2474, doi:10.1002/2014GL059608, 2014.
- Levy, M., Shankar, D., Andre, J., Shenoi, S. S. C., Durand, F. and Montegut, C. D. B.: Basinwide seasonal evolution of the Indian Ocean 's phytoplankton blooms, J. Geophys. Res.,
  112(C12014), 1–14, 2007.
- Madhupratap, M., Kumar, S., Bhattathiri, P., Kumar, M., Raghukumar, S., Nair, K. and
  Ramaiah, N.: Mechanism of the biological response to winter cooling in the northeastern
  Arabian Sea, Nature, 384(12), 549–552, 1996.
- Maritorena, S., Siegel, D. and Peterson, A. R.: Optimization of a semianalytical ocean color
   model for global-scale applications., Appl. Opt., 41(15), 2705–14, 2002.
- McGillicuddy, D., Kosnyrev, V., Ryan, J. and Yoder, J.: Covariation of mesoscale ocean
  color and sea-surface temperature patterns in the Sargasso Sea, Deep. Res. II, 48, 1823–1836,
  2001.
- Murtugudde, R., Seager, R. and Thoppil, P.: Arabian Sea response to monsoon variations,
   Paleoceanography, 22(4), 1–17, doi:10.1029/2007PA001467, 2007.
- 894 Naqvi, S. W. A., Moffett, J. W., Gauns, M. U., Narvekar, P. V, Pratihary, A. K., Naik, H.,
- Shenoy, D. M., Jayakumar, D. A., Goepfer, T. J., Patra, P. K., Al-Azri, A. and Ahmed, S. I.:
  The Arabian Sea as a high-nutrient, low-chlorophyll region during the late Southwest
- 897 Monsoon, Biogeosciences, 7, 2091–2100, doi:10.5194/bg-7-2091-2010, 2010.
- Piontkovski, S., Al-Azri, A. and Al-Hashmi, K.: Seasonal and interannual variability of
  chlorophyll-a in the Gulf of Oman compared to the open Arabian Sea regions, Int. J. Remote
  Sens., 32(22), 7703–7715, doi:10.1080/01431161.2010.527393, 2011.
- Piontkovski, S., Al-Gheilani, H., Jupp, B., Al-Azri, A. and Al-hashmi, K.: Interannual
  Changes in the Sea of Oman Ecosystem., Open Mar. Biol. J., 6, 38–52, 2012.
- 903 Resplandy, L., Lévy, M., Madec, G., Pous, S., Aumont, O. and Kumar, D.: Contribution of 904 mesoscale processes to nutrient budgets in the Arabian Sea, J. Geophys. Res. Ocean., 116(11),
- 905 1–24, doi:10.1029/2011JC007006, 2011.
- 906 Richlen, M. L., Morton, S. L., Jamali, E. A., Rajan, A. and Anderson, D. M.: The catastrophic

- 2008–2009 red tide in the Arabian gulf region, with observations on the identification and
  phylogeny of the fish-killing dinoflagellate Cochlodinium polykrikoides, Harmful Algae,
  909 9(2), 163–172, doi:10.1016/j.hal.2009.08.013, 2010.
- Sarma, Y. V. B., Al-hashmi, K. and Smith, L. S.: Sea Surface Warming and its Implications
  for Harmful Algal Blooms off Oman, Int. J. Mar. Sci., 3(8), 65–71,
  doi:10.5376/ijms.2013.03.0008, 2013.
- Shalapyonok, A., Olson, R. J. and Shalapyonok, L. S.: Arabian Sea phytoplankton during
  Southwest and Northeast Monsoons 1995 : composition , size structure and biomass from
- individual cell properties measured by # ow cytometry, Deep. Res. II, 48, 1231–1261, 2001.
- Simmons, H. L., Jayne, S. R., St. Laurent, L. C. and Weaver, A. J.: Tidally driven mixing in a
  numerical model of the ocean general circulation, Ocean Model., 6(3-4), 245–263,
  doi:10.1016/S1463-5003(03)00011-8, 2004.
- Stacey, M. W., Pond, S. and Nowak, Z. P.: A Numerical Model of the Circulation in Knight
  Inlet, British Columbia, Canada, J. Phys. Oceanogr., 25(6), 1037–1062, doi:10.1175/15200485(1995)025<1037:ANMOTC>2.0.CO;2, 1995.
- Tang, D., Kawamura, H. and Luis, A. J.: Short-term variability of phytoplankton blooms
  associated with a cold eddy in the northwestern Arabian Sea, Remote Sens. Environ., 81, 82–
  89, 2002.
- Veldhuis, M. J. W., Kraay, G. W., Van Bleijswijk, J. D. L. and Baars, M. A.: Seasonal and spatial variability in phytoplankton biomass, productivity and growth in the northwestern
- 927 Indian Ocean: the southwest and northeast monsoon, 1992-1993, Deep Sea Res. Part I,
- 928 44(3), 425–449, 1997.
- Wang, D. and Zhao, H.: Estimation of phytoplankton responses to Hurricane Gonu over the
  Arabian Sea based on ocean color data, Sensors, 4878–4893, doi:10.3390/s8084878, 2008.
- Wiggert, J. D., Murtugudde, R. G. and Christian, J. R.: Annual ecosystem variability in the
  tropical Indian Ocean : Results of a coupled bio-physical ocean general circulation model,
  Deep. Res. II, 53, 644–676, doi:10.1016/j.dsr2.2006.01.027, 2006.
- Wiggert, J. D., Murtugudde, R. G. and Mcclain, C. R.: Processes controlling interannual
  variations in wintertime (Northeast Monsoon) primary productivity in the central Arabian
  Sea, Deep. Res. II, 49, 2319–2343, 2002.
- Winton, M., Hallberg, R. and Gnanadesikan, A.: Simulation of Density-Driven Frictional
  Downslope Flow in Z -Coordinate Ocean Models, J. Phys. Oceanogr., 28(11), 2163–2174,
- 939 doi:10.1175/1520-0485(1998)028<2163:SODDFD>2.0.CO;2, 1998.
- 940







Figure 1. Monthly average for region from 56°E-66°E, 15°N-26°N: (a) <u>Climatological surface</u> chlorophyll-a, (SeaWIFS) for a nominal year of 2001; (b) Nitrate (WOA09) over top 100m; (c) Temperature over top 100m; (d) WOA09 seasonal mixed layer depth in meters<u>- black line</u> shows result from World Ocean Atlas, red line from ARGO climatology (ARGO, 2015).

Anand 1/7/2016 11:20 AM	
Deleted: S	
Anand 1/7/2016 11:20 AM	
Deleted: in 2001	

Shahabeddin Torab..., 1/10/2016 2:08 PM Formatted: Not Highlight





Figure 2. Monthly variation of organic matter in <u>SeaWiFS</u> satellite data between 1998 and
2005 within 56°-66°E, 15°-26°N (large region); and 60°-62°E, 22°-26°N (small region): (a)
and (d) chlorophyll; (b) and (e) particulate backscatter; (c) and (f) CDOM.



956 Figure 3. Satellite chlorophyll-a in mg/m<sup>3</sup> (colors) and sea-surface height anomaly (SSHA,

957 contours) in <u>cm (contour interval = 5cm), in the</u> Gulf of Oman and Northwest Arabian Sea

- 958 over the course of 2001.
- 959

	Anand 1/7/2016 11:22 AM
	Deleted: C
	Anand 1/7/2016 11:22 AM
$\langle \rangle$	Deleted: S
	Anand 1/7/2016 11:22 AM
	Deleted: S
	Anand 1/7/2016 11:24 AM
	Deleted: meter





Figure 4. Chlorophyll-a in mgm<sup>-3</sup> (colors) and sea surface height anomaly (SSHA, contours)



964

965 966 967

in <u>cm (contour interval=5 cm)</u>, in the Gulf of Oman and Northwest Arabian Sea during. November in different years.

968

Anand 1/7/2016 11:25 AM **Deleted:** meter Anand 1/7/2016 11:23 AM **Deleted:** in





Figure 6. Average monthly cross-correlation with <u>observed</u> SSHA and average monthly
values between 1998 and 2005 within 56°-66°E and 15°-26°N for (a, b) <u>satellite-estimated</u>
chlorophyll; (c, d) <u>satellite-estimated</u> backscatter; (e, f) <u>satellite-estimated</u> CDOM.



Figure 7. Monthly variation of organic matter in satellite data between 1998 and 2005 and GFDL models (8 characteristic years) within  $56^{\circ}-66^{\circ}E$ ,  $15^{\circ}-26^{\circ}N$ : (a) chlorophyll from GFDL models and GSM5 algorithm. (b) PO<sub>4</sub> from the BLING and miniBLING simulations,  $NO_3/16$  from the TOPAZ simulations and observed PO<sub>4</sub> from WOA09. (c) NO<sub>3</sub> from the TOPAZ simulations and observed NO<sub>3</sub> fromWOA09.

34

Formatted: Font color: Auto







Figure 8. Modeled biomass in CM2.6 in P units (mol P kg <sup>-1</sup>) versus: (a) Mixed layer irradiance (Wm<sup>-2</sup>); (b) Light-Saturated photosynthesis rate (carbon specific) (s<sup>-1</sup>) 56°-66°E, 15°-26°N for January of year 195. In the model, biomass is a function of the growth rate smoothed over several days, and the light-saturated photosynthesis rate indicates the extent to which this growth rate is controlled by nutrient limitation.



1002 Figure 9. PO<sub>4</sub> Advection, diffusion and tendency flux from the CM2.6 model over the whole

36

1003 region averaged over top 50 m ( $56^{\circ}$ - $66^{\circ}$ E,  $15^{\circ}$ - $26^{\circ}$ N).



1007Figure 10. Average monthly coefficient of variation (standard deviation/mean) of Chlorophyll1008a in satellite data between 1998 and 2005 and GFDL models (eight characteristics years)1009within (56°-66°E, 15°-26°N) for the satellite data (black), CORE-TOPAZ (red), COUPLED-1010TOPAZ (green) and CM2.6 model with miniBLING (blue) and within the south region (56°-101166°E, 15°-19°N) for CM2.6 (miniBLING, dashed blue).





1016 and 15°-26°N.



1019Figure 12. CM2.6 (miniBLING) Surface chlorophyll-a concentration and sea surface height1020anomaly (SSHA) November and December during a year where the observed eddy-bloom

- 1021 interaction is seen in the Southern part of the Arabian Sea.



1026 1027 1028 1029

Figure 13. (a-d) Seawater temperature (colors, °C) and Phosphate (PO4) concentration (contours, µM) for the northern (top row) and southern (middle row) parts of the central Arabian Sea. ; (e-f) yearly averaged subsurface distribution of salinity (black contours) and potential density (red contours). Left-hand column shows observations with the Persian Gulf 1030 plume centered at 300m, right-hand column results from CM2.6 model with a much broader 1031 mixing of salinity over the top 200m.





1035 Figure 14. Surface chlorophyll in mg/m<sup>3</sup>. (a-b), Advective flux of phosphate to top 50m in mol/m<sup>2</sup> (c-d,colors), and diffusive flux of phosphate in mol/m<sup>2</sup> (e-f, colors) with sea surface 1036 1037 height (contours in m, contour interval 0.02m) for eddy E1 (63°-66°E, 15°-18°N) for the 1038 month of December during the two CM2.6 model years 197 and 198. 1039

Anand 1/7/2016 11:36 Deleted: overlaid

Deleted:

60.8°E 61.2°E 61.6\*

Diffusion (mol m −2

60.8°E 61.2°E Surface Chl (mol/kg

61 6\*F





1044Figure 15. (a), (b) : CM2.6 (miniBLING)  $PO_{d}$  on 9 November and 28 December of year1045197.(56°-66° E, 15° - 26° N).

