Atmospheric Patterns over the Antarctic Peninsula

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Abstract

Using clustering analysis for the SLP field of the ERA Interim reanalysis between 1979 and 2016, five synoptic pressure patterns have been obtained for Drake area and Antarctic Peninsula (AP) region (45ºS 75ºS 20ºW 120ºW). The five patterns have been named according their most important features as Low over the Wedell Sea (LWS), Low over the Amundsen and Bellingshausen Seas (LAB), Low over the Drake Passage (LDP), Zonal over the Drake Passage (ZDP) and Ridge over the Antarctic Peninsula (RAP). Each atmospheric pattern has been described after analyzing their development and evolution. A frequency analysis shows that the 5 atmospheric patterns present a similar annual frequency but a large seasonal variability. Their transitions from one to other pattern tends to follow a cycle in which synoptic atmospheric waves displaces eastwards a quarter-wavelength. Four of the five atmospheric patterns (except RAP) are very influenced by ENSO and SAM, specially LAB and LWS. Occurrence of LAB pattern presents a positive trend showing
agreement with other studies that indicates an enhancement of the Amundsen-Bellingshausen Sea
Low. Finally, atmospheric circulation patterns have been related with the airmass advection and
precipitation in Livingston Island showing the potential application to study the changes in the surface
mass balance on the AP cryosphere.

Key words: South Shetland Islands, Antarctic Peninsula, South Shetland Islands, atmospheric
patterns, climatology, Antarctica

1. Introduction

Ice mass lost by mountain glaciers and island ice caps is one of the major contributor to sea
level rise (IPCC 2013). Compared to the large ice sheets, thermal and dynamic response of these ice
masses are quicker, so they are more vulnerable to the global warming. Most ice caps are located on
islands around the Antarctic Peninsula (AP) (Bliss et al. 2013) and many studies have shown evidence
of their acceleration and thinning (i.e. Hock et al. 2009; Radić and Hock 2010, 2011; Gardner et al.
2013).

Ice cap thinning occurs as a response of the changing conditions in the AP region. The AP has
warmed +3.7 °C/century during the second half of 20th century (Vaughan et al. 2003), and although
there has been a significant cooling between 1998 and 2015 (Carrasco 2013; Oliva et al. 2017), Turner
et al. (2016) demonstrated that this period is consistent with the natural variability of the region, since
it is affected by long-term persistence (Ludescher et al. 2016). Indeed, the AP region is characterized
by a large interannual variability and its temperature is very sensitive to the state of both the El Niño-
Southern Oscillation (ENSO) (Fogt and Bromwich 2006; Fogt et al. 2011; Clem and Fogt 2013) and
the Southern Annular Mode (SAM) (Thompson and Solomon 2002; Van Den Broeke and Van Lipzig
2003; Marshall 2003). It has been shown that these climatological modes of variability modify the
main climatic patterns in Antarctica (i.e. Amundsen-Bellingshausen Seas Low (ABSL), jet streams,
etc.), suggesting that their change may be closely linked to synoptic pressure patterns (Carleton 2003).

Synoptic scale systems eventually determine the temperature and the moisture of the airmass
that contribute to the surface mass balance of the ice caps in the AP. Therefore, a further understanding of the synoptic climatology and its relationship with other climatological parameters may contribute to interpret the changes in the AP cryosphere.

Although the AP climatology has been extensively studied, only few studies have addressed the synoptic climatology and its classification. Kejna (1993) used low-level analysis in the AP region between 1986 and 1989 to make a classification according the air mass advection to H. Arctowski Station in King George island. Govorukha and Timofeyev (2002) made a visual analysis with a set of synoptic charts, satellite imagery and other operational information during three years between 1997 and 2000 to describe the main synoptic processes over the AP. Turner et al. (1998) examined satellite imagery and operational data to analyze synoptic-scale low pressure systems in the AP region during a year, classifying them according to the environment in which cyclogenesis took place.

These studies have contributed to improve the knowledge of the synoptic activity in the AP region, but until now, no one has conducted an objective synoptic classification during an extended period of time and using clustering techniques. This classification is addressed in this paper. Datasets and the methodology used are detailed in Chapter 2. In Chapter 3 synoptic classification is described, with an analysis of the trends and the relationship between the synoptic patterns found and different modes of variability. In Chapter 4 we conduct an analysis comparing the synoptic patterns with the air mass advected over Livingston island, discussing the possible effects on the surface mass balance on the island glaciers. In chapter 5 we draw some conclusions. Recently, Cohen et al. (2013) developed the objective synoptic classification for the Ross Sea region using the NCEP reanalysis. This study may be considered as its natural extension to the AP region.

2. Datasets and methodology

a) Cluster analysis

To obtain the synoptic pressure patterns in Drake area, we selected the mean sea level pressure (SLP) field of the ERA Interim reanalysis (Dee et al. 2011). ERA Interim has been stated as one of the most realistic reanalysis to describe the SLP and geopotential height at 500 hPa in Antarctic region (Bracegirdle and Marshall 2012). The area selected was 45ºS 75ºS 20ºW 120ºW (Figure 1a), and is
wide enough to represent the climate patterns in the surroundings of Drake area and the AP. This area may include few points located at high altitude in the Andean Mountains (South America) and Eternity Range (Antarctic Peninsula), that are not expected to affect the cluster analysis. By contrast, we carefully selected the area to not cover the Antarctic Plateau, an area with widespread high altitudes where SLP reduction is not valid. The period analyzed ranges from 1 Jan 1979 until 31 Dec 2016. For each day, we employed the 12 UTC field as the SLP field that better represent the daily synoptic pattern to compare with mean daily values of the automatic weather station (AWS) [see Section 2c].

The approach we used for clustering was similar to that used by Cohen et al. (2013). Prior to clustering we detrended the data by removing the mean SLP and weighted each data point by the square root of latitude. Thereafter, we retrieved the atmospheric patterns using the k-means cluster algorithm. This algorithm has been successfully used to do a cluster analysis in other regions (e.g. Lana et al. 2007; Houssos et al. 2008; Cohen et al. 2013) and consists in minimizing the average squared distance between all points in the same cluster to identify the centers of the cells. Single k-means algorithm is very sensitive to the initial values and may give unsatisfactory solutions by achieving local minimums. Thus, we improved the seeding algorithm employing the k-means++ method (Arthur and Vassilvitskii 2007) to guarantee to find optimal solutions. This methodology was performed for K from 3 to 15 and repeated 10 times for each K to ensure the reproducibility (with this methodology the 10 cluster fields obtained for each K were almost identical). As in Cohen et al. 2013, the final K set was selected by visual inspection of the center fields according to the expertise of Antarctic AEMET forecasters in the area, selecting K5 as the one that best reproduce the synoptic patterns of the AP area. The same procedure was also employed with the 500 hPa geopotential height fields obtaining a similar set of clusters for K5.

b) Climate Indices

In order to evaluate the climatological variability of the synoptic patterns retrieved, they were compared with two main climate modes of variability linked to the AP climate: Southern Annular Mode (SAM) and El Niño Southern Oscillation (ENSO).
SAM is the principal mode of variability of the southern hemisphere circulation and its changes has a large impact in the AP climate (Marshall et al. 2006). Surface temperatures over AP have been linked to the strengthening or weakening of circumpolar westerlies associated to the SAM (Thompson and Solomon 2002; van den Broeke and van Lipzig 2003; Marshall et al. 2006). Monthly station-based SAM index calculated by Marshall (2003) [available on-line at https://legacy.bas.ac.uk/met/gjma/sam.html] was used. This index is based on the zonal pressure difference between 40°S and 65°S calculated using 12 stations, and it extends back until 1957.

Variability in the AP temperatures has been also associated with the phase of ENSO by modulating the depth and the extension of the ABSL. We selected El Niño 3.4 index calculated from HadlSST1 dataset (Rayner et al. 2003) [available on-line at https://www.esrl.noaa.gov/psd/gcos_wgsp/Timeseries/Nino34/] as an index representative of the ENSO state (Bamston et al. 1997).

c) Station based data

In Section 4, we compare the synoptic patterns with the distribution of daily mean temperature, relative humidity and accumulated precipitation observed to study the air masses advected over the AP. Data from the Spanish Base Juan Carlos I AWS (JCI, 62.66°S 60.39°W) (Bañon and Vasallo 2015) in Livingston Island (South Shetland Islands) have been used to analyze the air mass advected to the South Shetland Islands for each synoptic type (Figure 1b). A dataset of mean daily values of temperature, relative humidity and precipitation accumulated from 2005 to 2016 was obtained if at least 80% of the daily 10-minutes values were available. To standardize the observations and fill the gaps the dataset has been homogenized against the daily values in another AWS located at Gabriel de Castilla Station (GdC, 62.98°S 60.68°W) in Deception Island, and the surrounding grid points of ERA Interim reanalysis (4 points located at the intersections between 62.25°S 60.75°W and 63.00°S 60.00°W) (Figure 1b). The homogenization was performed by means of the R package Climatol (Guijarro 2017), applied on the monthly aggregates to improve the break-point detection. Two significant break-points were detected in the JCI temperature series, and three in GdC. The dates of the break-points were refined with the aid of the meta-data of the stations before adjusting their
daily series. The same procedure was applied to the relative humidity (one break-point in JCI and two
in GdC) and precipitation (one break-point in JCI) series. Finally, the monthly means were subtracted
from daily means to obtain daily anomalies to compare them with the atmospheric patterns retrieved.

3. Synoptic Patterns in Maritime Antarctica

a) Description of synoptic patterns

Figure 2 shows the five SLP synoptic patterns (the centroid of each cluster) calculated between
1979 and 2016 from ERA Interim reanalysis. Besides evaluating the centroids, we considered to
analyze each pattern by examining some individual days (about 10 for each one) using Integrated
Data Viewer (UCAR/Unidata) software. Figure 3 presents some examples of synoptic maps and their
classification. For clarity, each cluster has been named according to the most important feature
presented as follows:

Low over the Weddell Sea (LWS)
This cluster shows the presence of a low east of Weddell Sea, produced either from the
maxima of cyclogenesis located in the southern part of the Weddell Sea (Simmonds and Keay 2000),
or from a low crossing the Drake Passage to the east. As a counterpoint, there is a presence of the
South Pacific subtropical anticyclone extending to the south east to the Bellingshausen Sea,
sometimes even coupling with the Continental Antarctic Anticyclone. This pattern presents an intense
meridian circulation from the south or southwest over the Antarctic Peninsula.

Low over the Amundsen and Bellingshausen Seas (LAB)
This cluster is characterized by the semi-permanent ABSL located around 100° W at the
This pattern is composed by wide quasi-stationary lows often surrounded by one or more small mobile
lows. East to the Drake Passage the South Atlantic subtropical anticyclone extends southwards to the
Weddell sea, being this pattern almost symmetric with LWS. This cluster transports warm and moist
air over the Antarctic Peninsula often associated to precipitation in the area (van Loon 1967).

Low over the Drake Passage (LDP)
This cluster aggregates the lows over and west to the Drake Passage. These lows may have
both a zonal path crossing the Drake Passage to the east, or a meridian path either northwards or
southwards as a result of a blocking high. Blocking highs are suggested in the cluster center by a slight ridge located over South Atlantic that sometimes couples with the Antarctic continental anticyclone.

**Zonal over the Drake Passage (ZDP)**

This cluster is characterized by an intense westerly flow over the Drake Passage and Tierra del Fuego with mobile cyclones moving eastwards at high latitudes along the Circumpolar Trough. Those cyclones crosses from Bellingshausen Sea to Weddell Sea through the AP and may suffer a variety of orographic processes on their paths (Mayes 1985).

**Ridge over the Antarctic Peninsula (RAP)**

This cluster presents a meridian circulation characterized by an anticyclonic ridge extending southward from South America to the Drake Passage and the AP associated to the wavenumber-3 pattern in the region (van Loon and Jenne 1972). This ridge is flanked by two quasi-stationary wide lows, which enhance the meridian circulation. Western low may be associated to a western displacement of the ABSL.

**b) Frequency of occurrence**

The five atmospheric patterns present a similar frequency all ranging between 18 and 22 % (Table 1). The most common pattern is ZDP with a frequency of 22.0% of days and the less common is LWS with 18.4% of days. There is a notable variability in pattern frequencies between seasons. For example, LAB shows an increase of frequency during the equinoxes, especially in the spring, associated to the Semi-Annual Oscillation (SAO) when the wavenumber-3 pattern of the circumpolar trough (one of the troughs located over Amundsen-Bellingshausen seas) contracts and deepens southeastward (van Loon 1967; van den Broeke 1998, 2000). ZDP pattern also presents a noteworthy frequency in spring when circumpolar trough is contracted and enhances the zonal winds around Antarctica (Raphael 2004). In summer, when the polar jet slightly moves northwards, LDP pattern becomes dominant. It is possible that the classification method associates systems north to the ABSL inside LDP pattern. This would explain the low occurrence of LAB in summer and the shift to the west of the LDP pattern in Figure 1. Autumn and winter present a more equilibrated distribution of
patterns, with a slight increase of LWS in winter associated to a larger density of cyclones near the Flichner-Ronne ice shelf (Simmonds and Keay 2000) and a most western position of the ABSL (Hosking et al. 2013).

c) Pattern persistence and sequences

Table 2 shows the pattern sequence between two consecutive days. The persistence is large and correspond to the 64.7% of the days. The high values of persistence of atmospheric patterns between two consecutive days indicates that the characteristic time for moving the different elements that configure each cluster is longer than a day. The most common daily sequences of patterns are the persistence of ZDP and LAB with 14.1% and 13.8% respectively. This is not surprising since they are the two most common patterns in the region. Nonetheless, persistence is longer in LWS and LAB, in which the mean number of consecutive persisting days is 3.2 and 3.1 respectively. RAP shows the smallest value of both frequency of persistence (11.3%) and mean number of consecutive days (2.5 days).

Non-persistence sequences correspond to the 35.3% of the days being each individual sequence below 4%. The most common sequences are the change from RAP to LAB, LWS to RAP and LAB to ZDP. It is not surprising that LWS to RAP (3.1%) and RAP to LAB (3.9%) along with LAB to LDP (2.7%) and LDP to LWS (2.6%) sequences are so common. These four sequences (LWS → RAP → LAB → LDP → LWS) constitute a chain where the synoptic waves move forward a quarter-wavelength. In fact, the opposite sequences are in general very uncommon since they imply a backward motion of the synoptic waves. Not surprisingly, there are no sequences between LWS and LAB and vice versa since they imply a complete inversion of the ridges and lows in a day.

LAB to ZDP (3.1%) is also a common transition that occurs when the low over the Bellingshausen Sea moves southeastward and crosses the Antarctic Peninsula. After this situation is also common the change from ZDP to LWS (2.8%) probably due to the orographic cyclogenesis leeside of Antarctic Peninsula (Mayes 1985). This chain (LAB → ZDP → LWS) may replace the chain LAB → LDP → LWS when the flow is mainly zonal over the Drake Passage, and explains why RAP to LAB has a larger frequency than the others. The complete cycle with these two chains is
outlined in Figure 4. The fact that all clusters are implicated into one or both transition chains may explain the similar frequencies of all circulation patterns.

e) Climate variability and linear trends

Correlations between annual and seasonal occurrence of atmospheric patterns with SAM and ENSO indices are shown in Table 3. Annual occurrence of most patterns significantly correlates with SAM, with the only exception of RAP that show a weak correlation ($p$-value < 0.1). Seasonal occurrences of patterns also show significant correlations with SAM with the exception of correlations with RAP in autumn, winter and spring, and with LWS in summer. A positive SAM increases the occurrence of ZDP and LAB since SAM is associated to the contraction and intensification of the circumpolar westerly belt that strengths zonal conditions (Lefebvre et al. 2004; Orr et al. 2008; Lubin et al. 2008) and it modulates the depth of the ABSL (Turner et al. 2013; Raphael et al. 2016; Clem and Fogt 2013). These two patterns contribute to bring mild air to Antarctic Peninsula. By contrast, positive SAM reduces the occurrence of LDP and LWS. SAM – LDP anticorrelation is consistent with other studies that show a reduction of the number of cyclones between 50º-60º S due to the poleward shift of their path during positive SAM conditions (Reboita et al. 2015).

ENSO has much lower influence in the atmospheric patterns over the AP region than SAM. Yet, there is a significant correlation between El Niño 3.4 SST anomalies and annual occurrences of LAB and LWS. At seasonal scale, El Niño 3.4 correlates with LAB in spring, summer and autumn. Through a tropospheric wave train, the cold (warm) phase of ENSO related to La Niña (El Niño) strengthens (weakens) the ABSL (Trenberth et al. 2002; Genthon and Cosme 2003), especially in spring when the correlations between ENSO and temperatures in the AP are strongest (Clem and Fogt 2013; Clem et al. 2016). El Niño 3.4 also correlates with LWS in winter and spring agreeing with the increase of the cyclonic density east to the AP during the positive phase of ENSO (Reboita et al. 2015, in their figure 5). It is worth to note that the correlations between ENSO or SAM and LAB occurrence are in agreement with the correlation between those teleconnections and central pressure of ABSL calculated by Hosking et al. (2013, in their Table 2).
Related to stratospheric ozone depletion and greenhouse gas increases, SAM has shown a positive trend since 1958 (Thompson and Solomon 2002; Marshall et al. 2004) specially in summer and autumn. This increase in positive SAM conditions has contributed to an increase of the temperature in the AP during the last 60 years (Marshall 2007) by intensifying cyclonic conditions in high latitudes (Thompson and Solomon 2002), enhancing lee-side foehn winds on the eastern Peninsula (Orr et al. 2008) and increasing meridional winds by amplifying the ABSL (Fogt et al. 2011, 2012; Turner et al. 2013; Hosking et al. 2013). Therefore, it is expected to find a change in the patterns occurrence associated to the increase of the SAM.

Table 4 shows the linear temporal evolution of frequency for each atmospheric pattern between 1979 and 2016. The only significant trends found at >95% significance are an increase of LWS in summer and an increase of LAB in autumn and all-year round. Weak trends are also observed in summer to increase ZDP and decrease LDP. The increase of LAB occurrences for both in all-year round and autumn may be associated to the increase of SAM (Hosking et al. 2013; Raphael et al. 2016). Increase of zonal conditions in summer may enhance lee cyclogenesis around Antarctic Peninsula (Mayes 1985; Orr et al. 2008) that would explain the increase of LWS occurrence in DJF.

4. Application to Livingston Island

South Shetland Islands, as well as the northern edge of the Antarctic Peninsula, are located near the center of the study area. Thus, it is expected that Livingston Ice Cap response may be influenced by the occurrence of the different circulation patterns advecting different air masses. This effect may be especially important in summer, when the ice caps are more sensitive to temperature changes. Indeed, Jonsell et al. (2012) have shown that Livingston Ice Cap surface mass balance is very sensitive to small changes in temperature during the melt season, calculating that 0.5 ºC increase results in 56% higher melt rates. They also noticed that high peaks in melt coincide with moist and warm fluxes arising from NW. This suggests that LAB pattern may produce those peaks in melt.

Figure 5 shows the bivariate distribution of daily temperature and moisture anomalies with respect of the monthly mean for each circulation pattern at JCI AWS in Livingston Island calculated
using a kernel density estimation. As expected, LAB is prone to transport warm and moist air to the Shetland Islands from south-east Pacific while LWS and RAP advects cold and dry air from the continent. Therefore, those two patterns may largely affect the Livingston Ice Cap surface balance being mainly negative during LAB conditions and stabilizing during LWS or RAP conditions. Both, ZDP and LDP presents mild temperatures, but whereas ZDP bears mainly moist air, LDP presents a dual behavior. LDP shows two peaks in the density function with different moisture. This behavior may be produced by the different air advection depending on the position of the low respect to the island. When the low is located west to Livingston, warm moist air is transported to JCI. When the low crosses to east, the air becomes cooler and dryer. Nonetheless, those relationships and the real effect into the cryosphere should be further explored.

Figure 6 shows the distribution of daily precipitation anomalies with respect the monthly mean in JCI AWS for each circulation pattern. Days with LAB pattern present more positive precipitation anomalies. ZDP in general shows positive precipitation anomalies although it also presents many days with dry anomalies. Most of the dry days were classified as RAP, LWS or LDP. It is worth to note that dichotomy presented by LDP for moisture is also conspicuous for precipitation, with two relative peaks in the density function for both positive and negative precipitation anomalies. Indeed, the largest anomalies of precipitation in JCI (the right tail of the plot) are not produced by LAB but for both LDP and ZDP patterns. This indicates that LAB does not determine the largest precipitation episodes but light drizzle events. Those results should be carefully considered due to the limitations of the rain gauges when snowfall is collected.

5. Conclusions

Five synoptic pressure patterns have been computed from ERA Interim reanalysis using cluster analysis in the AP region (45ºS 75ºS 20ºW 120ºW). All five types present a similar frequency during the year but they manifest a large seasonal variability as a result of changing climatological structures as SAO, polar jet movement, etc. Transition between synoptic patterns tends to follow a cycle where synoptic waves move eastwards a quarter-wavelength. This cycle has two modes, one
with larger latitudinal amplitude and meridional circulation, and another one with enhanced zonal circulation. Climate modes of variability as ENSO or SAM have a large influence in the frequency of patterns over the AP region. For example, as other studies suggest (i.e. Clem and Fogt 2013), cold phase of the ENSO in spring is associated with an enhanced ABSL and a reduction of the Weddell Low. All synoptic patterns show a larger correlation with SAM than with ENSO. During the positive SAM, the ABSL becomes strengthened and zonal circulation increases. By contrast, lows over Drake Passage and Weddell Sea prevails during negative SAM. Indeed, increase of LAB conditions may be explained by the increase of SAM as a result of stratospheric ozone depletion and greenhouse gas increases (Thompson and Solomon 2002; Marshall et al. 2004).

As pointed out by Cohen et al. (2013), there exists many potential uses to the synoptic patterns. In this paper, we conducted a cursory analysis to explore the possibility of determining the air mass characteristics associated to each pattern that may affect the surface mass balance of the ice cap on Livingston Island. Synoptic patterns have been shown to be a useful tool to study air mass advections that may affect the surface mass balance of ice caps and ice sheets in the periphery of Antarctica. Different patterns may carry different air masses. For example, LAB transports a warm and moist air and often slight precipitation to the South Shetland Islands while LWS is associated with dry and cold air. We suggest that changing of synoptic types occurrence, especially in summer may dramatically affect to the stability of glaciers and island ice caps. Future work will further explore the effects in the cryosphere with changing synoptic types.

Acknowledgments

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References


———, and N. P. M. van Lipzig, 2004: Changes in Antarctic temperature, wind and precipitation in


Table 1. Seasonal and annual frequency (%) of each synoptic pattern during the period comprised between 1979 and 2016.

<table>
<thead>
<tr>
<th></th>
<th>LWS</th>
<th>LAB</th>
<th>LDP</th>
<th>ZDP</th>
<th>RAP</th>
</tr>
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<tbody>
<tr>
<td>DJF</td>
<td>16.5</td>
<td>16.0</td>
<td>28.9</td>
<td>21.4</td>
<td>17.1</td>
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<td>MAM</td>
<td>17.4</td>
<td>20.6</td>
<td>20.8</td>
<td>19.4</td>
<td>21.8</td>
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<tr>
<td>JJA</td>
<td>23.0</td>
<td>19.3</td>
<td>17.5</td>
<td>19.5</td>
<td>20.6</td>
</tr>
<tr>
<td>SON</td>
<td>16.5</td>
<td>26.0</td>
<td>13.7</td>
<td>27.8</td>
<td>15.9</td>
</tr>
<tr>
<td>Total</td>
<td>18.4</td>
<td>20.5</td>
<td>20.2</td>
<td>22.0</td>
<td>18.9</td>
</tr>
</tbody>
</table>

Table 2. Frequency (%) of pattern sequences between two consecutive days during the period comprised between 1979 and 2016. Values in the diagonal (in bold) show the daily persistence for each synoptic pattern. Values in brackets show the mean consecutive days in which each pattern persist. Values in parenthesis show the frequency considering only the changing sequences, that is, when the persistence is removed.

<table>
<thead>
<tr>
<th>To pattern</th>
<th>LWS</th>
<th>LAB</th>
<th>LDP</th>
<th>ZDP</th>
<th>RAP</th>
</tr>
</thead>
<tbody>
<tr>
<td>From pattern</td>
<td>LWS</td>
<td>LAB</td>
<td>LDP</td>
<td>ZDP</td>
<td>RAP</td>
</tr>
<tr>
<td>LWS</td>
<td>12.6 [3.2]</td>
<td>0.0 (0.0)</td>
<td>0.9 (2.5)</td>
<td>1.8 (5.1)</td>
<td>3.1 (8.8)</td>
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<tr>
<td>LAB</td>
<td>0.0 (0.0)</td>
<td>13.8 [3.1]</td>
<td>2.7 (7.6)</td>
<td>3.1 (8.8)</td>
<td>0.9 (2.5)</td>
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<tr>
<td>LDP</td>
<td>2.6 (7.4)</td>
<td>1.0 (2.8)</td>
<td>12.9 [2.8]</td>
<td>2.0 (5.7)</td>
<td>1.8 (5.1)</td>
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<tr>
<td>ZDP</td>
<td>2.8 (7.9)</td>
<td>1.8 (5.1)</td>
<td>1.5 (4.2)</td>
<td>14.1 [2.8]</td>
<td>1.9 (5.4)</td>
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<tr>
<td>RAP</td>
<td>0.4 (1.1)</td>
<td>3.9 (11.0)</td>
<td>2.2 (6.2)</td>
<td>1.0 (2.8)</td>
<td>11.3 [2.5]</td>
</tr>
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</table>

Table 3. Correlations and statistical signification between seasonal occurrence of synoptic patterns and SAM and El Niño 3.4 indices during the period comprised between 1979 and 2016. p-values are indicated in parenthesis. Bolded values indicate statistically significant correlations at 95% confidence.

<table>
<thead>
<tr>
<th></th>
<th>Annual</th>
<th>DJF</th>
<th>MAM</th>
<th>JJA</th>
<th>SON</th>
</tr>
</thead>
<tbody>
<tr>
<td>SAM</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>LWS</td>
<td>(-0.44 (0.006))</td>
<td>(-0.07 (0.674))</td>
<td>(-0.62 (0.000))</td>
<td>(-0.39 (0.010))</td>
<td>(-0.45 (0.005))</td>
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<tr>
<td>LAB</td>
<td>(0.56 (0.000))</td>
<td>(0.32 (0.048))</td>
<td>(0.57 (0.000))</td>
<td>(0.57 (0.000))</td>
<td>(0.54 (0.000))</td>
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<tr>
<td>LDP</td>
<td>(-0.53 (0.001))</td>
<td>(-0.67 (0.000))</td>
<td>(-0.41 (0.010))</td>
<td>(-0.37 (0.020))</td>
<td>(-0.71 (0.000))</td>
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<td>ZDP</td>
<td>(0.49 (0.002))</td>
<td>(0.80 (0.000))</td>
<td>(0.64 (0.000))</td>
<td>(0.37 (0.023))</td>
<td>(0.42 (0.008))</td>
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<td>RAP</td>
<td>(-0.28 (0.094))</td>
<td>(-0.40 (0.011))</td>
<td>(-0.15 (0.379))</td>
<td>(-0.10 (0.538))</td>
<td>(-0.29 (0.080))</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>Annual</th>
<th>DJF</th>
<th>MAM</th>
<th>JJA</th>
<th>SON</th>
</tr>
</thead>
<tbody>
<tr>
<td>El Niño 3.4</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>LWS</td>
<td>(0.38 (0.017))</td>
<td>(0.30 (0.062))</td>
<td>(0.16 (0.350))</td>
<td>(0.33 (0.043))</td>
<td>(0.61 (0.000))</td>
</tr>
<tr>
<td>LAB</td>
<td>(-0.39 (0.015))</td>
<td>(-0.38 (0.018))</td>
<td>(-0.34 (0.038))</td>
<td>(-0.22 (0.183))</td>
<td>(-0.52 (0.001))</td>
</tr>
<tr>
<td>LDP</td>
<td>(0.12 (0.452))</td>
<td>(0.19 (0.257))</td>
<td>(0.06 (0.738))</td>
<td>(-0.14 (0.403))</td>
<td>(0.20 (0.219))</td>
</tr>
<tr>
<td>ZDP</td>
<td>(-0.20 (0.219))</td>
<td>(-0.16 (0.322))</td>
<td>(0.02 (0.908))</td>
<td>(0.05 (0.789))</td>
<td>(-0.01 (0.939))</td>
</tr>
<tr>
<td>RAP</td>
<td>(0.01 (0.941))</td>
<td>(-0.09 (0.579))</td>
<td>(0.10 (0.559))</td>
<td>(-0.12 (0.468))</td>
<td>(-0.07 (0.684))</td>
</tr>
</tbody>
</table>
Table 4. Annual and seasonal linear trends of synoptic pattern frequencies during the period comprised between 1979 and 2016. *p*-values are indicated in parenthesis. Bolded values indicate statistically significant trends at 95% confidence.

<table>
<thead>
<tr>
<th></th>
<th>Annual</th>
<th>DJF</th>
<th>MAM</th>
<th>JJA</th>
<th>SON</th>
</tr>
</thead>
<tbody>
<tr>
<td>LWS</td>
<td>-0.82</td>
<td><strong>2.72</strong></td>
<td>-2.43</td>
<td>-2.46</td>
<td>-0.42</td>
</tr>
<tr>
<td></td>
<td>(0.329)</td>
<td>(0.025)</td>
<td>(0.096)</td>
<td>(0.114)</td>
<td>(0.754)</td>
</tr>
<tr>
<td>LAB</td>
<td><strong>1.56</strong></td>
<td>0.40</td>
<td><strong>3.15</strong></td>
<td>-0.22</td>
<td>3.10</td>
</tr>
<tr>
<td></td>
<td>(0.047)</td>
<td>(0.678)</td>
<td>(0.009)</td>
<td>(0.868)</td>
<td>(0.086)</td>
</tr>
<tr>
<td>LDP</td>
<td>-0.79</td>
<td>-3.32</td>
<td>-0.24</td>
<td>1.18</td>
<td>0.09</td>
</tr>
<tr>
<td></td>
<td>(0.196)</td>
<td>(0.065)</td>
<td>(0.817)</td>
<td>(0.225)</td>
<td>(0.937)</td>
</tr>
<tr>
<td>ZDP</td>
<td>0.63</td>
<td>3.11</td>
<td>0.35</td>
<td>0.80</td>
<td>-1.33</td>
</tr>
<tr>
<td></td>
<td>(0.409)</td>
<td>(0.054)</td>
<td>(0.778)</td>
<td>(0.431)</td>
<td>(0.409)</td>
</tr>
<tr>
<td>RAP</td>
<td>-0.58</td>
<td>-1.20</td>
<td>-0.84</td>
<td>0.69</td>
<td>-1.16</td>
</tr>
<tr>
<td></td>
<td>(0.210)</td>
<td>(0.109)</td>
<td>(0.470)</td>
<td>(0.557)</td>
<td>(0.194)</td>
</tr>
</tbody>
</table>
Figure 1. a) Map of the geographic area showing the area used to perform the cluster analysis (red) and the area depicted in b (black). b) South Shetland Islands region indicating JCI station (in red), GdC station (in blue) and the four ERA Interim reanalysis points used to homogenize the data (in green).

Figure 2. Synoptic patterns (cluster centroids) calculated using ERA Interim reanalysis for the AP region over the period comprised between 1979 and 2016.

Figure 3. Examples of ERA Interim synoptic analysis in the AP for each cluster. LWS: 13 May 2002; LAB: 15 Sep 2008; LDP: 25 Feb 2003; ZDP: 19 Aug 2008; RAP: 30 May 2000.

Figure 4. Scheme of the most common transitions between synoptic patterns in the AP. The width of the arrows is proportional to the frequency of each transition.

Figure 5. Bivariate density function of daily temperature and relative humidity anomalies respect to the monthly mean for each synoptic pattern at JCI AWS. The period considered is comprised between 2005 and 2016. Each pattern displays a different scale to better visualize the peak of the function (shaded).

Figure 6. Density function of daily precipitation anomalies respect to the monthly mean for each synoptic pattern at JCI AWS. The period considered is comprised between 2005 and 2016.
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