

Delineating environmental control of phytoplankton biomass and phenology in the Southern Ocean

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Main point #1: Phytoplankton phenology and biomass are mostly organized in the Southern Ocean at a large latitudinal scale and a regional scale.

Main point #2: The timing of bloom occurrence appears tightly linked to the seasonal cycle in irradiance, with some exceptions in specific light-limited regimes.

Main point #3: Zonal asymmetries in regional-scale phytoplankton biomass are mainly driven by local advective and iron supply processes.

Running head (40 characters max): Phytoplankton dynamics in the SO

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Abstract

The Southern Ocean (SO), an area highly sensitive to climate change, is currently experiencing rapid warming and freshening. Such drastic physical changes might significantly alter the SO's biological pump. For more accurate predictions of the possible evolution of this pump, a better understanding of the environmental factors controlling SO phytoplankton dynamics is needed. Here we present a satellite-based study deciphering the complex environmental control of phytoplankton biomass (PB) and phenology (PH; timing and magnitude of phytoplankton blooms) in the SO. We reveal that PH and PB are mostly organized in SO at two scales: a large latitudinal scale, and a regional scale. Latitudinally, a clear gradient in the timing of bloom occurrence appears tightly linked to the seasonal cycle in irradiance, with some exceptions in specific light-limited regimes (i.e. well-mixed areas). Superimposed on this latitudinal scale, zonal asymmetries, up to three orders of magnitude, in regional-scale PB are mainly driven by local advective and iron supply processes. These findings provide a global understanding of PB and PH in SO, which is of fundamental interest for identifying and explaining ongoing changes as well as predicting future changes in the SO biological pump.

1. Introduction

The Southern Ocean (SO) plays a key role in the global carbon cycle, absorbing 40% of the total oceanic inventory of anthropogenic CO₂ [Khaliwala *et al.*, 2009]. However, current and predicted changes in ocean circulation and hydrology associated with climate variability might impact the efficiency of the SO carbon sink, although the extent of such changes and their specific impact on the biological pump are still vigorously debated [Landschützer *et al.*, 2015; Le Quéré *et al.*, 2010; Munro *et al.*, 2015]. Such debate cannot advance without a

better understanding of the factors that shape the complex biogeography of the SO, and what controls their time-scale of variability (i.e. daily, seasonal to decadal).

Delimiting marine bio-regions has proven highly valuable for disentangling multiple limiting growth factors affecting the efficiency of the biological pump [Longhurst, 2007] mediated by both PB and PH, and for ultimately underlining effects of climate change such as areal changes in trophic regime [Polovina *et al.*, 2008] or in fisheries stock and species [Fossheim *et al.*, 2015]. Changes in PB profoundly alter the efficiency of the biological pump, thereby modulating energy transfer to upper trophic levels and carbon export to deep water [Legendre and Rassoulzadegan, 1995]. Changes in PH may also have important consequences for the marine food web, due to potential mismatch between primary and secondary producers and apex predators [Ardyna *et al.*, 2014; Edwards and Richardson, 2004], and for the carbon cycle through export of unexploited carbon aggregates to the deep ocean [Dall'Olmo *et al.*, 2016].

The efficiency of the SO's biological pump is strongly associated with phytoplankton biomass (PB) and its phenology (PH), which themselves are mediated by nutrient availability (e.g. mainly iron, and/or possibly nitrate and silicic acid), light, temperature and mortality factors [Behrenfeld and Boss, 2014; Boyd, 2002]. These factors are essentially controlled by vertical mixing, advection, seasonal fluctuations in solar irradiance, and the extent and typology of sea-ice cover. As a result, a variety of patterns of net primary production (NPP) are found in the SO [Arrigo *et al.*, 2008], with their own associated phytoplankton species succession, stock, and distribution [Quéguiner, 2013]. Delineating the interplay of environmental forcing on PB and PH remains challenging and requires novel approaches, which are available through satellite-derived observations. Using these integrative and multidisciplinary approaches (i.e. satellite-derived ocean color and altimetry, Argo floats,

lagrangian modeling and biogeographic-derived analysis), we provide here a global understanding of the phytoplankton biomass and phenology in a potentially changing SO.

2. Material and methods

2.1 Satellite-derived, climatological and modeling products

Satellite-derived Level-3 data sets of chlorophyll-a concentration (chl *a*; mg m⁻³) and photosynthetically available radiation (PAR; E m⁻² d⁻¹) were obtained from the European Space Agency's GlobColour project (<http://www.globcolour.info>). Eight-day composite chl *a* concentrations using standard Case 1 water algorithms were used (i.e. OC4v5 for SeaWiFS, OC4Me for MERIS and OC3v5 for MODIS/VIIRS sensors; see *O'Reilly et al.* [2000] and *Maritorena et al.* [2010] for details). The climatological annual net primary production (NPP) was derived from the adapted-SO NPP model of *Arrigo et al.* [2008]. A detailed description of the NPP model can be found in the supporting information [*Dobson and Smith*, 1988; *Gregg and Carder*, 1990; *Markus*, 1999]. The climatologies of mixed layer depth and bathymetry were extracted respectively from *Pellichero et al.* [2016] and the GEBCO, version 2014 (General Bathymetric Chart of the Oceans; <http://www.gebco.net>). The climatology of the length of the sea-ice cover was derived from the Special Sensor Microwave Imager (SSM/I; 1998–2002) and the Advanced Microwave Scanning Radiometer - Earth Observing System (AMSR-E; 2002–2014) sensors, and made available by National Snow and Ice Data Center (NSIDC; <https://nsidc.org>). The locations of the major fronts (i.e. the Subantarctic Front SAF, the Antarctic Polar Front APF, the Southern Antarctic Circumpolar Current (ACC) Front SACCf and the Southern Boundary of the ACC SBdy, following *Swart et al.* [2010]) were determined based on the Maps of Absolute Dynamic Topography (MADT) product from CLS/AVISO. The Lagrangian model of horizontal

dispersion of iron fluxes used also altimeter products produced by CLS/AVISO, with support from the CNES (Centre National d'Etudes Spatiales; <http://www.aviso.altimetry.fr/duacs/>).

2.2. Clustering K-means method

The bio-regions were defined here using a cluster K-means analysis (see the SI for more details), previously applied successfully in the Mediterranean Sea [D'Ortenzio and Ribera d'Alcalà, 2009; Mayot *et al.*, 2016], in the North Atlantic [Lacour *et al.*, 2015] and at the global scale [D'Ortenzio *et al.*, 2012]. The analysis was performed on climatological and normalized annual chl *a* cycle, in order to statistically organize the GLOBcolour time series (1998–2014) and to create clusters representing regions of similarity (i.e. annual chl *a* cycles). To maintain consistency in the seasonal data availability throughout the study area, the period study covers the period from September to March (as referred here to “annual”).

This technique regroups pixels with seasonal cycles shapes - i.e. similar phenologies. A single characteristic seasonal cycle, that is statistically representative of the group as a whole, is then determined by calculating the center (average cycle) within each group or cluster. Each group representative of a characteristic seasonal cycle constitutes a phenological characteristic regime. A more detailed discussion of the cluster K-means analysis can be found in the supporting information [Devred *et al.*, 2007; Hartigan and Wong, 1979; IOCCG, 2009; Lund and Li, 2009; Milligan and Cooper, 1985]

A one-way analysis of variance by ranks (Kruskal–Wallis H test; Zar [2010]) was performed to test whether eco-regions differed in their biological characteristics (i.e. annual mean chlorophyll *a* concentration, seasonality and timing of bloom). A significant result of the Kruskal–Wallis H test implies that at least one eco-region differs from all others.

2.3 Lagrangian modeling of horizontal iron fluxes

In order to estimate iron delivery due to horizontal stirring, we use an advection scheme based on altimetry, extending to the Southern Ocean a Lagrangian model used for predicting

the development of the Kerguelen phytoplanktonic plume [d'Ovidio *et al.*, 2015]. The model has been extensively calibrated and validated in the Crozet and Kerguelen regions by integrating satellite data (altimetry and ocean color), lithogenic isotopes, iron measurements, and drifters [d'Ovidio *et al.*, 2015; Sanial *et al.*, 2014; Sanial *et al.*, 2015]. The main characteristics of the model were resumed here and we refer to d'Ovidio *et al.* [2015] for the full description of the model. The model seeds each open ocean location with a particle and finds the particle's most recent contact with a shallow bathymetry by a back-trajectory issued from altimetry data. It provides the time and the position at which the contact took place. An exponential scavenging relation (with a time constant based on in situ iron data from the KEOPS2 cruise) is used to estimate the decreasing of bio-available iron along the trajectory from the time of the contact with the potential iron source to the current position. Here, the model was applied to the entire Southern Ocean using a grid of 0.25° , a time window of 2004-2015 (one map every 4 days), potential sources of iron in the bathymetric band shallower than 500m, and a diffusion term of $40 \text{ m}^2 \text{ s}^{-1}$.

3. Results and discussion

Based on satellite-derived observations, a large range of both annual mean chlorophyll *a* (chl *a*) concentration and net primary production (NPP) ($>0.1 \text{ mg chl } a \text{ m}^{-3}/20 \text{ g C m}^{-2} \text{ y}^{-1}$ to more than $2 \text{ mg chl } a \text{ m}^{-3}/400 \text{ g C m}^{-2} \text{ y}^{-1}$, respectively) were observed across the SO. It occurs at a variety of scales, including the latitudinal and the regional (i.e. Indian, Pacific and Atlantic; Fig. 1a) scales in the SO. The annual mean chl *a* concentration is associated with the amplitude of its seasonal cycle (Fig. 1a&c). In contrast, the timing of blooms (i.e. the maximum of the annual chlorophyll cycles, Fig. 1d) appears entirely decoupled from both large-scale patterns in annual chl *a* concentration and its seasonality (Fig. 1c), hence revealing a complex organization of phenological patterns of the chl *a* annual cycle.

Here, we disentangle the spatial complexity of chl *a* seasonal cycle using a K-mean clustering method [D'Ortenzio *et al.*, 2012] with the goal of defining distinct bio-regions with similar large-scale patterns in PH (Fig. 2). In brief, this technique statistically gathers regions that exhibit similarly shaped seasonal chl *a* cycles (see the section 2.2 for more details). The seven distinct bio-regions so identified are distributed along a latitudinal gradient (Fig. 3a). At temperate latitudes, two bio-regions (i.e. 1 and 2) exhibit an early bloom in October. Moving south to 45-63°S, two other bio-regions (i.e. 4 and 5) display an annual chl *a* cycle with a delayed bloom in November/December. Finally, near the Antarctic continental shelf (< 63°S), the dominant bio-regions 6 & 7 were characterized by a late bloom in January/February. Interestingly, one bio-region (i.e. 3) differs drastically from the general phenological trend of a delayed bloom moving from subtropical to polar latitudes, having a late bloom at a moderate latitude ($\approx 43^\circ\text{S}$, Fig. 2 & 3d). Furthermore, the co-existence of bio-regions at a same latitude indicates variations in the seasonality of the annual chl *a* cycle (e.g., 1 & 2, 4 & 5 and 6 & 7) which are related to distinct levels of PB (Figs 3b-c).

From temperate to high latitudes, the latitudinal gradient in bloom timing clearly follows changes in light regime (expressed as the annual mean photosynthetically available radiation in the double boxplot; Fig. 4a). The unique bio-region (i.e., 3), characterized by a paradoxically late bloom with respect to the timing of light availability, is mainly located in the vicinity of the Antarctic Circumpolar Current. Interestingly, this region corresponds to the highest Southern Ocean wind stress and depth of the mixed layer (see SI, Fig. S1). Deep mixed-layers likely impose severe light limitation to phytoplankton growth in spring, hence delaying the bloom later in season. Close to the Antarctic shelf, as shown locally in the western Antarctic Peninsula [Venables *et al.*, 2013], the timing of the bloom (Fig. 2) was not related to winter sea-ice extent but more to the seasonal cycle of irradiance.

To address the potential causes of regional-scale variability in PB (Fig. 1a), we hypothesize that iron availability represents its first order driver [Hutchins and Boyd, 2016]. Using integrative and multidisciplinary approaches, we disentangle the impact of four important iron supply mechanisms on PB. These surface-layer iron-sources are shallow plateaus (<500 m) which both (i) locally recharge the surface-layer in iron [Arrigo *et al.*, 2015; Graham *et al.*, 2015] and (ii) remotely recharge the surface layer by lateral advection of non-consumed iron [d'Ovidio *et al.*, 2015; Graham *et al.*, 2015]; (iii) sea-ice, which recharges the surface as ice melts [Arrigo *et al.*, 2015; Lannuzel *et al.*, 2016]; and (iv) vertical mixing through deep convection, which can entrain iron from deep water [Tagliabue *et al.*, 2014b]. Below, the potential impact of these four likely iron-sources on the intensity of PB is investigated. Note that the Aeolian iron deposition was not considered here as a major iron supply mechanism due to its intermittency and its unclear role in regulating PB [Boyd *et al.*, 2012; Cassar *et al.*, 2007; Tagliabue *et al.*, 2009; Tagliabue *et al.*, 2017].

Clearly, the highest PB ($> 1.2 \text{ mg m}^{-3}$) are detected in shallow areas (< 500 m), on continental and island shelves where iron fluxes are expected to be significant [Boyd *et al.*, 2012; Tagliabue *et al.*, 2014a]. Furthermore, we use a Lagrangian model based on altimetry in order to track which water parcels have been recently in contact with a shallow area – and when – to assess the probable locations of some of the iron plumes downstream of these iron sources and to estimate their loss of bio-available iron during advection by scavenging [d'Ovidio *et al.*, 2015]. We find a well-defined decline in PB (from 0.8 to 0.3 mg m^{-3}) with distance and time from the most recent contact of the water parcel with a shallow area (Fig. 4b): the further away a water-parcel is from the initial shallow area iron source, or the longer after a water-parcel has been iron-enriched in shallow areas, the weaker is the PB signal. We note however that seamounts and submerged plateaus do not enhance local or downstream PB, as already observed north of Kerguelen plateau [Graham *et al.*, 2015] (Figs. 1a&4c),

suggesting that not all shallow areas are active and/or bioavailable iron sources. Close to the Antarctic shelf and coastal polynyas, the potential role of the seasonal melting of sea-ice [Arrigo *et al.*, 2015; Lannuzel *et al.*, 2016] as iron source is revealed with enhanced PB ($\approx 0.4 \text{ mg m}^{-3}$).

Moving away from potential iron delivery from shallow areas (either locally or downstream) and from sea-ice, we investigated the possible role of mixed-layer depth on regulating PB. If deepening of the mixed layer in winter can recharge the surface layer in nutrients (iron, but also nitrate and silicic acid in the temperate regimes) by deep convection, it can also limit PB by reducing light availability for phytoplankton growth during the growing season. Indeed, region of deep winter mixing in the Southern Ocean are mostly located directly north of the ACC, where strong winds and isopycnal tilts, are associated with an season-wide weak stratification, leading to deep winter mixed layer but also to relatively deep, compared to their surrounding, spring and summer mixed-layers.

Our findings clearly translate this complex interplay between availability of nutrient and light related to the winter mixed-layer depth. An increase in PB is observed with the deepening of the winter mixed-layer until a maximum winter depth of about 150 m. When the winter mixed layer becomes deeper, a decrease in PB is observed, likely resulting from a stronger light limitation of phytoplankton growth during the growth season (Sep-Mar; on average, winter mixed layer deeper than 150 m are associated to Sep-Mar mixed layer deeper than 110 m or so; Fig 4b). We thus depict here a tipping winter MLD at $\sim 150\text{m}$, which likely represents a shift between an iron-limited to a light-limited environment. Such threshold in the MLD is certainly dependent on the local structure of the iron profile [Tagliabue *et al.*, 2014b]. Despite the HNLC (High Nutrient Low Chlorophyll) nature of the Southern Ocean, an intermediate mixing mode (i.e., with a winter MLD maximum from 120 to 200 meters) appears to enhance PB by supplying an optimal combination of both light and nutrient

requirements for phytoplankton growth. This apparently modest effect of optimal mixing on PB may nevertheless have major implications due to the large areal extension of those zones relying on the vertical mixing to support phytoplankton iron requirements (Fig. 4c).

Deciphering the complexity of the SO's phytoplankton phenology and biomass is particularly critical to assess how climate variability and change might regionally impact the biological pump. Our results reveal two main scales of organization for both PB and PH, i.e. a large latitudinal and a regional scale. Alterations of stratification that are expected with global warming may modify light-mixing regimes, and thus potentially modulate the bloom timing. However, since the phenology appears here to be strongly linked to the circumpolar seasonal cycle in irradiance, we suppose that only drastic changes in stratification could significantly alter bloom timing, and therefore the impact on the biological pump might be expected to be relatively minor.

Concerning the fate of phytoplankton biomass in the changing SO, no particular changes at local scale are expected in shallow bathymetry (i.e. where iron recharge from shallow plateaus (<500 m) are not prone to be affected). Potential changes also likely to remain minor in downstream transport from ACC, depending on whether the ACC has shifted locally in some regions over the past decades according to climate model [*Kim and Orsi*, 2014; *Sallée et al.*, 2008] but not others [*Gille*, 2014; *Shao et al.*, 2015]. Even if ACC changes remain unclear under next century climate change scenarios [*Meijers et al.*, 2012], past studies indicate that if there is a change, the expected position change should be small compared to the size of the bathymetry structure it interacts with to give rise to ocean surface iron plumes.

The main uncertainties regarding the fate of PB thus remains linked to the role of the mixed-layer depth and the sea-ice, which have a more direct response to climate variability and change. Climate modes have direct regional impact on both sea-ice and mixed-layer

depth [Sallée *et al.*, 2010; Simpkins *et al.*, 2012]. In response to SAM and ENSO, sea-ice extends in some regions and is reduced in others, which, according to our results, would have regional impact on iron delivery and PB intensity. Similarly, SAM tends to deepen mixed-layer depth regionally but shallows it in other regions. Under next century climate change, the response is more regionally consistent, with an overall shallowing of the mixed-layer associated with increased stratification (freshwater input from glacier and sea-ice, and increased heat forcing from the atmosphere), and an overall reduction of sea-ice cover.

However, according to our findings, a shallowing (or deepening) of the mixed-layer is not expected to translate to the same PB response everywhere (see SI, Fig. S2). Indeed, the response would be state-dependent: a shallowing of a deep mixed-layer (typically deeper than 200 m) would result in an increase in PB, while a shallowing of an already relatively shallow mixed-layer (typically less than 120 m), would reduce PB. At the SO scale, a global shallowing of the mixed-layer depth appears to have negative consequences on PB (loss of $\approx 0.3 \text{ \% m}^{-1}$; see SI), mainly due to the areal increase of depleted-iron regions with less efficient vertical iron recharge. Conversely, a positive response of PB (increase of $\approx 0.15 \text{ \% m}^{-1}$; see SI) would be observed with a global deepening of the mixed-layer until a critical threshold (+40 m compared to the actual mixed-layer depth climatology), from which severe light limitation damps phytoplankton growth and biomass. Given the more direct response of the MLD to climate variability and change in the SO, these results highlight the crucial role of the local vertical mixing and the associated subtle balance between light/nutrient availability on regulating PB in the majority of the HNLC areas of the SO.

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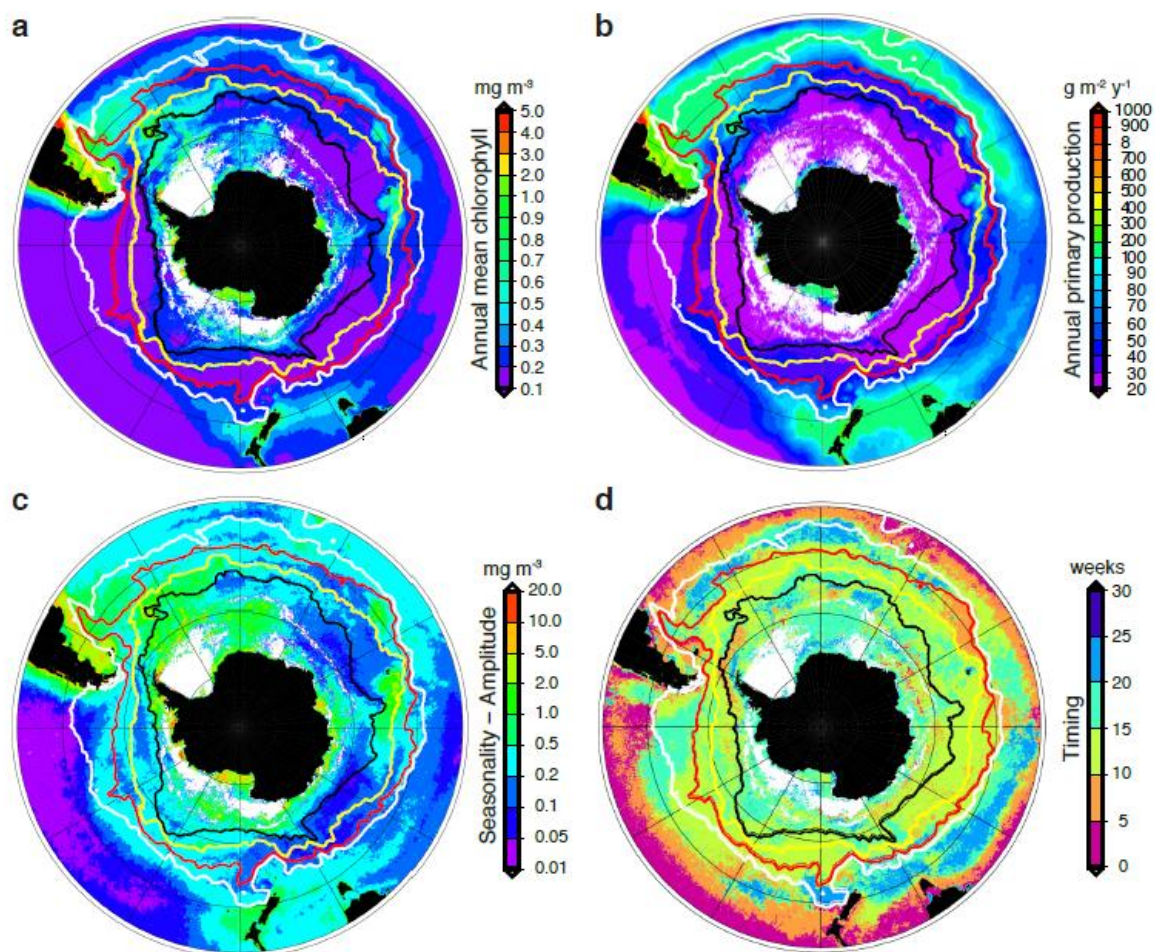


Figure 1: Phytoplankton biomass, net primary production and phenology in the Southern Ocean. Map showing the climatological (a) annual mean chlorophyll *a* concentration (mg m^{-3}), (b) annual net primary production ($\text{g C m}^{-2} \text{d}^{-1}$), (c) the chlorophyll *a* seasonality (mg m^{-3}), (d) the timing of the bloom (week) maximum based on the GLOBcolour time series (1998–2014). The climatological annual net primary production (b; $\text{mg C m}^{-2} \text{d}^{-1}$) was derived from the adapted-SO NPP model of Arrigo et al. (2008). Frontal positions calculated from MADT contours are shown for the STF (white), the SAF (red), the PF (yellow) and the SACCF (black).

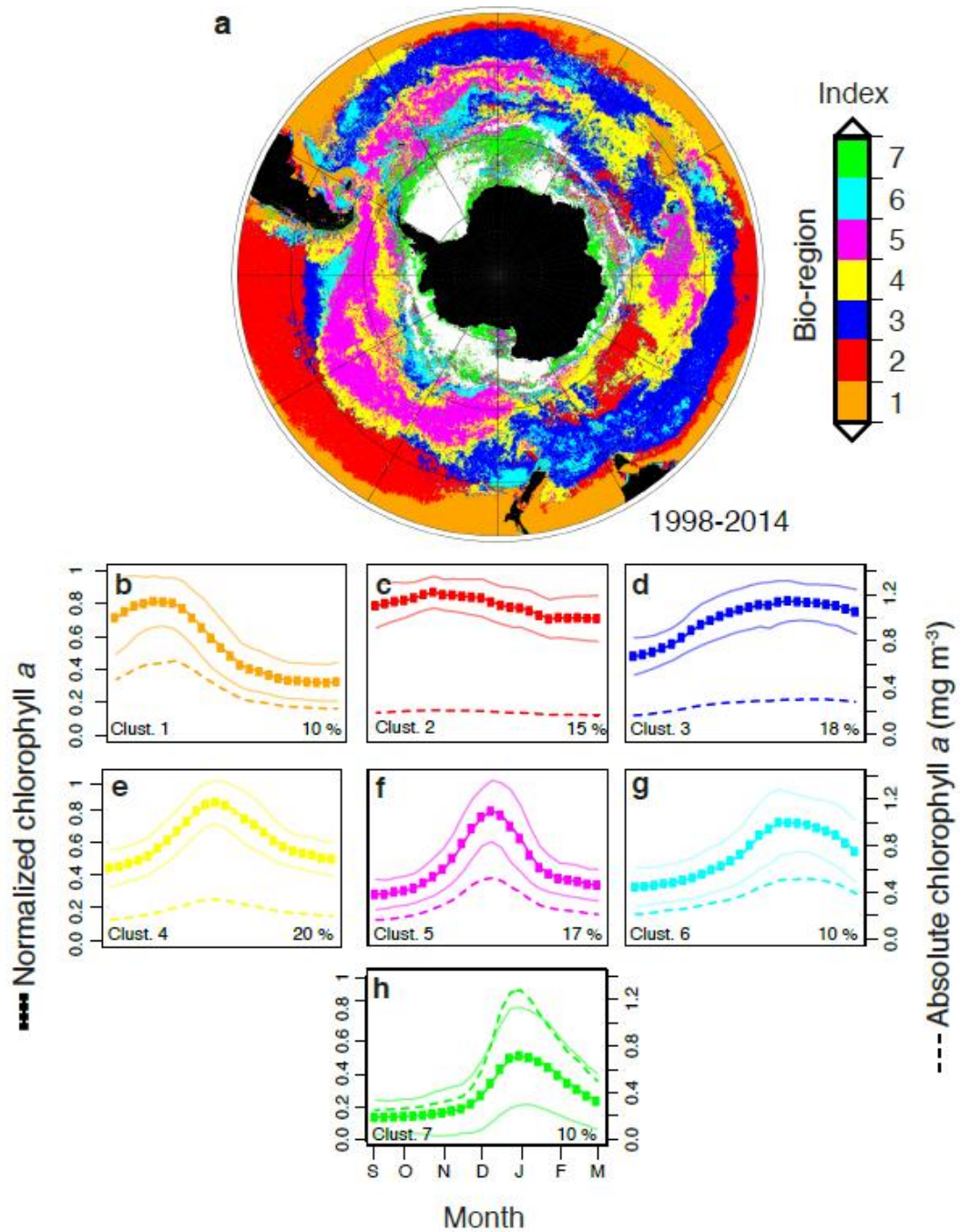


Figure 2: Biogeography of the Southern Ocean. Spatial distribution (a) of the 7 bio-regions (i.e. bio-regions 1 to 7) of the Southern Ocean obtained from the k-means analysis. The normalized annual chlorophyll cycles (continuous lines) of the centers of the clusters (b – h) obtained from the k-means analysis are presented, as well as their respective standard deviation. The absolute annual chlorophyll cycles (dashed lines) corresponding to the centers of the clusters (b – h) are also indicated.

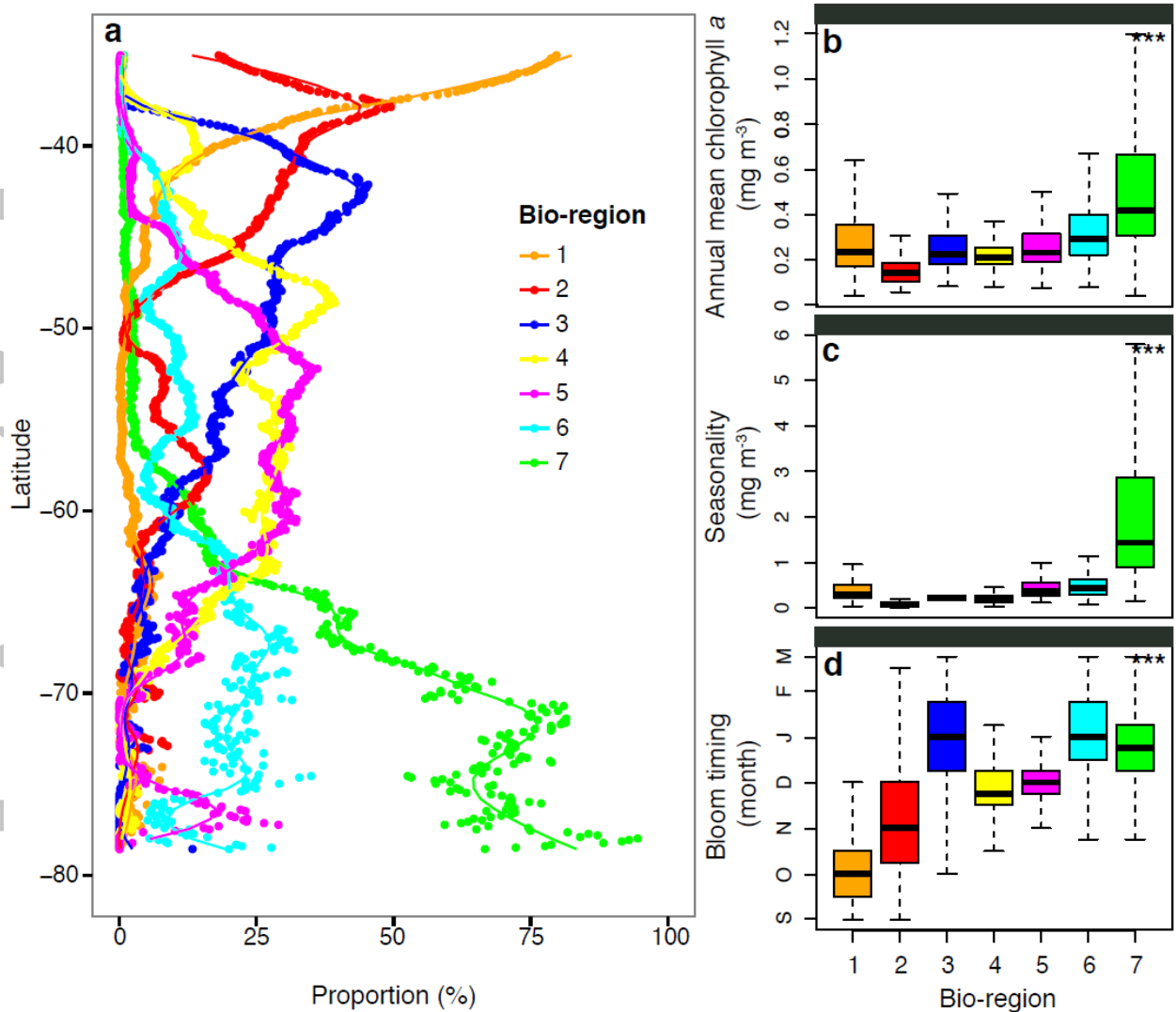


Figure 3: Latitudinal repartition of the bio-regions and their respective biological characteristics. Latitudinal proportion (a) of the different bio-regions. Box plots of the bio-regions (1 to 7; x axis) against (b) annual mean chlorophyll *a* concentration (mg m⁻³), (c) the chlorophyll *a* seasonality (mg m⁻³), (d) the timing of the bloom maximum (month; y axis). The line in the middle of each box represents the region median. The top and bottom limits of each box are the 25th and 75th percentiles, respectively. The lines extending above and below each box, i.e., whiskers, represent the full range of non-outlier observations for each variable beyond the quartile range. The results of the Kruskal–Wallis H test are shown in figures (b) through (d), and depict regions with statistically significant differences between the

climatological input variables at the 95 % level ($p < 0.05$). The codes of the test significance are as follow $\therefore p < 0.05$, *: $p < 0.01$, **: $p < 0.001$ and ***: $p < 0.0001$.

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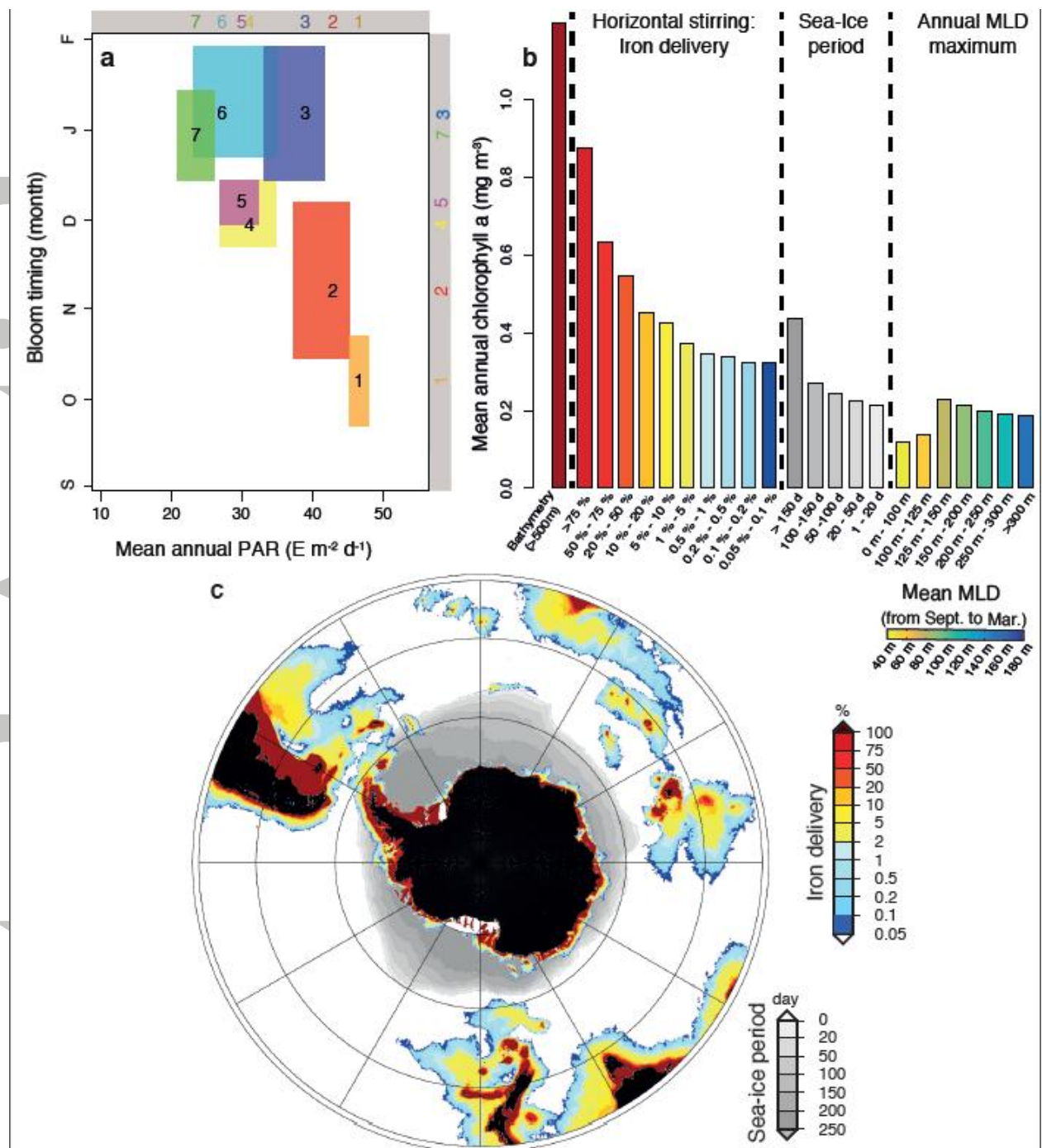


Figure 4: Environmental control of phytoplankton biomass and phenology in the Southern Ocean. Double boxplot (a) of the maximum bloom timing (weeks) versus the annual mean PAR ($E\ m\ y^{-1}$) of the different bio-regions. The top/right and bottom/left limits of each box are the 25th and 75th percentiles, respectively. The numbers in each box are located at the median of both maximum bloom timing and annual mean PAR. Barplot (b) of the mean annual chlorophyll *a* concentration according (1) the shallow areas (< 500 m; red),

(2) areas where iron delivery downstream take place (%; percent of iron remaining in a water parcel after scavenging in respect to its initial concentration acquired in shallow areas; red to blue), (3) areas characterized by a seasonal sea-ice cover (grey) and (4) areas where variations in the annual maximum MLD are analyzed (white). See the map (c) delineating the distinct areas listed just above.