Decadal variability and trends of the Benguela Upwelling System as simulated in a high-resolution ocean simulation

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Abstract

Detecting the atmospheric drivers of the Benguela Upwelling Systems is essential to understand its present variability and its past and future changes. We present a statistical analysis of an ocean-only simulation driven by observed atmospheric fields over the last decades with the aim of identifying the large-scale atmospheric drivers of upwelling variability and trends. The simulation is found to reproduce well the seasonal cycle of upwelling intensity, with a maximum in the June-to-August season in North Benguela and in the December-to-February season in South Benguela. The statistical analysis of the interannual variability of upwelling focuses on its relationship to atmospheric variables (sea level pressure, 10 m-wind, wind stress). The relationship between upwelling and the atmospheric variables differ somewhat in the two regions, but generally, the correlation patterns reflect the common atmospheric pattern favoring upwelling: southerly wind/wind stress, strong subtropical anticyclone, and an ocean-land sea level pressure gradient. In addition, the statistical link between upwelling and large-scale climate variability modes was analyzed. The El Niño Southern Oscillation and the Antarctic Oscillation exert some influence on austral summer upwelling velocities in South Benguela. The decadal evolution and the long-term trends of upwelling and of ocean-minus-land air pressure gradient do not agree with Bakun’s hypothesis that anthropogenic climate change should generally intensify coastal upwelling.

1 Introduction

The Benguela Upwelling System (hereafter BUS) is one of the four major Eastern Boundary Upwelling Systems (EBUs) of the world (Shannon, 1985; Leduc et al., 2010) and among the most productive oceanic regions (Bakun et al., 2010; Leduc et al., 2010). Nutrient-rich coastal upwelling in the EBUs are mainly driven by wind patterns that cause offshore Ekman transport that cannot be balanced by the horizontal advection of water (Bakun and Weeks, 2004). Further offshore, the wind stress curl causes
upwelling by Ekman pumping. These areas are thus rich in pelagic fish biomass and important for coastal fisheries (Bakun et al., 2010). The BUS, off Angola, Namibia and South Africa (Blanke et al., 2005), has its northern boundary at the Angola-Benguela Front (between 14 and 17° S) (Shannon and Nelson, 1996; Veitch et al., 2010). The southern boundary of the BUS is defined by the Agulhas retroflection at the southern tip of the continent where Agulhas Current water penetrates into southern Benguela at 37° S (Shannon and Nelson, 1996). The main driver of the BUS is thought to be the wind stress off Southwest Africa (Nelson and Hutchings, 1983) while north of the front the upwelling is more strongly related to the larger-scale equatorial upwelling (Hardman-Mountford et al., 2003). The strongest upwelling takes place near Lüderitz (27° S) (Shannon and Nelson, 1996), resulting in intense cold sea surface temperature (SST) that persist throughout the year (Parrish et al., 1983). This upwelling cell naturally divides the BUS into a northern and a southern part (Shannon and Nelson, 1996; Hutchings et al., 2009). In southern Benguela the upwelled water occurs near the coast, whereas in northern Benguela it extends farther off the coast, westward up to about 150–250 km (Shannon and Nelson, 1996; Fennel et al., 2012).

The upwelling intensity is highly seasonal in the temperate latitudes (30–34° S), with generally higher intensity in boreal spring and summer, but more uniformly distributed across seasons in regions closer to the equator (15–30° S) (Parrish et al., 1983; Mackas et al., 2006; Chaigneau et al., 2009; Chavez and Messié, 2009). However, a clear picture of the upwelling seasonality is not established yet. Bakun and Nelson (1991) stress that the strongest upwelling takes place in the austral summer seasons, Blanke et al. (2005) widen this seasonal frame to October to March. In contrast, Hagen et al. (2001) and Hagen et al. (2005) argue that the main upwelling season is rather the austral winter to spring. Veitch et al. (2010) clarify that upwelling peaks in southern Benguela in austral spring and summer, whereas in the north, with weaker seasonal variations, it does in austral autumn and spring. Central Benguela upwelling by Lüderitz does not show a clear seasonal cycle (Shannon and Nelson, 1996). This distinction supports
the division in two distinct regimes (Shannon, 1985) and highlights the complex nature of the BUS, which warrants a closer look at the large-scale climate drivers.

The Benguelan upwelling is thought to be driven by several factors. The coastal topography and the climatological winds frame the areas of upwelling (Shannon, 1985; Chavez and Messié, 2009). The region is dominated by southerly and southeasterly winds (Hagen et al., 2001; Risien et al., 2004) that are influenced by the high pressure system over the South Atlantic, the cyclones moving westward over southern Benguela and the pressure over the southern African (Shannon and Nelson, 1996; Risien et al., 2004). Seasonal trade winds also influence the dynamics of the upwelling (Shannon, 1985; Chavez and Messié, 2009). In addition, the BUS displays some particular characteristics, compared to other EBUs, determined by its geophysical boundaries: the Agulhas retroflection, the Angola-Benguela Front and the passage of westerly winds in the south (Shannon and Nelson, 1996). Zooming more closely into the BUS, Lachkar and Gruber (2012) characterized the Benguela subregions as having a shallow mixed layer depth, but with central Benguela dominated by strong upwelling due to a wide shelf and low eddy activity, whereas southern Benguela is characterized by weaker upwelling, a wide shelf, and moderate to high eddy activity. In northernmost Benguela, the upwelling is moderate in a narrow shelf, with moderate to high eddy activity.

The upwelling intensity is not constant, but presents intraseasonal variations. In general, the variability of the Benguela system is dominated by the intraseasonal (eddy) variability (Chavez and Messié, 2009). Lachkar and Gruber (2012) identified factors that inhibit the net primary production and thus have indirectly been linked to upwelling dynamics, namely strong eddy activity, a narrow continental shelf and a deep mixed layer. These factors help to provide a spatial characterization of the Benguela system and its spatial heterogeneity (Lachkar and Gruber, 2012).

From a larger scale climatic perspective, climate modes seem to have an impact on the interannual variability of upwelling through modulation of the local conditions. Hagen et al. (2001) mentioned a possible influence of the Quasi-Biennial Oscillation (QBO), a mode of variability that describes quasi-oscillations of low-stratospheric
winds. The St. Helena Index (HIX, first mode of the Empirical Orthogonal Function (EOF) Analysis of air temperature, sea level pressure (SLP) and precipitation) is positively correlated with the southeasterly trades and thus should modulate the strength of the upwelling (Hagen et al., 2005). Furthermore, the Antarctic Oscillation (AAO), that is related with the strength of the circumpolar westerly flow (Jones and Widmann, 2004), could also influence the upwelling intensity.

In addition to the mentioned climate modes, the impact of El Niño Southern Oscillation (ENSO) on the Benguela upwelling could, according to some authors (e.g. Dufois and Rouault, 2012), be of major importance, even stronger than the one of the AAO (Rouault et al., 2010). According to this study, El Niño and La Niña events tend to weaken and strengthen upwelling, respectively. The proposed physical mechanism is related to an equatorward shift in the high pressure over the southern Atlantic which leads to weaker upwelling-favourable trades (and conversely for La Niña events) (Dufois and Rouault, 2012). The results obtained by Shannon and Nelson (1996) support this by identifying the relation to ENSO as the most significant one on interannual time scales. Deutsch et al. (2011) also presented changes in the upwelling intensity and the depth of the thermocline associated with ENSO.

Some authors (e.g. Huang et al., 2005) suggest that ENSO, affecting the whole tropical belt around the Earth, also has an influence on the so called Benguela Niños. These extreme events are a type of Atlantic Niños that reach the BUS when warm waters extend into the eastern Atlantic around 600 km further south than usually (Shannon et al., 1986). The Benguela Niño events have been explained by two different causal chains. According to the first, SST anomalies along the southwest African coast develop due to zonal wind anomalies in equatorial South America and the equatorial western Atlantic (McCartney, 1982; Wang, 2006). Kelvin waves crossing the basin are induced due to the relaxation of the trades along the equator (Rouault et al., 2007). The unusually sea surface height reaches the Angola-Benguela Front which leads to increased intrusion of tropical water into the Angola-Benguela Upwelling System (McCartney, 1982; Rouault et al., 2007). According to second scenario (Richter et al., 2010), the
main drivers of Benguela Niños are alongshore wind anomalies, which occur due to the weakening of the South Atlantic High that develops two to three months in advance of Benguela Niños and five to six months before the Atlantic Niños. Furthermore, they found that Kelvin waves had an insignificant influence on Benguela Niño event in 1995. Most past major warming events, which usually occurred between February and May and lasted a few months, were related to Atlantic Niño, and cooling events to Atlantic Niña state, but the relationship does not seem to be linear (Rouault et al., 2010; Dufois and Rouault, 2012). The warming events have been found to be weaker and less frequent than the Pacific ENSO (Shannon et al., 1986; Latif and Grötzner, 2000). In contrast, Hardman-Mountford et al. (2003) estimate a much higher frequency of warm events in the Atlantic than in the Pacific, but these authors found that the warming events penetrate more rarely into the Benguela system. For this to occur, southerly winds at the equator are required to blow simultaneously with the southward water flow (Hardman-Mountford et al., 2003). Shannon et al. (1986) find that, in any case, the effect on southern Benguela is small, whereas in northern Benguela the intrusions of warm water persist for at least six months, leading there to extreme events in precipitation in Namibia and Angola (Rathmann, 2008), and weaker or shorter (about two months) upwelling season (Hagen et al., 2001). Hagen et al. (2001) found a periodicity of the occurrence of Benguela Niños around every 11 years (1909–1963) and between 1974 and 1999 every five years, and interpreted this shift as a tendency of these extreme events to occur more often.

Another ongoing discussion hints to a more complex interaction from the Tropical Pacific into the Tropical Atlantic as a simple unidirectional influence from the Pacific to the Atlantic. Rodríguez-Fonseca et al. (2009) mentioned that ENSO events are preceded by Atlantic events of opposite sign and that the Atlantic could strengthen ENSO. Ham et al. (2013) agree and explain that easterlies over the equatorial far-eastern Pacific are considerably weaker under Atlantic Niña conditions. The same study adds that the North Tropical Atlantic is more strongly connected with the Pacific ENSO than with the Atlantic Niño. In contrast, Wang (2006) detected no correlation between both El Niños
(Pacific and Atlantic) but found that anomalous SSTs of the two equatorial oceans can induce an inter-Pacific-Atlantic SST gradient that leads to anomalies in surface zonal wind over parts of both oceans and equatorial South America. These anomalies of the zonal wind constitute the bridge conveying the interaction between the two oceanic basins. Wang (2006) stated further that the SST gradient between the equatorial Pacific and Atlantic is of higher importance than the individual ocean SST anomalies as driver of the atmospheric circulation across northern South America. Contrasting to the results of Rodríguez-Fonseca et al. (2009), Latif and Grötzner (2000) mentioned that the equatorial Atlantic responses to ENSO with a lag of six months, this time being needed by the equatorial Atlantic to adjust to low-frequency wind stress variations. Also Colberg and Reason (2006) found that ENSO-induced wind anomalies play a major role in driving upper South Atlantic temperatures by an atmospheric teleconnection with a lag of one season. El Niño changes the Walker Circulation so that an ascending branch is located over Benguela during this phase (Rathmann, 2008). Thus, even if the direction of teleconnection is unclear, the influence of ENSO and the Benguela Niños on the Benguelan upwelling could be of major importance when investigating the atmospheric drivers due to their affects on the Benguelan ecosystem.

The long-term trend of the Benguela upwelling brought about by anthropogenic climate change is another important reason to identify its large-scale atmospheric drivers. In his landmark paper, Bakun (1990) put forward his hypothesis that can be summarized as follows: as a consequence of rising greenhouse gas concentrations the surface temperature over the continents warm faster than oceans. This leads to a strengthening of the continental lows and oceanic highs. The land–sea pressure gradient is thereby enhanced, causing a strengthening of alongshore winds and enhancing upwelling. In addition, the warming ocean surface results in more humidity in the otherwise very dry atmosphere over land, which reinforces the greenhouse gas effect (Bakun, 1990). Observations do indicate that over Southwest Africa surface temperatures over land have increased more rapidly than over the ocean from 1980 onwards, although in previous periods the statistical significance of the trend is compromised due to a poorer
In agreement to Bakun’s hypothesis, Narayan et al. (2010) found decreasing trends in coastal SSTs in four major EBUs over 1960–2000 and attributed them to meridional wind stress intensification. This was supported by Demarq (2009), who found a positive trend in southerly winds in the Benguela EBU for 2000–2007. In addition, Rathmann (2008) report an intensification of the trade winds due to SLP trends. However, Narayan et al. (2010) recognized the spread in estimation of wind stress trend using different data sets (reanalyses and observations) and Belmadani et al. (2014) could not identify the mechanisms linking land–sea thermal contrast and upwelling-favourable winds in the Peru upwelling system in future climate model simulations. Bakun et al. (2010) also expressed some doubts about the robustness of the estimated upwelling trends in Benguela. To explain this apparent lack of clear upwelling trends in BUS, Bakun et al. (2010) later reasoned that the influence of ENSO on atmospheric humidity in Benguela, combined with the succession of recent strong ENSO, events may have stifled a long-term intensification of upwelling that should have occurred as a result of the anthropogenic greenhouse effect. However, climate models still give an unclear picture about the long-term effect of anthropogenic greenhouse gas forcing on ENSO (Vecchi and Wittenberg, 2010), and still underestimate the atmospheric thermodynamical response to SST changes (Kim et al., 2014).

In addition, global climate models still display clear systematic errors in simulating the SST in the eastern ocean basins (Richter and Xie, 2008; Echevin et al., 2012; Grodsky et al., 2012), the origin of these errors still being under investigation.

Against the backdrop of these competing hypothesis about the role of large-scale climate modes on Benguela upwelling and the long-term evolution of upwelling itself, the purpose of this paper is to analyse the atmospheric drivers of the interannual variability and the long-term trends of upwelling in the BUS as simulated in one high-resolution (about 0.1° lon–lat) ocean simulation with the ocean model MPI-OM developed at the Max-Planck-Institute for Meteorology in Hamburg, and spanning the last five decades driven by the NCEP/NCAR meteorological reanalysis. The focus of this analysis lies on the longest timescales possible for such high resolution simulations. This focus re-
quires a trade-off between a possibly higher quality atmospheric forcing data provided by most recent meteorological reanalysis and the period covered by those data sets, which is considerably shorter that those of the NCEP/NCAR reanalysis. Although the ocean simulation was driven by the NCEP/NCAR reanalysis, some analysis were confirmed with the more recent and higher resolution ERA Interim reanalysis. However, through this study, it has to be borne in mind that the conclusions obtained may be dependent on the realism and homogeneity of the NCEP/NCAR product.

The paper has the following structure. The data and methods are described in Sect. 2, followed by a basic validation of the model output regarding the upwelling. The results are presented in the following two sections. First, the analysis of the upwelling index constructed from the modelled vertical velocity is displayed, and second, the results of the investigation on the SLP gradient and the long-term trends in the simulating upwelling are shown. Finally, these results they are discussed in Sect. 6, with our most important conclusions in Sect. 7.

2 Data and methods

For this study we analyse climate variables derived from observational data sets, atmospheric data from meteorological reanalysis and from a high resolution global ocean-only simulation. A summary of the characteristics of these data sets is presented in Table 1, with Fig. 1 showing a sketch of the geographical boxes used to define some indices.

2.1 Gridded observational data

Atmospheric variables (10 m wind, wind stress, and SLP) were obtained from the National Centers for Environmental Prediction/ National Center of Atmospheric Research (NCEP/NCAR) reanalysis 1. This data set is available from 1948 onwards with a horizontal resolution of 2.5° (Kalnay et al., 1996). Monthly means were downloaded and
averaged to seasonal means of the area 50°W–40°E, 0–50°S. An analysis of the gradient of the SLP field was performed with the SLP of ERA Interim reanalysis (Dee et al., 2011) as well as of NCEP/NCAR reanalysis 1. Here we chose a region over the ocean (20°W–10°E, 15–35°S) and one over land (12–25°E, 10–20°S), subtracted the SLP over the land from the SLP over ocean, calculated its trend and the significance of the trend (p value), and correlated it with the upwelling indices derived from the vertical velocity (described in Sect. 2.3). The chosen regions are identified as most closely correlated to the upwelling.

In addition to the reanalyses data, the observational gridded sea surface temperature data set HadISST1 with a grid spacing of 1° and data between 1870 and 2012 (Rayner et al., 2003) was used in this study for the validation of the model output. For the same purpose, the advanced Very High Resolution Radiometer (AVHRR, Casey et al., 2010) version 5.0 in 4 km resolution remote-sensed SST, a radiometer on board of the NOAA satellites was used to detect the possible SST bias of the model data described in the following. However, it is known that, this version of pathfinder has a warm bias of up to 5°C close to the coast in austral summer in southern Benguela (Dufois et al., 2012).

2.2 Model data

The ocean simulation analysed here is a global simulation (hereafter denoted STORM) with a high-resolution version of the state-of-the-art model MPI-OM (Marsland et al., 2003) covering the period 1950–2010 driven by the global NCEP/NCAR global meteorological reanalysis (von Storch et al., 2012; Li and von Storch, 2013). The horizontal resolution of this model version is 0.1° and it has 80 levels.

2.3 Upwelling indices

To identify the large-scale atmospheric drivers, an upwelling index had first to be defined. The definition may depend on the available data, since not always direct data of the vertical water mass transport are readily available. Chen et al. (2012) defined an
index derived from the SST and mentioned that it represents well the upwelling intensity in terms of the spatial variation while the index defined derived from the offshore Ekman transport represents the temporal variations. The same study found that the index constructed from the SST is not able to correctly provide the upwelling intensity during warm water intrusions of the Angola Current.

In general, SSTs are not only affected by upwelling but also by a complex interaction between horizontal ocean advection, ocean–atmosphere heat fluxes and vertical mixing (McCabe et al., 2015). Therefore, the simulated vertical mass transport at 52 m depth, close to the modeled mixed layer depth in the Benguela region, is used as upwelling index. All data set were seasonally averaged using the standard seasons definition: December-to-February (DJF), March-to-May (MAM), June-to-August (JJA) and September-to-November (SON). The index derived from the vertical velocity is denoted here up_wvelo. Previous studies (Shannon and Nelson, 1996; Hutchings et al., 2009) suggested that there exist a strong difference in the behaviour of the upwelling in northern and southern Benguela. Thus, the area was divided in two regions with the border at Lüderitz (28° S) (Fig. 1). The South Benguela region covers around 2.7 times the oceanic region of North Benguela, because the south coast of South Africa with the Indian Ocean is included too. There, upwelling takes place as well (Shannon and O’Toole, 2003). Thus, the region denoted by South Benguela contains the southern BUS and additionally the south coast upwelling.

### 2.4 Climate indices

The upwelling index were not only correlated with the atmospheric variables but also with climate indices. The influence of the Multivariate ENSO index (MEI, Wolter and Timlin, 1993), the Antarctic Oscillation (AAO, Marshall, 2003), and the tropical Atlantic (ATL-3, 20° W–0° E, 3° N–3° S, Rodríguez-Fonseca et al., 2009) were investigated, as well as the impact of the Atlantic Meridional Mode (AMM, Chiang and Vimont, 2004), the St. Helena Index (HIX, Feistel et al., 2003), and the Quasi Biennial Oscillation...
(QBO, Labitzke and van Loon, 1988) on the upwelling off southern Africa were examined. The ATL-3 were calculated with the data sets of HadISST1 and STORM.

For the analysis of the statistical significance of the long-term trends in the upwelling indices and the linear correlations between the indices, a two-sided significance level of $p = 0.05$ was adopted. The $p$ value of the correlation and of the trend significance indicates the significance of the correlation (or trend) under the usual assumption of serially uncorrelated and normally distributed regression residuals.

3 Validation

Figure 2 shows the long-term annual means of some important variables obtained from the STORM simulations in the South Atlantic and in the Southeast Atlantic: the vertical velocity at 52 m depth, SST (compared to the corresponding fields derived from HadISST1) and surface currents. Figure 2a and b display the upwelling cells at the western and southern coast of southern Africa. The SST pattern of STORM (c, d) and HadISST1 (e, f) look very similar, with colder SST at the coast due to upwelling. The SST of the STORM simulation shows lower values than the observational data set HadISST1. This may be due to relatively coarse resolution of the HadISST1. In Fig. 2g one can see the realistic representation of the equatorial currents, the Agulhas current, Benguela current and the westward drift along the west coast caused by the trade winds. The simulated SST in STORM compared to that of AVHRR is shown in Fig. 3. To have the same spatial resolution in both data sets, the AVHRR data set was spatially smoothed by averaging two grid points in both spatial directions. If the two grid-points include one missing value, the missing value is ignored in the mean. The STORM data set was interpolated onto the grid of AVHRR. The STORM simulation displays a warm bias in the upwelling region, as it is well-known for general circulation models. The cold bias located along the coast may be also partially caused by the bias of the AVHRR data mentioned by Dufois et al. (2012). The stronger bias of AVHRR underlines the more realistic cold SSTs simulated by STORM along the coast compared to HadISST1.
The annual cycle of the upwelling index in North and South Benguela is displayed in Fig. 4. The up_wvelo index in South and North Benguela display clear differences. Whereas in the north upwelling shows a broad maximum in JJA, in South Benguela tends to be stronger in DJF, due to the position of the subtropical Atlantic high (Mackas et al., 2006). This is in broad agreement with observations, considering the discrepancies already indicated in the introduction (Sect. 1).

The relationship between the upwelling index up_wvelo and the SST are displayed in Fig. 5 as a map of the correlation coefficient between the upwelling index and the SST in each model grid cell. These correlations are negative near the coast and more pronounced in North Benguela. The correlations between the up_wvelo of South Benguela and the SST field in STORM are negative along the entire coast in DJF, with stronger correlations in the south. In JJA the correlations are negative only at the south coast of South Africa.

4 Simulated upwelling over the last decades

4.1 Linear trends and variability of upwelling over the last 50 years

Table 2 summarizes the main findings of the analysis of the upwelling index in the STORM simulation, including the estimated linear trends (and their two-sided 95% uncertainty ranges) expressed as the change of upwelling intensity over the period 1950–2010 relative to their long-term mean. The long-term mean values themselves may depend on the depth, area, and quite presumably on model resolution, but they are in the reasonable range of decades of meters per year, which agrees with estimations using simplified models of upwelling dynamics (Rykaczewski and Checkley, 2008).

The time evolution of the upwelling indices (deviations from their long-term mean) through the period of the STORM simulation (Fig. 6) indicates that the variations of upwelling in the north and in the south do not go hand in hand. The significant linear correlation between up_wvelo in North and South Benguela is 0.30 in DJF and −0.52
in JJA, indicating that the interannual variations of upwelling during JJA (the season with maximum upwelling in North Benguela) is out of phase in both regions. These results strongly suggest that the Benguelan upwelling cannot be considered as a homogeneous system, in accordance with previous observational studies (see Sect. 1).

The linear trends of the upwelling indices in the STORM simulations are positive in North Benguela in MAM, JJA and DJF and positive in South Benguela in MAM, SON and DJF (Table 2). Upwelling in North Benguela is about double as intense as in the South Benguela (Fig. 6) and it is less variable relative to its long-term mean. The interannual standard deviation (SD) in the north lies between 15 to 29% of the long-term mean, whereas in the south the relative variability is in the range 48 to 84%. The second characteristic is that the decadal variations are much more coherent across seasons in the north than in the south (Table 2). There is obviously some persistent factor in the tropical upwelling region that coherently drives the upwelling decadal variability in all seasons. In the south, in contrast, the seasonal upwelling indices display a more incoherent behaviour across the four seasons. In addition, the time scales of variations are longer in the north than in the south, as it will be more clearly shown later with a spectral analysis of these indices.

The analysis of the significance of the trends shows that only in North Benguela the positive trends in MAM and DJF are nominally significant (i.e. assuming normal and serially uncorrelated trend residuals, see confidence intervals in Table 2). However, Fig. 6 clearly shows that a linear trend in the north is not a good description of the long-term evolution of upwelling. In general, upwelling intensity remains stable from 1950 to 1985 and undergoes a decadal surge and subsequent slowdown until the end of the simulation. This behaviour is not what one would expect from a steady increase in external climate forcing over the second half of the 20th century. These long-term changes over the whole simulation period are of the same order as the coefficients of variation (SD divided by the long-term mean), if not weaker, which further confirms that the decadal variations are dominating the temporal variability. A more sophisticated analysis of the statistical significance considering the serial correlation of the trend
residuals indicates that both trends are not significant at the 95% level. In the south, the trend in DJF is close to the $p = 0.05$ significance level but the long-term evolution, as displayed in Fig. 6, does show a steady, albeit weak, increase through the simulation. The trend residuals are in this case not serially correlated. The central estimate of the trend in South Benguela upwelling in DJF implies a linearly increase of about 40% from the start through the end of the simulation. We will return to this point when discussing the connection of upwelling to the large-scale climate modes.

4.2 Correlations with large-scale climate fields

Correlations with atmospheric fields were calculated for the up_wvelo index. Figure 7 depicts the correlation patterns between the upwelling index and wind stress (defined as positive when directed downward) for DJF and JJA. This figure not only illustrates the expected relationship between upwelling and wind stress, but also the seasonality in this relationship. In North Benguela, wind stress variability is more important for upwelling during JJA – the season with stronger upwelling – whereas in South Benguela the seasonality is not very marked. In both regions, the direction of upwelling-favourable winds does not change with season. These wind stress correlation patterns can also explain the mutual correlations between the north and south upwelling indices. As indicated before, the correlation between the interannual variations in upwelling in both regions in JJA is negative ($r = -0.52$). North Benguela upwelling is statistically connected to winds south of the southern tip of Africa that actually inhibit upwelling in South Benguela, and vice-versa. In DJF, by contrast, this does not happen, and North Benguela upwelling is statistically linked to winds south of South Africa that are also upwelling-favourable for South Benguela, and vice-versa, explaining the weak but positive correlation between both upwelling indices in this season. In the STORM simulation, upwelling is strongly correlated with upwelling-favourable wind stress in North Benguela. South Benguela upwelling in DJF is strongly driven by wind stress at the south coast and not so pronouncedly by the wind stress at the southern west coast.
This could happen due to our definition of the region chosen as South Benguela, which includes parts of the coast of the Indian Ocean as well.

Summarizing, the up_wvelo in North Benguela is driven mainly by the expected wind pattern. In contrast, the correlations with South Benguela show upwelling-favourable wind stress that are more pronounced at the south coast than at the west coast south of Lüderitz. The correlation pattern with the SLP supports this with a positive pressure anomaly south of the continent instead of an intensified subtropical high located in the central South Atlantic (not shown here).

4.3 Correlations with climate modes

To identify the connection between climate modes and the upwelling index up_wvelo, the correlations of this index with ENSO (MEI index), the Antarctic Oscillation (AAO), the Tropical Atlantic SST (ATL-3), St. Helena index (HIX), the Quasi Biennial Oscillation (QBO), and the Atlantic Meridional Mode (AMM) were also calculated (Table 3). The upwelling in North Benguela in STORM is significantly positively correlated with MEI in MAM, with the AAO in MAM and DJF and HIX in DJF. Significant negative correlation were found with AAO in JJA and ATL-3 in DJF (Table 3). In contrast, the upwelling in South Benguela is significantly negatively correlated with ENSO in MAM and DJF and positively with AAO in all four seasons and AMM in DJF (Table 3). The correlations with the QBO are all insignificant.

The correlation pattern between wind stress and the ENSO index in DJF (Fig. 8) clearly explains the stronger negative correlations with South Benguela upwelling and the negligible correlation in the north. The wind stress anomalies associated with ENSO tend to weaken the climatological winds in the south and are very small in the north. The pattern of wind stress anomalies related to ENSO have a large-scale character, spanning almost the whole South Atlantic at mid-latitudes, being clearly strongest in South Benguela. The correlation between ENSO and wind stress in North Benguela is quite weak, consistent with the weak correlation found between ENSO and North Benguela upwelling. The wind stress correlation pattern indicates that, if anything,
ENSO may contribute to drive upwelling anomalies in South Benguela but not in the north. Even if the correlation is weak, the wind stress pattern favoured by ENSO shows the upwelling-unfavourable influence of a positive phase of ENSO. We suggest that these correlation patterns, which display systematically large values at mid-latitudes and very small values in the tropical regions, arise as a result of a large-scale dynamical teleconnection with ENSO via the mid-latitudes, rather than due to a tropical bridge between the Tropical Pacific and the Tropical Atlantic, as it can be seen in Fig. 8. Also, a link between ENSO and Benguela upwelling resulting from the modulation of smaller scale humidity dynamics in the Benguela region, as discussed by Bakun et al. (2010), does not seem to be required, since the correlation patterns display a large-scale teleconnection structure.

Similarly, the AAO index displays suggestive teleconnection patterns with the wind stress field in DJF and JJA (Fig. 8). These patterns indicate that the influence of the AAO is restricted to South Benguela and suggests the reason why the correlation between the North Benguela upwelling and the AAO is weaker. In both seasons, the AAO is associated with upwelling-favourable winds in South Benguela, and this link is somewhat stronger in DJF.

### 4.4 Spectral analysis

To identify the more important time scales for upwelling variability, we perform spectral analyses on the upwelling index up_wvelo.

The spectra of indices (Fig. 9) highlight their different behaviour in North and South Benguela. In the north, the spectra are red through the whole frequency range, with some superimposed broad and weak peaks. In the south, the spectra flatten for frequencies lower than $2 \times 10^{-1}$ (periods longer than 5 years). In both regions the spectra show some enhanced variability that is not strictly statistically significant, but which are nevertheless documented here as it supports some of the results of the correlation analysis and as it can be useful for future analysis with longer simulations.
The up_wvelo index of North Benguela in STORM displays a very broad spectral peak at periods of around 5 years in the season DJF and 4 years in JJA. In the season SON, the variabilities with periods of 3 years are slightly enhanced, whereas in MAM this occurs for slightly shorter periods of 2.5 years. In South Benguela, the spectra hint at an enhanced variability in the period range of 3 to 5 years, with other maxima at 2.5 years. However, whereas the spectra of the up_wvelo index in both North and South Benguela display enhanced variability at the typical ENSO periodicities, correlations to the ENSO index differ between the two regions (Table 3). The index in North Benguela shows mainly weak correlations to ENSO. The index in the south displays stronger and significant correlations in DJF and MAM.

In summary, the red spectral background in the north is probably a result of tropical dynamics, whereas the more white spectra in the south may reflect the atmospheric forcing at mid-latitudes, which generally has a flatter spectrum itself. Both upwelling spectra display enhanced variability at periods of around 2.5, 3.3 and 5 years. The periods of 5 years in the DJF season suggest a link to ENSO, as it was found in the correlation analysis between the upwelling indices in South Benguela and the ENSO index. The 2.5 period could indicate an influence of the Quasi Biennial Oscillation (QBO). However, the QBO and up_wvelo of STORM are insignificantly correlated (Table 3). Possible reasons for these variations have to be further investigated.

5 SLP gradient and long-term trends in upwelling

Bakun (1990) postulated a long-term intensification of coastal upwelling driven by the increasing greenhouse gas concentration. This hypothesis would formally agree with the positive trend found here in the North Benguela upwelling, although the upwelling evolution cannot be well described by a steady linear trend. In the south, the long-term evolution is described by a linear trend better than in the North, but its magnitude lies just below the 95 % significance level. The mechanism proposed by Bakun (1990) involves a stronger increase of near-surface temperatures over land than over the ocean,
which would lead to an intensification of the continental thermal low pressure relative to the ocean SLP. Consequently, the SLP gradient between land and ocean would tend to increase, the SLP difference between ocean and land becoming more positive, leading to a strengthening of the upwelling favourable wind and hence to an increase of upwelling. We have tested Bakun’s hypotheses concerning the link between the SLP gradient and upwelling in the framework of the ocean simulation STORM in Benguela, and concerning the long-term trend in upwelling in this simulation.

We defined two regions, one over land (12–25° E, 10–20° S), and one over the ocean (20° W–10° E, 15–35° S), and calculate the difference of their respective area-averaged SLP difference ocean minus land. Analysing the trend of this SLP difference in NCEP/NCAR reanalysis and ERA-Interim reanalysis does not support Bakun’s hypothesis in this region. Bakun’s hypothesis would predict a positive trend in this difference – land SLP decreasing relative to ocean SLP – whereas the NCEP/NCAR and ERA-Interim SLP provide significantly negative trends for the seasons MAM (Fig. 10, Table 4), with the trend in other seasons being not significant.

The trends of the SLP over land and over ocean are positive for both data set and all seasons, except the SLP in MAM of NCEP (Table 4). This could indicate a missing imprint of increasing greenhouse gas concentrations on the SLP, especially over land, which is maybe due to the relatively short temporal period covered by the reanalyses data or the reanalyses data itself.

The correlation at interannual time scales yields a clear relationship between the upwelling intensity and the ocean-land SLP gradient. Bakun’s hypothesis, applied at interannual timescales, predicts a positive correlation between upwelling and this gradient. The correlations with the up_wvelo is positive and significant in North Benguela in all four seasons (with NCEP/NCAR and ERA-Interim). South Benguela upwelling is significantly influenced by the SLP difference in MAM, JJA and DJF when correlating with NCEP/NCAR SLP gradient, but not significantly in all seasons when correlating with ERA-Interim SLP gradient.
In other words, the SLP difference between ocean and land derived from these two meteorological reanalyses do not support Bakun’s hypothesis of a long-term intensification due to increasing greenhouse gas concentrations, although there is a relationship at interannual timescales as Bakun (1990) and Bakun et al. (2010) envisaged. The lack of a significant trend does not necessarily indicate that Bakun’s hypothesis is not correct. The reason could lie in an insufficient quality of the reanalysis SLP in this region to identify long-term trends or that the long-term trend in SLP gradient over the last decades is still overwhelmed by other factors and it has not emerged from the background noise yet. Bakun et al. (2010) indicated that this possible factor that blurs the long-term trend in upwelling in BUS is its relationship to ENSO, to which we briefly direct our attention now.

According to the results of the correlation between upwelling and climate modes, and the correlation patterns between the climate modes ENSO and AAO with the wind stress, it appears that the influence of both climate modes is mostly restricted to South Benguela. We attempt to estimate the contribution of ENSO and the AAO to the long-term trend in upwelling in South Benguela in DJF, the season in which the interannual correlations are stronger and the wind stress patterns also display a clearer signal. For this, we set up a linear regression model between upwelling as predictand and ENSO and the AAO as predictors:

\[
\text{up}_\text{wvelo}(t) = \alpha_{\text{enso}} \text{ENSO}(t) + \alpha_{\text{aao}} \text{AAO}(t) + \epsilon
\]

where \(\text{up}_\text{wvelo}\) represents the upwelling index in South Benguela in the DJF, \(t\) represents the year, \(\alpha_{\text{enso}}\) and \(\alpha_{\text{aao}}\) are the regression coefficients and \(\epsilon\) represents the unresolved variance in upwelling. Using the data from the period covered by all three data sets (ENSO, AAO and STORM, 1957–2010) we can estimate the values of \(\alpha_{\text{enso}}\) and \(\alpha_{\text{aao}}\). The contribution of ENSO and AAO to the long-term upwelling trend is then estimated as:

\[
\alpha_{\text{enso}} \text{trend}_\text{enso}
\]
and

\[ \alpha_{\text{aao}} \text{trend}_{\text{aao}} \]

respectively, where trend\textsubscript{\text{enso}} and trend\textsubscript{\text{aoo}} are the long-term trends in the ENSO and AAO indices, respectively. It turns out that the ENSO trend in DJF is very weak and by far not statistically significant, whereas the AAO trend in DJF is positive and clearly statistically significant over the \( p = 0.05 \) level. The South Benguela upwelling in the STORM simulation increases linearly over 1957–2010 by 39\% of its long-term mean value. The estimated contribution of ENSO to the long-term trend in upwelling in South Benguela in DJF is \(-3\%\) of the mean upwelling and the estimated contribution of the long-term trend in the AAO is 41\% of the mean upwelling. This means that, under the assumptions of this simple statistical analysis, the trend in upwelling in South Benguela in DJF simulated in the STORM simulation could be almost entirely explained by the long-term trend in the AAO, with the ENSO contribution remaining negligible.

6 Discussion

The large-scale atmospheric drivers of Benguela upwelling in this ocean-only simulation were identified using an upwelling index based on the vertical velocity.

An important result is the disagreement between our analysis of the SLP gradient and the hypothesis of Bakun (1990). There are several possibilities for this disagreement. The results of the correlation analysis between the upwelling indices and the SLP gradient do show the expected relation of wind forcing driving the upwelling on interannual scales, but the long-term trend in the SLP may be more strongly burdened by uncertainties in the SLP data from NCEP/NCAR reanalysis in the Southern Hemisphere. For instance, Marshall (2003) found considerable deviations in the magnitude of SLP trends between those derived from the NCEP/NCAR reanalysis and those from station data in the Southern Hemisphere, although the sign of the trend seems to agree...
in both data sets. Since STORM was driving by the NCEP/NCAR reanalysis, this un-
certainty is a strong caveat when estimating long-term upwelling trends. Unfortunately,
it is difficult to ascertain which data set (reanalysis, station wind records) is closer to
reality, as wind station records may also be strongly influenced by relocation of the
stations and instrumental inhomogeneities.

The SLP mechanisms later augmented by Bakun et al. (2010) involve small scale
dynamics of humidity transport, which may not be properly represented in the coarse
resolution atmospheric model used to generate the NCEP/NCAR reanalysis. Unfortunately,
no long instrumental SLP records spanning this region are available, so this
must remain an open question for the moment. Other possibility is that, although the
effect of the external forcing on the SLP field as envisaged by Bakun (1990) may be cor-
rect, the amplitude of internally generated wind variability at decadal and multidecadal
timescales blurs the long-term signal. Finally, as also proposed by Bakun et al. (2010),
the influence of other large-scale climate modes, such as ENSO, on BUS upwelling
may mask the forcing by anthropogenic greenhouse gases. Our correlation analysis
sheds some light on this last question, as explained below.

The statistically significant negative correlation between ENSO and upwelling in
South Benguela during the DJF has been reported by previous studies. Rouault et al.
(2010) stated that El Niño leads to reduced upwelling whereas La Niña enhances up-
welling. The same study argues that the upwelling-favourable winds are more pro-
nounced during La Niña and weaker during El Niño, which explains the SST anomaly
in the BUS during ENSO events. The connection between extreme SSTs (in the False
Bay, South Africa) and ENSO is also supported by the study by Deutsch et al. (2011)
and Dufois and Rouault (2012). The former authors also found significant correlations
between El Niño and the first principal component of austral summer, autumn and
partly spring SST (strongest with a 4 month lag, which is contained in the seasonal
resolution of our analysis).
The spectral analysis emphasized the connection between the upwelling and ENSO in DJF. The source of the other time scales in the spectrum (2.5, 3.3) could not be determined yet.

The link between upwelling in South Benguela and ENSO has been invoked to explain the parent lack of long-term trend in BUS upwelling. Earlier studies, for instance Vecchi et al. (2006), indicated a weakening of the trade winds across the Tropical Pacific (i.e. a tendency towards El Niño state) as a response to anthropogenic greenhouse gas forcing. A trend towards more intense or more frequent El Niño phases would thus tend to weaken Benguelan upwelling, counteracting the direct effect on upwelling of the anthropogenic forcing. However, more recent studies of the response of ENSO to anthropogenic greenhouse gas forcing recognize much larger uncertainties in the predictions of the future evolution of ENSO (Vecchi and Wittenberg, 2010), although one recent multimodel study has found that the frequency of extreme El Niño events may increase as a result of anthropogenic greenhouse gas forcing (Cai et al., 2014). Nevertheless, the model belonging to the Climate Model Intercomparison Project CMIP5, used by the Intergovernmental Panel on Climate Change, still underestimate the atmospheric, thermocline and advection feedbacks to ENSO (Kim et al., 2014). In the observations analysed here, the trend in the ENSO index since 1950 is weak and not significant, so that it is unclear how the long-term trend in ENSO could contribute to robust weakening of upwelling counteracting the greenhouse gas forcing in the past decades. Our analysis rather shows that the climate modes ENSO and the AAO could at most have influenced upwelling in South Benguela. The observed long-term trend in the indices of these two climate modes could explain almost 100% of the upwelling trend in DJF in South Benguela in the STORM simulation, but this contribution stems almost entirely from the AAO. In North Benguela, the influence of ENSO and the AAO is very weak and the evolution of the simulating upwelling there cannot be described well by a linear steady trend (Fig. 6) associated with the observed evolution of greenhouse gas forcing.
7 Conclusions

The large-scale atmospheric drivers of the Benguela Upwelling System, as simulated in a global high resolution ocean simulation (STORM) driven by a prescribed atmosphere in the last decades, have been described in this paper. The major results are summarized as follows:

- The BUS is better described by two subsystems, North Benguela and South Benguela, as their mean seasonality, their time evolution, and correlations with atmospheric drivers and large-scale climate modes differ.

- As a general characteristic for upwelling favourable atmospheric conditions in both subsystems are an intensified South Atlantic High, strong and southerly winds/wind stress along the coast, and a SLP contrast between land and ocean.

- There is some evidence of an influence of ENSO and the AAO on upwelling in South Benguela. El Niño phase in the Tropical Pacific tends to weakly hinder upwelling, whereas a stronger AAO tends to reinforce upwelling in this region. In North Benguela, there is no clear influence of these climate modes.

- The long-term trends of the simulating upwelling over approximately the last 50 years do not clearly support the hypothesis put forward by Bakun (1990) that anthropogenic greenhouse gas forcing should lead to more vigorous upwelling. The analysis of the trend of the SLP gradient between land and ocean as suggested by Bakun (1990) does not support an influence of the sea–land SLP contrast on the long-term behaviour of upwelling. The estimated influence of the trends of the large-scale climate modes on Benguela upwelling do not indicate that these climate modes may have disturbed the hypothesized connection between anthropogenic greenhouse forcing and upwelling in this region in the last decades.
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Table 1. Description of the data sets used in this study.

<table>
<thead>
<tr>
<th>Acronym</th>
<th>Data description</th>
<th>Time period</th>
<th>Grid [degree]</th>
</tr>
</thead>
<tbody>
<tr>
<td>NCEP/NCAR</td>
<td>reanalysis data</td>
<td>1948–2011</td>
<td>2.5</td>
</tr>
<tr>
<td>ERA-Interim</td>
<td>reanalysis data</td>
<td>1979–2010</td>
<td>0.75</td>
</tr>
<tr>
<td>HadISST1</td>
<td>global gridded observations</td>
<td>1870–2012</td>
<td>1</td>
</tr>
<tr>
<td>STORM</td>
<td>global ocean-only simulation/model MPI-OM</td>
<td>1950–2010</td>
<td>0.1</td>
</tr>
<tr>
<td>MEI</td>
<td>Multivariate ENSO (El Niño Southern Oscillation) Index</td>
<td>1950–2012</td>
<td></td>
</tr>
<tr>
<td>AAO</td>
<td>Antarctic Oscillation</td>
<td>1957–2012</td>
<td></td>
</tr>
<tr>
<td>ATL-3</td>
<td>Tropical Atlantic SST (HadISST1, STORM)</td>
<td>1870–2012 and 1950–2012</td>
<td>1 and 0.1</td>
</tr>
<tr>
<td>AMM</td>
<td>Atlantic Meridional Mode</td>
<td>1948–2012</td>
<td></td>
</tr>
<tr>
<td>HIX</td>
<td>St. Helena Index</td>
<td>1892–2012</td>
<td></td>
</tr>
<tr>
<td>QBO</td>
<td>Quasi-Biennial Oscillation</td>
<td>1979–2012</td>
<td></td>
</tr>
<tr>
<td>AVHRR</td>
<td>pathfinder 5.0, remote-sensed SST</td>
<td>1985–2009</td>
<td>4 km</td>
</tr>
</tbody>
</table>
Table 2. Coefficients of variation (CV, ratios between interannual SD and long-term mean, in %) and trends in upwelling intensity over the period 1950–2010 as simulated in the STORM simulation. The magnitude of the trends is expressed as total linear changes over the period 1950–2010 relative to the long-term mean. The figures in parenthesis bracket the nominal 95 %-uncertainty range in the trend estimation, assuming uncorrelated and gaussian distributed trend residuals.

<table>
<thead>
<tr>
<th></th>
<th>North Benguela CV</th>
<th>North Benguela trends</th>
<th>South Benguela CV</th>
<th>South Benguela trends</th>
</tr>
</thead>
<tbody>
<tr>
<td>MAM</td>
<td>24</td>
<td>26 (5, 46)</td>
<td>84</td>
<td>15 (−62, −91)</td>
</tr>
<tr>
<td>JJA</td>
<td>15</td>
<td>3 (−11, 17)</td>
<td>−72</td>
<td>−9 (−75, 56)</td>
</tr>
<tr>
<td>SON</td>
<td>16</td>
<td>−11 (−25, 3)</td>
<td>81</td>
<td>38 (−35, 110)</td>
</tr>
<tr>
<td>DJF</td>
<td>29</td>
<td>31 (10, 52)</td>
<td>48</td>
<td>39 (−3, 82)</td>
</tr>
</tbody>
</table>
Table 3. Correlation coefficients of the upwelling indices derived from the vertical velocity of STORM of North Benguela and South Benguela with ENSO, AAO, ATL-3 of HadISST1, ATL-3 of STORM, HIX, AMM, and QBO, significant correlations are highlighted by **.

<table>
<thead>
<tr>
<th></th>
<th>ENSO</th>
<th>AAO</th>
<th>ATL-3 (HadISST1)</th>
<th>ATL-3 (STORM)</th>
<th>HIX</th>
<th>AMM</th>
<th>QBO</th>
</tr>
</thead>
<tbody>
<tr>
<td>North Benguela:</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>MAM</td>
<td>0.26**</td>
<td>0.27**</td>
<td>−0.16</td>
<td>−0.21</td>
<td>0.06</td>
<td>0.08</td>
<td>−0.04</td>
</tr>
<tr>
<td>JJA</td>
<td>−0.13</td>
<td>−0.31**</td>
<td>0.02</td>
<td>0.00</td>
<td>0.22</td>
<td>0.14</td>
<td>−0.16</td>
</tr>
<tr>
<td>SON</td>
<td>−0.21</td>
<td>0.06</td>
<td>0.01</td>
<td>−0.19</td>
<td>0.18</td>
<td>−0.02</td>
<td>−0.14</td>
</tr>
<tr>
<td>DJF</td>
<td>0.01</td>
<td>0.29**</td>
<td>−0.04</td>
<td>−0.32**</td>
<td>0.33**</td>
<td>0.17</td>
<td>−0.02</td>
</tr>
<tr>
<td>South Benguela:</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>MAM</td>
<td>−0.26**</td>
<td>0.40**</td>
<td>0.04</td>
<td>−0.06</td>
<td>0.07</td>
<td>0.05</td>
<td>0.19</td>
</tr>
<tr>
<td>JJA</td>
<td>0.17</td>
<td>0.37**</td>
<td>0.03</td>
<td>−0.05</td>
<td>−0.06</td>
<td>−0.07</td>
<td>0.07</td>
</tr>
<tr>
<td>SON</td>
<td>0.23</td>
<td>0.32**</td>
<td>−0.06</td>
<td>−0.15</td>
<td>0.01</td>
<td>0.01</td>
<td>0.14</td>
</tr>
<tr>
<td>DJF</td>
<td>−0.46**</td>
<td>0.61**</td>
<td>0.13</td>
<td>−0.02</td>
<td>0.05</td>
<td>0.28**</td>
<td>0.04</td>
</tr>
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</table>
Table 4. Sea level pressure over land, over ocean, and gradient (ocean-land) trend in hPAy$^{-1}$ and its correlation coefficients with the upwelling index of STORM, significant correlations and trends are highlighted by **.

<table>
<thead>
<tr>
<th>Trend:</th>
<th>SLP (NCEP/NCAR)</th>
<th>SLP (ERA-Interim)</th>
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</thead>
<tbody>
<tr>
<td></td>
<td>Gradient</td>
<td>Land</td>
</tr>
<tr>
<td>MAM</td>
<td>$-0.02^{**}$</td>
<td>+0.02**</td>
</tr>
<tr>
<td>JJA</td>
<td>$-0.01$</td>
<td>+0.01</td>
</tr>
<tr>
<td>SON</td>
<td>$-0.01$</td>
<td>+0.02**</td>
</tr>
<tr>
<td>DJF</td>
<td>$-0.01$</td>
<td>+0.02**</td>
</tr>
</tbody>
</table>

North Benguela:

<table>
<thead>
<tr>
<th>Trend:</th>
<th>SLP (NCEP/NCAR)</th>
<th>SLP (ERA-Interim)</th>
</tr>
</thead>
<tbody>
<tr>
<td>MAM</td>
<td>+0.69**</td>
<td>+0.57**</td>
</tr>
<tr>
<td>JJA</td>
<td>+0.72**</td>
<td>+0.65**</td>
</tr>
<tr>
<td>SON</td>
<td>+0.42**</td>
<td>+0.56**</td>
</tr>
<tr>
<td>DJF</td>
<td>+0.51**</td>
<td>+0.65**</td>
</tr>
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</table>

South Benguela:

<table>
<thead>
<tr>
<th>Trend:</th>
<th>SLP (NCEP/NCAR)</th>
<th>SLP (ERA-Interim)</th>
</tr>
</thead>
<tbody>
<tr>
<td>MAM</td>
<td>+0.27**</td>
<td>+0.23</td>
</tr>
<tr>
<td>JJA</td>
<td>$-0.32^{**}$</td>
<td>$-0.33$</td>
</tr>
<tr>
<td>SON</td>
<td>+0.14</td>
<td>+0.07</td>
</tr>
<tr>
<td>DJF</td>
<td>+0.28**</td>
<td>+0.17</td>
</tr>
</tbody>
</table>
Figure 1. Map of the studied regions. The displayed domain is the area used for the atmospheric variables of NCEP/NCAR. The boxes indicate the areas used to define the upwelling indices and the SLP gradient. North Benguela: 8–30° E, 15–28° S (grey), South Benguela: 8–30° E, 28–40° S (violet), and the SLP regions (blue) over the ocean (20° W–10° E, 15–35° S) and over land (12–25° E, 10–20° S).
Figure 2. Long-term mean of the annual mean upwelling velocity at 52 m depth for two regions in the South Atlantic (a, b), the long-term annual mean simulated sea surface temperature (c, d), the long-term mean annual sea surface temperature of HadISST1 in the period (1870–2012) (e, f) and the simulated long-term annual mean surface (6 m depth) currents (g).
Figure 3. Long-term difference between the simulated SST in the STORM simulation and the AVHRR-derived SST in austral summer (DJF). Units are in Kelvin.
Figure 4. Annual cycle of the upwelling indices of STORM derived from the simulated vertical velocity at 52 m depth in North and South Benguela.
Figure 5. Correlation between upwelling indices derived from the vertical velocity and the SST field of STORM in austral summer (DJF) (North Benguela a, South Benguela b), and in austral winter (JJA) of STORM (North Benguela c and South Benguela d).
Figure 6. Time evolution of the seasonal upwelling indices (deviations from their long-term mean) in North Benguela (a) and South Benguela (b) derived from the vertical velocity of the STORM simulation in the period 1950–2010.
Figure 7. Correlation patterns between the simulated upwelling indices and the NCEP/NCAR wind stress for North and South Benguela and for DJF and JJA. The zonal and meridional components of the arrows represent the correlation with the corresponding component of the wind stress. The colours code the square root of the sum of these zonal and meridional correlations squared.
Figure 8. Correlation pattern between the downward wind stress from the NCEP/NCAR re-analysis and the Multivariate ENSO index MEI in DJF (a) and the Antarctic Oscillation (AAO) index in DJF (b), JJA (c) in the period 1950–2010. The zonal and meridional components of the arrows represent the correlation of MEI and accordingly AAO with the zonal and meridional component of the wind stress, respectively. The colours code the square root of the sum of these zonal and meridional correlations squared.
Figure 9. Spectral density of the upwelling indices derived from the vertical velocity in the STORM simulation in North Benguela (a) and South Benguela (b). The vertical black lines indicate the 90 % confidence interval.
Figure 10. Time evolution of the SLP gradient over the Benguela Upwelling System (SLP over land minus SLP over the ocean), derived from the NCEP/NCAR reanalysis data.