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Dominant biological control over upwelling on pCO_2 in sea east of Sri Lanka

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Abstract

Upwelling enhances pCO_2 levels due to injection of carbon rich water to the surface de-spite the removal of carbon due to increase in primary production supported by enhanced nutrients. It is hypothesized that in the Bay of Bengal, upwelling may decrease pCO_2 due to existence of low saline and pCO_2 poor waters in the subsurface layer. In order to test this hypothesis, a high resolution state of the art ocean biogeochemical model (Regional Ocean Modeling System; ROMS) runs are examined at the Sea East of Sri Lanka (SESL) where intense upwelling occurs during summer monsoon (May to August, SM). Upwelling enhances pCO_2 by 34 μatm whereas decrease in surface temperature and increase in surface salinity reduce pCO_2 by 24 μatm. The estimated net effect of upwelling is an increase in pCO_2 by 10 μatm. In contrast, soft and hard tissues together contribute to a decrease in pCO_2 by 21 μatm suggesting that the biological effect dominates over upwelling, resulting in a net decrease of pCO_2 by 11 μatm in the SESL. This striking contrast between the increase in pCO_2 due to physical dynamics (upwelling) and the removal of pCO_2 due to biological processes is caused by shallow (deep) nitracline (DIC-cline) in the SESL.

Key Points:

- Phytoplankton bloom, in response to upwelling, decreases pCO_2 in sea east of Sri Lanka (SESL) during summer monsoon.
- Relatively shallow nitracline and deep DIC-cline is observed in the SESL.
- Removal of pCO_2 due to biological production dominates increase of pCO_2 due to up-welling in SESL.

1. Introduction

The sea east of Sri Lanka (SESL) is known for its unique physical and biological char- acteristics during the summer monsoon (June-September, SM) [*Vinayachandran et al.,* 2004; *Le´vy et al.,* 2007; *De Vos et al.,* 2013]. The satellite images during May indicate high surface chlorophyll concentrations in the southern coast of Sri Lanka and Indian Peninsula. High phytoplankton biomass spreads all around Sri Lanka coast during June which eventually develops as a bloom in July. The SESL also witnesses intense phytoplankton bloom during summer as a response to the nutrient input associated with the Sri Lanka Dome (SLD) and also due to the augmented nutrients from the coast of Sri Lanka and Peninsular India by the intruding summer monsoon current (SMC; *Vinayachandran et al.* [2004]).

The nutrients for the phytoplankton bloom in the SESL are supplied by the following mechanisms: (a) a part of nutrients that upwell at the southern coast of Sri Lanka intrudes into the east via the SMC, (b) the open ocean upwelling associated with the SLD nourishes the SESL, and (c) the nutrients upwelling south of Indian Peninsula are advected to the SESL by the intruding SMC [*Vinayachandran et al.,* 2004]. The enhancement of nutrients by the intruding SMC is thus an important factor promoting the biological bloom in the SESL. This mechanism supplements the demand of nutrients even in cases where meso-scale eddies are suppressing the upwelling and the SLD [*Jyothibabu et al.,* 2015].

The sediment traps mooblack close to the southern part of the SESL show that the blooms during June to September are dominated by diatoms with an implication of high biogenic export (~3.6 gC m^{-2} month $^{-1}$) [*Vidya and Kumar,* 2013]. Numerical studies on biogenic export indicate that the enhanced primary production blackuces the surface ocean pCO_2 of this region by 10 μatm and the corresponding air-sea CO_2 fluxes by 0.1-0.2 gC m^{-2} $month^{-1}$ [*Sharada et al.,* 2008]. Observational and numerical simulations have consistently shown the significance of bloom on surface ocean pCO_2 of SESL [*Le´vy et al.,* 2007; *Sharada et al.,* 2008; *Kone´ et al.,* 2009]. pCO_2 is the key oceanic variable that determines the net exchange of CO_2 across the ocean-atmosphere interface. However, a precise quantification of processes responsible for the variability of pCO_2 is still elusive.

Although a large number of studies examined the bloom characteristics of SESL, no significant effort has been made in quantifying the role of bloom in the seasonality of oceanic pCO_2 of SESL [*Valsala and Maksyutov,* 2013]. The SESL is an important region with respect to the carbon cycle of Bay of Bengal because it is one of the regions where the largest seasonal and interannual variability

of surface ocean pCO_2 and air-sea CO_2 fluxes of Indian Ocean are found [*Valsala and Maksyutov*, 2013; *Takahashi et al.*, 2009]. Moreover, as evident from long-term observations, the presence of large phytoplankton biomass and vertical opal and organic carbon flux in the SESL warrants a detailed study of carbon cycle variability in this region, in order to assess the role of bloom in the carbon sink [*Vidya and Kumar*, 2013; *Jyothibabu et al.*, 2015]. Hence, in this study we have examined the controlling factors of surface ocean pCO_2 and air-sea CO_2 flux variability of SESL during SM.

The significance of assessing the role of bloom in pCO_2/CO_2 fluxes of SESL can be conceived as follows. The compilation of various model results and synthesized estimates clearly show that the northern Indian Ocean is a net source of CO_2 to the atmosphere with an exception of Bay of Bengal which is a mild sink in which ocean biology plays an integral role [*Levy et al.*, 2007; *Sharada et al.*, 2008; *Sarma et al.*, 2012, 2013; *Valsala and Maksyutov*, 2013]. This is further confirmed by high volume export of organic carbon to deep waters in the SESL region [*Vidya and Kumar*, 2013]. This is unique compablack to the rest of the Bay of Bengal which is relatively less productive in the summer season mostly due to the lack of supply of nutrients across the stratified upper ocean and shallow photic depth due to high suspended load [*Prasanna Kumar et al.*, 2002; *Madhupratap et al.*, 2003]. In contrast the eddy mediated biological pump is evident during most of the year [*Sabu et al.*, 2015; *Sarma et al.*, 2016]. SESL, on the other hand, experiences phytoplankton blooms owing to the SLD formation and SMC intrusion. Therefore assessing the oceanic pCO_2 variability of this highly biogenic export region is of prime importance to the carbon cycle of the Bay of Bengal.

There has been no routine monitoring of surface ocean pCO_2 of the Bay of Bengal except for a few research based cruises focusing on the coastal areas of the Bay [*Sarma et al.*, 2012, 2015]. A data based assessment of carbon cycle of the upper ocean Bay of Bengal is thus not feasible and that forces studies to resort on biogeochemical models [*Sarma et al.*, 2013]. Therefore, by using a high resolution physical-biogeochemical coupled model, we have investigated pCO_2 variability and its controlling mechanisms in the SESL region.

2. Data and Methods

We have used SeaWiFS (Sea-viewing Wide Field-of-view Sensor) satellite's 8-day averaged chlorophyll-a (Chl-a) data product with approximately 9 km spatial resolution from NOAA (National Oceanic and Atmospheric Administration) Coast Watch data portal. Further details for SeaWiFS and Aqua missions can be accessed through the Ocean Color portal managed by NASA's Goddard Space

Flight Center (GSFC). These data are normalized to match spatial resolution of the model for comparison.

A Bio-Argo profiling float (WMO No. 2902086) equipped with temperature and salinity (SBE41CP), dissolved oxygen (AanderraOptode 4330), and chlorophyll fluorescence (WET Labs FLBB), deployed in the central Bay of Bengal, at the end of December 2012, near 12.2° N, 88.7°E is used for assessment of the model simulated Chl-a variability in the subsurface waters of the Bay of Bengal. The daily vertical profiles of Chl-a were recorded by the float from 30-December-2012 to 17-January-2013. Subsequently the mission cycle was changed to 5 days. These floats were manufactublack by Sea-Bird Electronics and used in the Indian ARGO project managed by the Indian National Centre for Ocean Information Services (INCOIS). The floats acquire data during their ascent at pblacketermined depths. In the present study, Bio-Argo profiles in the upper 100 m of the water column between 01-January-2013 and 31-December-2015 are used. The data quality of the profiles is checked following *Takeshita et al.* [2013]. *Takahashi et al.* [2009] prepablack monthly climatological maps of pCO_2 (Takahashi climatology) using the available global oceanic pCO_2 observations. These maps are used to assess the capability of the model simulated surface ocean pCO_2.

2.1 Model description

A fully coupled high resolution physical-biogeochemical model has been successfully configured for the entire Indian Ocean basin (30°S to 30°N; 30°E to 120°E) using Regional Ocean Modeling System (ROMS) version 3.7 [*Haidvogel et al.*, 2008].

The model has a horizontal grid resolution of 1/12°(approximately 9 km), and uses 40 vertical layers in a terrain-following s-coordinate system. In the physical part of the model setup, the narrow channel between India and Sri Lanka is kept as closed and isolated land mask points were removed. The shallow Mallacca strait and the part of the South China Sea are also removed from the model grid. On the western and northern continental land boundaries, no slip and no normal flow boundary conditions are applied. The lateral boundaries in the east and south are kept open. It is challenging to configure global models with sufficiently fine horizontal and vertical resolutions due to computational limitations. A global model with coarse resolution can't capture all fine scale variability in the regional oceans. Also, the inherent inaccuracies in initial and boundary conditions can cause issues such as numerical drifting in model simulations. Therefore, the one-way nested modeling approach is adopted in which a global ocean general circulation model provides boundary

data to a regional model that is further coupled to an ecosystem model [*Shulman et al.,* 2003]. This approach utilizes strong relaxation at the open boundaries, but weaker relaxation in shallow waters and productive zones to allow the model to respond more freely to surface atmospheric forcing. The K-Profile Parameterization (KPP) scheme of *Large et al.* [1994] is used to represent vertical mixing. Bi harmonic viscosity and diffusion schemes are chosen for horizontal mixing [*Griffies and Hall berg*, 2000].

The tracer fields (temperature and salinity), sea surface height and the components of momentum are initialized on 01-January-2003 with the state of the ocean as available in the Global Ocean Data Assimilation System (GODAS), configublack at the INCOIS [*Ravichandran et al.,* 2013]. The weekly averaged boundary on interannual time scale derived from the GODAS was prescribed for the tracers, sea surface height and momentum fields along the east and south lateral boundary. The National Centre for Medium Range Weather Forecast (NCMRWF, New Delhi, India) has developed a data assimilated Global Forecast Systems (GFS) at a horizontal resolution of 0.25°which provides real time forecasts of forcing fields to force ocean models configublack at the INCOIS [*Rajagopal et al.*, 2012]. Six-hourly analysed atmospheric forcing fields obtained from the NCMRWF are used to force the model set-ups during 01-January-2003 to 31-December-2017. Atmospheric forcing component includes air pressure and temperature at 2 m, net shortwave and longwave flux, rain fall rate and wind velocity. Surface heat and momentum fluxes are internally calculated by ROMS using the bulk parameterizations of *Liu et al.* [1979] and *Fairall et al.* [1996a,b].

The biological component of our coupled set-up consists of the nitrogen cycle model with parameterized sediment denitrification described by *Fennel et al.* [2006]. The nitrogen cyclemodel includes seven state variables viz. phytoplankton, zooplankton, nitrate, ammonium, large and small detritus classes with nitrogen concentration and phytoplankton chlorophyll. The time rate of change of concentration of each state variable describes the balance of advection-diffusion and source-sink terms among the related state variables of the nitrogen cycle [*Fennel et al.,* 2006]. The biological model also resolves the full carbon cycle. The model carbonate chemistry is described in the coupled set-up following *Zeebe and Wolf-Gladrow*[2001] and *Fennel et al.* [2008]. The full carbon cycle is represented in the model using four state variables viz. alkalinity, dissolved inorganic carbon, large and small detritus class with carbon concentration. The dynamics of dissolved inorganic carbon includes the primary production and respiration as sinking and source processes respectively following blackfield stoichiometry besides gas exchange at the air-sea interface. The biogeochemical processes, such as calcite formation and dissolution, nitrate uptake and regeneration, and

sulfateblackuction are represented in the dynamics of alkalinity. The air-sea gas exchange of CO_2 doesn't have any impact on the dynamics of alkalinity [*Millero et al.*, 1998]. The partial pressure of carbon dioxide (pCO_2)is calculated in the surface layer as described in the *Fennel et al.* [2008]. Oxygen is included as model tracer and biogeochemical dynamics of oxygen is described in the model following *Fennel et al.* [2013].

The biological variables are initialized using the climatological state of January generated during climatological run of the coupled set-up. Considering the unavailability of sufficient observations along the boundaries, the monthly climatological boundary derived from the NODC World Ocean Atlas 2013 is prescribed for the nitrate, oxygen, dissolved inorganic carbon and alkalinity whereas small uniform values along the east and south lateral boundary are prescribed for the remaining state variables. Daily averaged output has been averaged from 01-Jan-2003 to 31-Dec-2017 for the analysis.

3 Results and Discussion

The evolution of surface Chl-a concentration surrounding the Sri Lanka and SESL are faithfully reproduced in the model (Figure 4 top). During March, no significant upwelling is observed in the sea surrounding Sri Lanka resulting in low Chl-a concentrations. By early May, winds change the direction and the southern coast of Sri Lanka experiences upwelling [*Vos et al.*, 2014] and subsequent enhancement in the Chl-a concentrations is observed from the satellite imageries. The model faithfully captures this signal. Figure 4 (top) compares the mean satellite Chl-a concentration for the months of March, May and July with that of the model. The comparison shows reasonably good agreement except a slight difference in amplitude. The contours of Chl-a surrounding Sri Lanka and its intruding shape to the SESL is captured in the model fairly well. The upwelling sets in the southern peninsula of India by May-June and propagates northward along the eastern Arabian Sea shelf during SM. The evolution of model simulated upwelling and phytoplankton bloom over its southern shelf are in good agreement with the observations reported by *Gupta et al.* [2016].

A similar comparison of model Chl-a concentrations with that derived from a Bio-Argo float (WMO No. 2902086) deployed in the south central Bay of Bengal further evaluates the capacity of the model to reproduce the observed biological variability in the region (Figure 2). Although the float didn't traverse strictly through the SESL region, a comparison to the nearest available Bio-Argo float increases the confidence in our model performance. A subsurface Chl-a maxima at about 40-60 m and its spatial variability from January-2013 to December 2015 is well captured in the model (Figure 2). To facilitate the analysis, the model outputs are extracted at the time and location (nearest grid

point) of Argo profiles. The undulations in the simulated Chl-a spatial variability is well comparable with the observations albeit the magnitude is over estimated in the model.

A similar attempt to compare the model surface ocean pCO_2 of Bay of Bengal with the observations [*Takahashi et al.*, 2009] suggest that except for a systematic bias of 10 μatmthroughout the basin from 5°N to 16°N, the overall gradient of pCO_2 from head Bay to 5°N is reproduced well in the model (Figure 3). Nevertheless, as the pCO_2 observations compiled in *Takahashi et al.* [2009] is centeblack for a year in the middle of 2000s, therefore the basin-wide bias of 10 μatm can be readily attributed to the increasing dissolved inorganic carbon concentration (DIC) that has been added in the recent times to the ocean via increased anthropogenic forcing [*Sarma et al.*, 2015].

Higher pCO_2 values are noticed during April-May as compablack to other months by 0-30 μatm (Figure 3 bottom). The onset of the phytoplankton bloom draws down the pCO_2 from June to September by 10-15 μatm. The model fairly captures observed seasonality of pCO_2 in the SESL region.

Considering the dynamics and associated biological response during summer monsoon (June to September), upwelling of subsurface waters in the SESL favours (a) decrease in pCO_2 by the enhanced biological production due to intrusion of 'nutrient rich' subsurface water via SMC, (b) vertical mixing that enhances surface DIC/pCO_2 levels due to pumping of enriched waters from below (Figure 4), (c) change in sea surface temperature (SST) associated with the SLD [*Vinayachandran et al.*, 2004] and SMC intrusion [*De Vos et al.*, 2013] that alters solubility pump of the region (Figure 5), and (d) variation in air-sea exchange [*Udayshankar*, 2015] may intricately alter the pCO_2 levels. The effect of all these processes eventually determine the net pCO_2 variability of the region.

The net pCO_2 variability of surface ocean is also controlled by abiotic factors or buffer chemistry of surface carbon and usually follows the pattern of SST especially in the region like Bay of Bengal where biological production is relatively weak [*Prasanna Kumar et al.*, 2004]. The change in pCO_2 of the SESL region gradually follows the evolution of SST in the region (Figure 4) from January to June when the biological pump is weak as evidenced from satellite Chl-a concentrations (Figure 4 top). However from July to September the variability of pCO_2 of the SESL is not in phase with that of the SST indicating decoupling of SST-pCO_2 relation by the interplay of biology and physical dynamics (upwelling) of this region. The spatial pattern of evolution of pCO_2 from March to July shows that the surface ocean pCO_2 in the SESL region has coherent 'trends' with that of the

observed satellite Chl-a concentrations by July (Figure 4). The relatively higher sinking biogenic fluxes from this region [*Vidya and Kumar*, 2013] further infers the significant role of biological pump in regulating surface ocean pCO_2. This affirms our hypothesis that the intricate relation between surface ocean dynamics and biology has to be analyzed in tandem for identifying the mechanisms responsible for the variability of pCO_2 during the bloom.

The component equation depicting the total variability of surface ocean pCO_2 as represented elsewhere [*Valsala and Maksyutov*, 2013; *Takahashi et al.*, 2009] is as follows:

$$dpCO_2/dt = \partial pCO_2/\partial DIC * dDIC/dt + \partial pCO_2/dALK + \partial pCO_2/\partial SST * dSST/dt + \partial pCO_2/ dSSS * dSSS/dt \text{ --- (1)}$$

The left term of the equation represents the total evolution of pCO_2. The terms on the right side represent the pCO_2 variability due to dynamics of surface ocean dissolved inorganic carbon (DIC, biologically linked to alkalinity), alkalinity (ALK), temperature (SST) and salinity (SSS) respectively.

In order to segregate the individual components in the total pCO_2 seasonal cycle, we did an offline component analysis. The coupled physical-biogeochemical model simulated daily averaged outputs of DIC, ALK, SST and SSS for the entire simulation period were considered for the analysis. The solubility abiotic routines of the carbon chemistry model was run offline for the entire simulation period with daily averaged surface DIC, ALK, SST and SSS and the corresponding pCO_2 is reconstructed. The reconstruction of net pCO_2 results in insignificant residual error (±2 μatm) as is evident in the RMSE analysis (Figure 6). In the subsequent offline abiotic runs, we supplied the annual mean of a given parameter at each grid and reproduced the net pCO_2. The supply of annual mean for a particular variable to run the abiotic model in offline means that we were eliminating the impact of a particular term in the seasonality of pCO_2 in the above equation at each run. Therefore the difference between reconstructed control pCO_2 and the pCO_2 of sensitivity runs gave the contribution of individual components in the net pCO_2 variability. Furthermore, we calculated the tendency of control pCO_2 per month and estimated its daily standard deviation as standard error. This procedure is also applied for individual reconstruction of pCO_2 for each variable. Four such sensitivity abiotic carbonate pump runs have been done with the supply of respective annual mean of DIC, ALK, SST and SSS each time.

The change in the total ocean biology has a significant contribution to the pCO_2 variability as well.

The total biological contribution is consideblack to be the summation of individual contribution of soft and hard tissue. In order to obtain the individual contribution of the soft and hard tissues, the methodology prescribed by *Louanchi et al.* [1996] is used. The details of the separation of soft and hard tissue pumps from the model solutions are given in Appendix.

The offline component analysis quantified the contribution of individual term in Equation (1) in the net pCO_2 variability of SESL region while the quantification of individual contribution of soft tissue and hard tissue is done following the Equations (A2 & A3) of Appendix. A summary of monthly averaged contribution of individual components to pCO_2 variability and their uncertainties are given in Table 1. The monthly contribution of soft and hard tissues are listed in Table 2. The climatology of the sensitivity runs are then contrasted with the climatology of control pCO_2 (Figure 4).

The upwelling effect influences pCO_2 in two opposite directions - (1) input of CO_2 rich subsurface waters to surface, and (2) increase in solubility of CO_2 due to injection of cool and saline waters resulting in decrease in pCO_2. These two effects counteract each other, hence, the net effect can be taken as physical effect. For instance, decrease in SST between May and August blackucespCO_2 of the SESL by 33.6 μatm whereas increase in salinity enhances pCO_2 levels by 9.5 uatm (Table 1). This is due to cooling of the sea surface and enhanced salt content by SMS intrusion of upwelled water from south of Sri Lanka as well as by the upwelling related to the SLD [*De Vos et al.*, 2013] (see Table 1 for mean and RMSE). Therefore, the net solubility effect is a blackuction of pCO_2 by 24.1 μatm. In contrast, vertical mixing caused by upwelling enhanced monthly pCO_2 by 4.1, 12.5, 13.9 and 3.7 μatm, respectively from May to August, with a total of 34.2 μatm, due to intrusion of SMC and upwellling due to SLD. Therefore, the net effect was an increase of 10 μatm due to direct input of CO_2 through vertical mixing caused by upwelling and change in pCO_2 due to solubility variations (Table 1).

The Figure 4 suggests that the effect of hard tissue is smaller than that of soft tissue during study period. The effect of soft tissue formation is small during April (0.1 μatm) and then increases to -10.1 μatm in July and eventually decreases to -7.1 μatm in August. In contrast, the hard tissue formation increases pCO_2 by <1.3 μatm during our study period. The net biological effect (both hard and soft tissue) decreases monthly pCO_2 by 1.2, 5.1, 8.8 and 5.9 μatm respectively during May-August, with a total net decrease of pCO_2 by 21 μatm (Table 2). Therefore, the net effect of upwelling is to decrease pCO_2 by 11 μatm (21 μatm decrease by biology and 10 μatm net increase by vertical processes). In spite of the net contribution of physical dynamics being close to zero, the

biological impact remains high (6 μatm) during August due to the prolonged biological response (see *Le´vy et al.* [2012] for phase lead-lag between biological and solubility pumps at sub-monthly scales). Earlier studies pointed out that the impact of mixing on pCO_2 during a cyclone is almost negligible as mixing initially increases pCO_2 whereas biological removal compensates the physical mixing over a period of month [*Levy et al.*, 2012]. When nutrients are pumped, the change in pCO_2 is controlled by both mixing (which increases pCO_2) and solubility (cooling of the sea surface decrease pCO_2 and increases in salt content increases pCO_2). The balance of these controls lead to the net pCO_2 change.In presence of sufficient nutrients, phytoplankton production dominates and decreases the pCO_2. Later, once the production weakens, respiration dominates and increases the pCO_2. The DIC, on the other hand, counteracts the impact of net pCO_2 change as suggested by *Levy et al.* [2012]. Nevertheless, the averaged monthly responses that we reported here average out the possible shorter time scale lead-lag responses of biology and solubility pumps on the net pCO_2.

A conspicuous pCO_2 variability by DIC and SST is noticeable during July. At the peak of the bloom, the augmentation effect of DIC on pCO_2 appears to be strongest among all months examined here. The impact of SST starts weakening during July while the impact of mixing strengthens. The concomitant biological impact during July indicates its role in draw down of pCO_2. The significant decrease in pCO_2 is further confirmed by significant increase in sinking carbon flux in SESL (see *Vidya and Kumar* [2013]; *Jyothibabu et al.* [2015] for the sediment trap data) that may further enhance the DIC while decreasing the ALK. Therefore, the increase in pCO_2 due to DIC (during July) and decrease in pCO_2 due to alkalinity support the role of both soft tissue and carbonate pumps, leading to dominant biological exports in the SESL region. It is worth mentioning that the sediment traps were mounted 800 m below the surface to explore the variability of biogenic carbonate fluxes whereas the dome shape of DIC is observed around 100 m. Therefore, we infer that the variations in upper ocean DIC are caused mostly by upwelling waters rather than carbonate pump which is located below the depth of upwelled waters. High resolution sediment trap measurements are requiblack to further separate the role of soft tissue and carbonate pumps on pCO_2 in the SESL. The dominance of biological effect over physical dynamics in the SESL, which is a part of the Bay of Bengal, is in contrast with the domination of physical effect observed in the Arabian Sea, equatorial Indian Ocean and south tropical Indian Ocean [*Louanchi et al.*, 1996; *Sarma et al.*, 2000]. Despite the supply of both DIC and nutrients, it is interesting to know why biological processes dominate over physical dynamics in the Bay of Bengal. In order to examine this, the depths of 2100 mM DIC-isoline (DIC-cline) and 2 mM nitrate isoline (nitracline) is plotted both in the Arabian Sea

and Bay of Bengal (Figure 8). It is noticed that depth of DIC-cline is deeper while nitracline is shallower in the Bay of Bengal than Arabian Sea due to shallow mixed layer and existence of relatively fresh water in the upper layers due to river discharge. This results in lower DIC and release of nutrients at shallower depth (below mixed layer) through decomposition of organic matter. This may result in pumping of low DIC and higher nitrate to the mixed layer in response to upwelling in the Bay of Bengal comparable to Arabian Sea, which leads to the dominance of biological production over mixing in the former basin. Our study suggests that in CO_2 dynamics, biological processes dominate over physical processes in the SESL and lead to a net decrease of 11 μatm in pCO_2 during the summer monsoon.

4 Summary and Conclusions

Based on state of art high resolution ocean biogeochemical model (ROMS), the variability in carbon cycle in one of the prominent phytoplankton bloom areas of the Bay of Bengal in the Indian Ocean is studied. Focusing on the variability of pCO_2, as a key variable depicting the air-sea equilibrium of CO_2 in this region, for the first time, we have demonstrated how biology dominates over physical dynamics in the SESL region. Over global ocean, typically upwelling and mixing effects dominantly control pCO_2 over other effects, including biology, as mixing brings CO_2 rich waters to surface. In contrast, it is noticed that higher nutrients are brought to the surface comparable to DIC due to low saline waters in the upper 100 m resulting from river discharge. This leads to deep DIC-cline, as fresh water contains lower DIC than seawater, and shallow nitracline comparable to the Arabian Sea and other regions in the Indian Ocean.

Several cyclonic eddies and seasonal cyclones/depressions (summer and post monsoon) form in the Bay of Bengal and they are expected to churn the upper ocean resulting in pumping of DIC and nutrients rich waters to surface. It has been hypothesized in earlier studies that the net effect of tropical cyclones, due to mixing and biological response, on surface pCO_2 may be balanced over time, if phase lag is considered. Our study, in contrast, suggests that such balance may be unlikely due to less DIC input comparable to nutrients resulting in dominant biological removal of atmospheric CO_2. However, more work is required to confirm our hypothesis on modulation of carbon cycling due to physical dynamics in the Bay of Bengal vis-a-vis Arabian Sea.

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Table 1.

Monthly pCO_2 tendency due to variability in individual components according to Equation (1). CTRL indicates the net pCO_2 tendency in the control reconstruction of pCO_2 by the abiotic pump runs. Surface ocean dissolved inorganic carbon (DIC; mM m^{-3}), alkalinity (ALK, mEq m^{-3}), temperature (SST; $^{\circ}C$) and salinity (SSS; PSU) represent the contribution by individual components in the pCO_2 variability. The error shows the uncertainty in the mean of respective months.

(CTRL - individual components) gives the net impact of each variable. Units are in µatm.

Month	CTRL	DIC	SST	ALK	SSS
MAY	-5.55 ± 0.88	-9.68 ± 3.42	4.73 ± 0.49	-2.99 ± 1.86	-8.43 ± 0.91
JUNE	-8.33 ± 0.43	-20.84 ± 3.16	7.69 ± 0.51	-1.62 ± 2.07	-10.47 ± 0.44
JULY	-1.94 ± 0.38	-15.91 ± 1.94	6.24 ± 0.83	7.73 ± 1.62	-4.48 ± 0.38
AUGUST	0.03 ± 0.67	-3.68 ± 1.77	-0.79 ± 1.24	5.66 ± 1.68	-1.95 ± 0.66
Sum of CTRL - Sensitivity	0	34.32 ± 10.29	-33.66 ± 3.07	-24.58 ± 7.23	9.54 ± 2.39

Table 2. Monthly individual contribution of soft and hard tissues on pCO_2 following the Equations (2&3) of *Louanchi et al.* [1996]

Month	SOFT TISSUE	HARD TISSUE
MAY	-0.84 ± 0.10	-0.34 ± 0.01
JUNE	-5.44 ± 0.30	0.30 ± 0.05
JULY	-10.16 ± 0.51	1.37 ± 0.08
AUGUST	-7.06 ± 0.46	1.16 ± 0.07
Total	-23.5 ± 1.37	2.49 ± 0.21

Figure 1: (top panel) The model derived Chl-a (shade; mg m−3) and corresponding satellite observations (contour; mg m−3), (bottom panel) Model simulated monthly mean pCO2 (µatm) for March, May and July in the SESL. The box (830E-880E, 60N-100N) represents the intruding shape of Chl-a in the SESL.

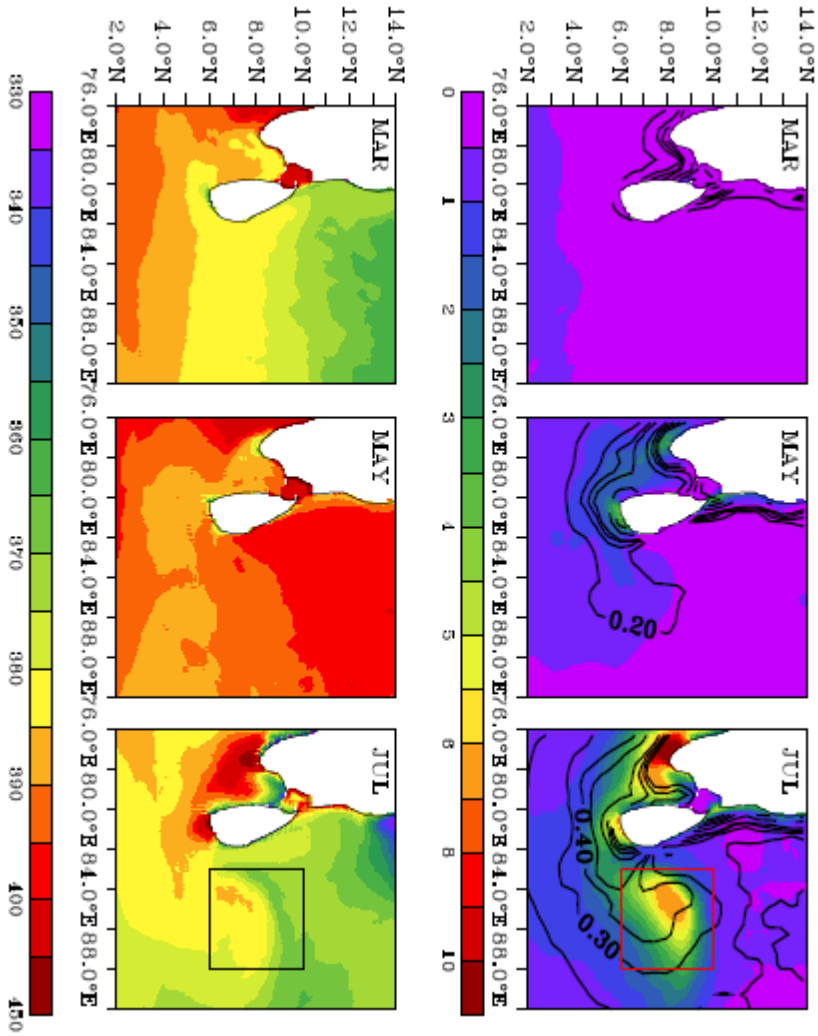


Figure 2. Comparison between model (bottom panel) simulated Chlorophyll-a (mg m−3) concentrations against observations (top panel) made by Bio-Argo float (WMO No. 2902086) deployed in the south-central Bay of Bengal (12.2°N, 88.7°E).

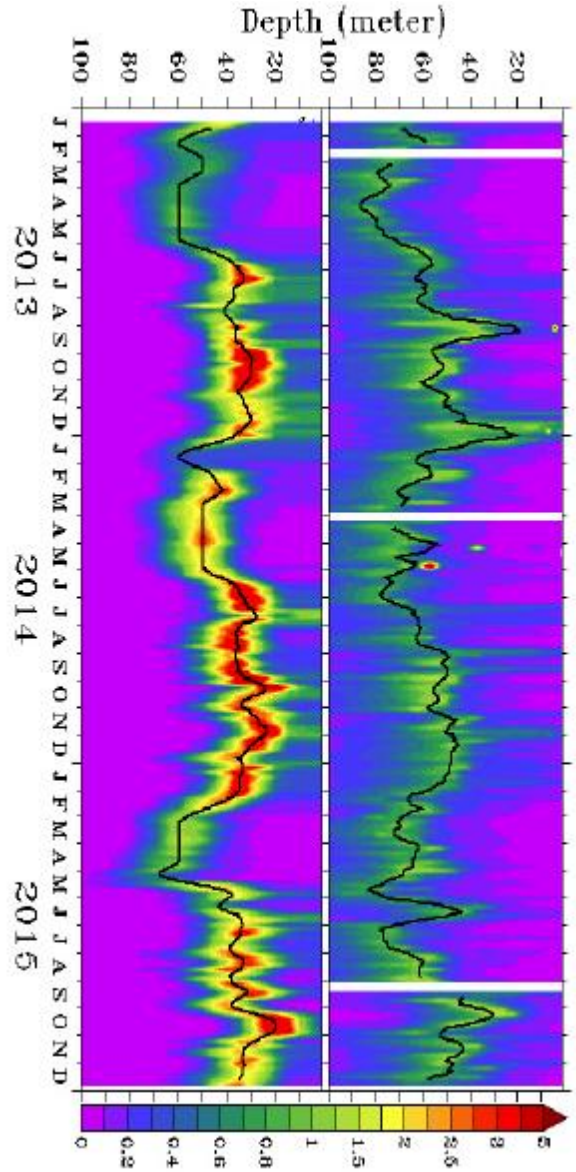


Figure 3. (top panel) Comparison of model simulated surface ocean pCO2 (µatm) with Takahashi et al. [2009] climatology at 900E. (bottom panel) Comparison of the same but averaged over the box (83°E-88°E and 6°N-10°N) in the SESL.

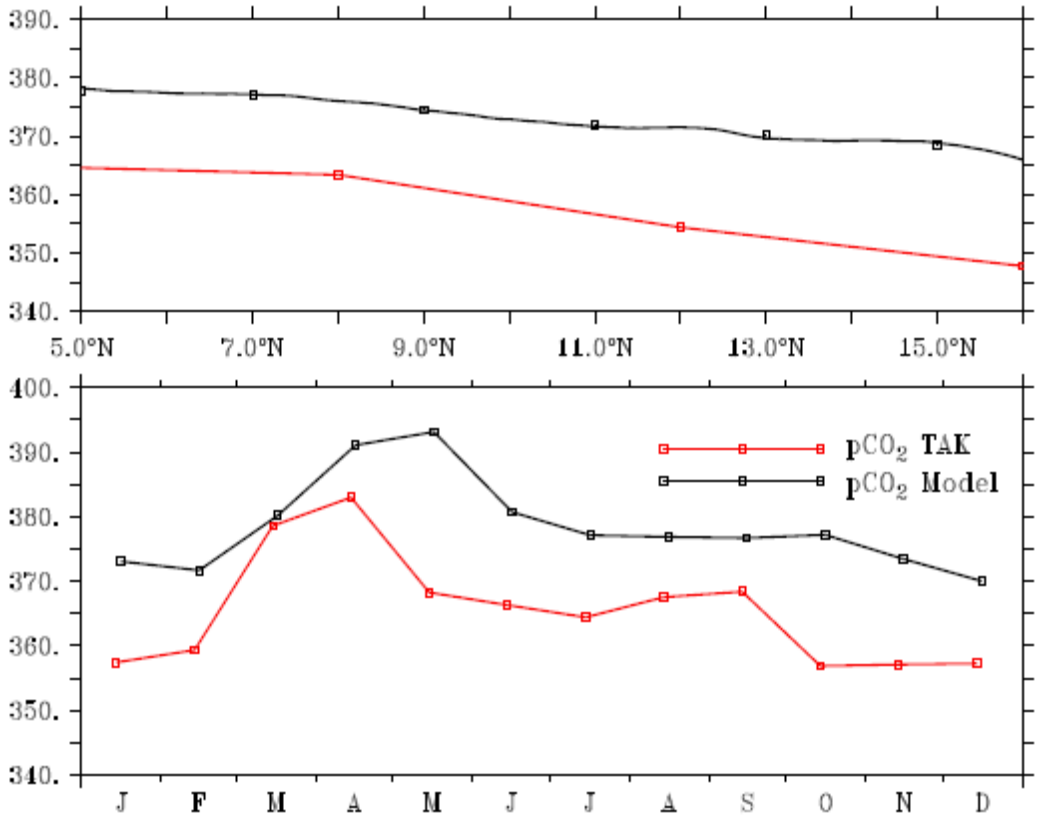


Figure 4. Model simulated monthly mean of the vertical section of DIC (mM m^{-3}) at 70N for March, May and July in the SESL.

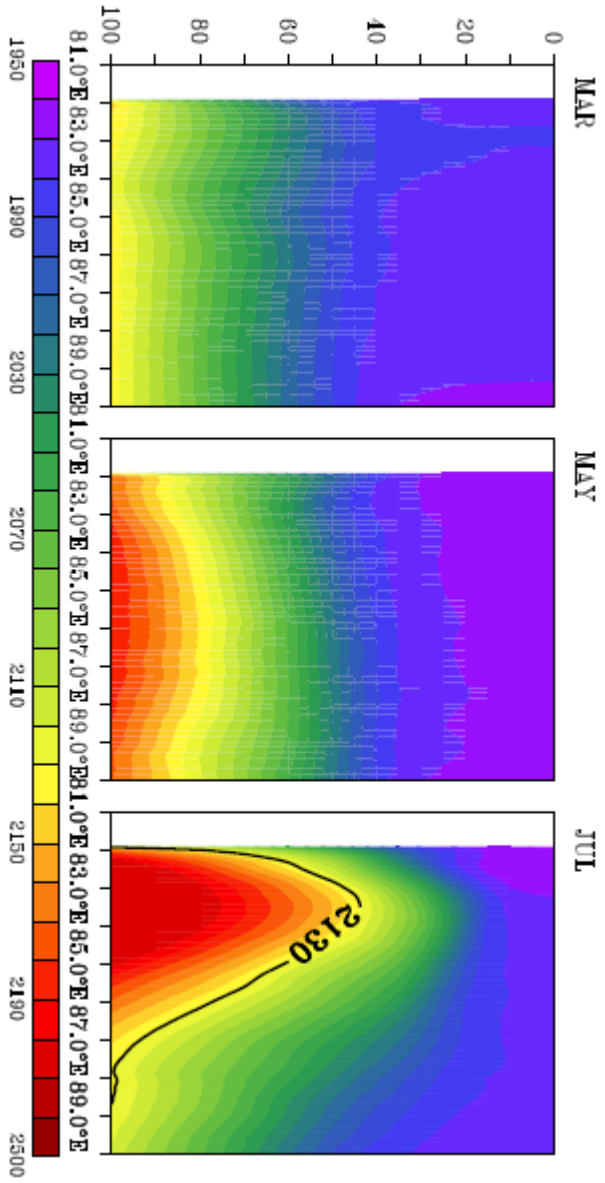


Figure 5. Seasonal cycle of model derived pCO2 (right axis) and sea surface temperature (left axis) averaged over the box (83°E-88°E and 6°N-10°N) in the SESL.

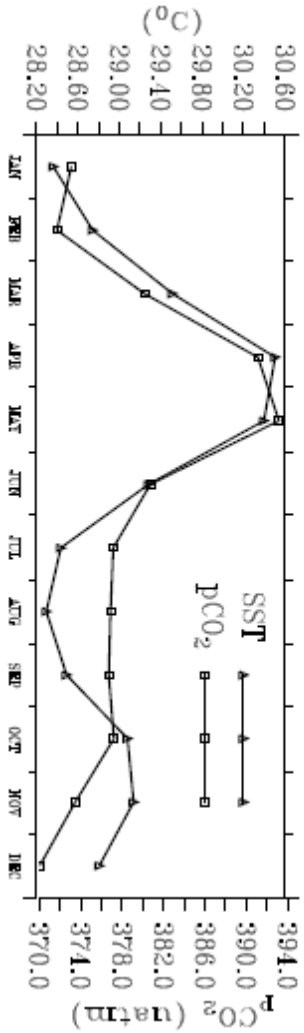


Figure 6. RMSE of the net pCO2 after reconstructing it by offline abiotic pump model. RMSE is calculated from the difference of reconstructed pCO2 and pCO2 of the original model simulation for the SESL region. RMSE is calculated from daily data of entire period of model simulation.

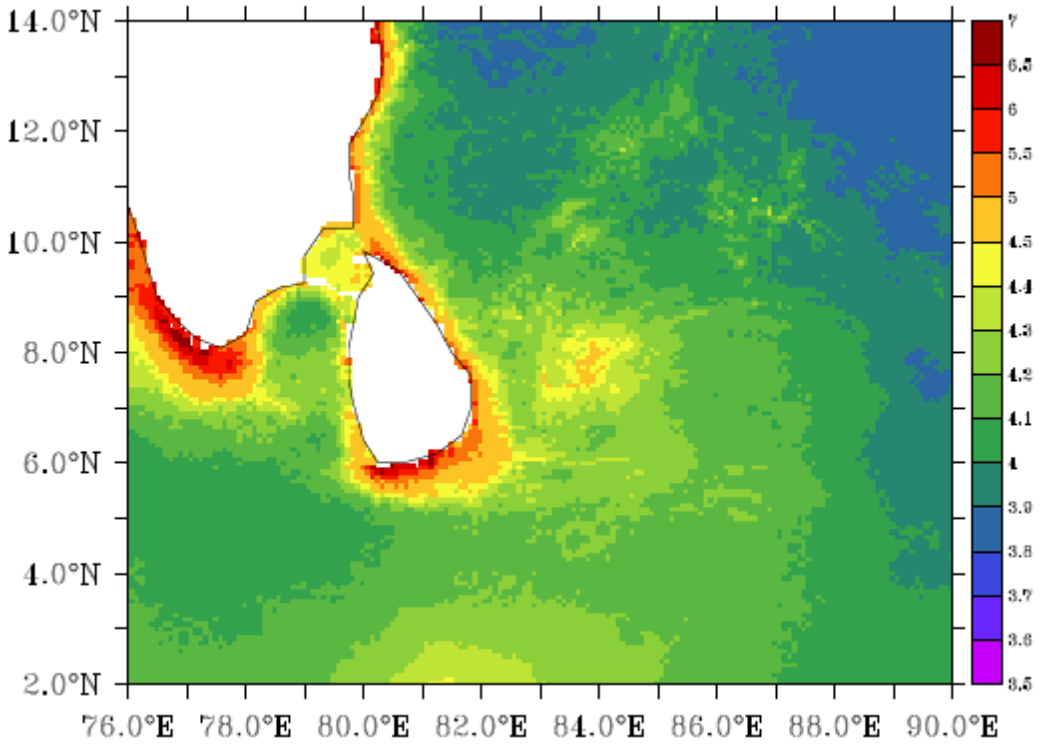


Figure 7. The contribution of the individual components (DIC (vertical mixing), SST, SSS and total biology (soft tissue and hard tissue)) on net pCO2 variability during March to August in the SESL region. Black dotted line represents the control run, black line represents the contribution of DIC (vertical mixing), Green line represents the contribution of SST, Cyan line represents the contribution of SSS, Black line represents the contribution of biological soft tissue and Magenta line represents the contribution of biological hard tissue.

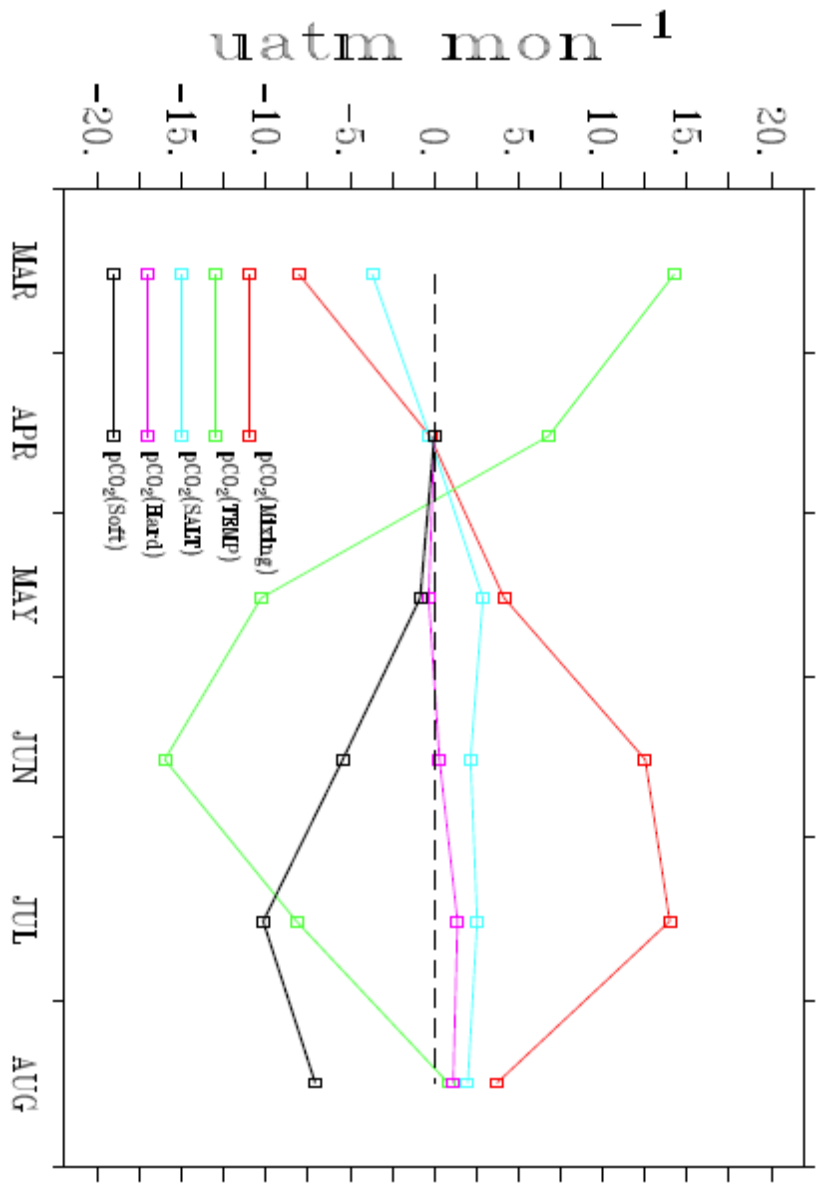
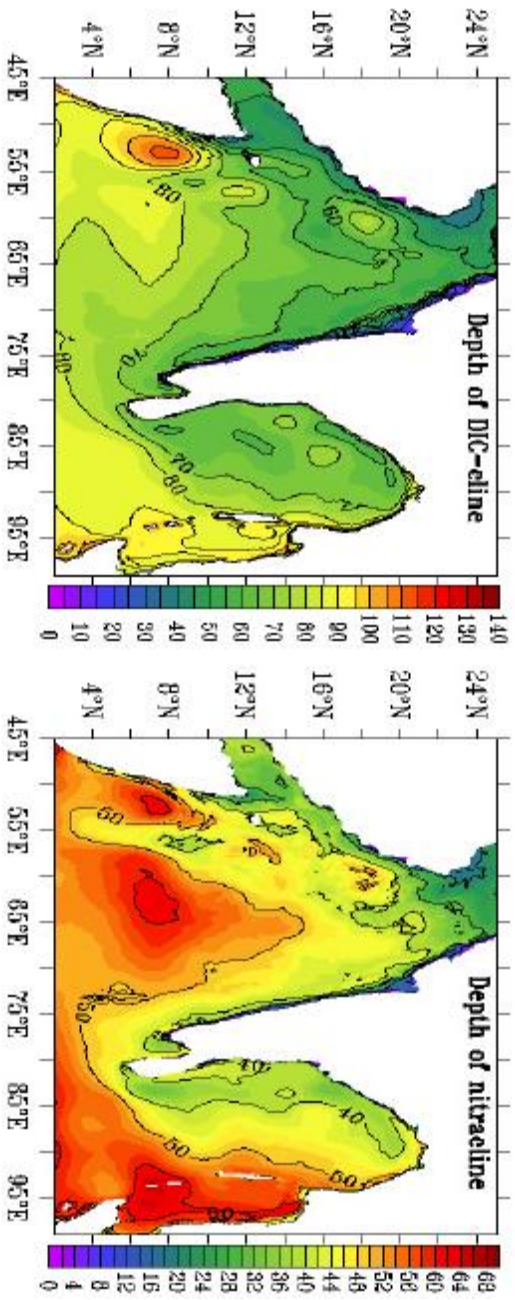


Figure 8. Annual mean depth of 2100 mM DIC-isoline (DIC-cline) and 2 mM nitrate isoline (nitracline) in the northern Indian Ocean.



5 Appendix

The change in pCO_2 due to biology is derived from DIC and ALK variations linked to the biomass production and the calcite formation in the euphotic layer and taken from *Lovanchi et al.* [1996]. These variations are expressed using the following equations:

$$\left(\frac{\partial DIC}{\partial t}\right)_{biology} = -\left(\frac{\partial N}{\partial t}\right)_{biology} * RCN - S_{biology} \quad (2)$$

$$\left(\frac{\partial ALK}{\partial t}\right)_{biology} = -2 * S_{biology} \quad (3)$$

The first term of the right hand side of the Eq. (2) represents the variation in nutrients with time resulting from photosynthesis, mortality and respiration, converted in terms of carbon unit from the C:N blackfield ratio RCN (blackfield et al., 1963). The rate of change in nutrients is computed at each time-step according to the following equation:

$$\left(\frac{\partial N}{\partial t}\right)_{biology} = -(u - r * q) * Chl * RN \quad (4)$$

where u is the growth rate which depends on the maximal growth rate u_m and on the single nutrient following a Michaelis- Menten formulation $[u = u_m * N/(N + K_m)]$, where K_m is the half-saturation constant, r is the removed factor and q is the the respiraton, mortality and grazing rate. u and q are calculated at each time-step using the following equation:

$$\left(\frac{\partial Chl}{\partial t}\right)_{biology} = (u - q) * Chl \quad (5)$$

where $\partial Chl/\partial t$ is the rate of change in chlorophyll. RN is the N:Chl blackfield ratio. The second term of the right-hand side of the Eq. (2) represents the temporal variation of the global calcite concentration which is expressed as 20% of the total exported production (P_{export}).

$$S_{biology} = 0.2 * (\partial P_{export}/\partial t) \quad (6)$$

Again, the exported production P_{export} consists of two parts of inorganic and organic carbon and is computed in terms of carbon at each time step by:

$$(\partial P_{export}/\partial t) = (1 - r) * q * Chl * RCC \quad (7)$$

where RCC is the C:Chl blackfield ratio.