Analysis of spatio-temporal changes in annual and seasonal precipitation variability in South America-Chile and related ocean–atmosphere circulation patterns

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ABSTRACT: Establishing relationships between coupled ocean–atmospheric patterns and precipitation accumulation is important to describe and predict spatio-temporal variability on annual or seasonal scales, and also to evaluate how this variability is influenced by global warming. The objective of this study was to examine the leading modes of interannual and seasonal (summer, autumn, winter, and spring) precipitation variability in South America-Chile, and their significant relationship to seasonally aggregated gridded data and climatic indices. Applying exhaustive data quality control measures to data from 238 rain gauges with different lengths of records between 1893 and 2013, a new data set was created with the objective of obtaining reliable records for further analysis. A comprehensive analysis through empirical orthogonal functions (EOF) allowed for determination of the leading modes of precipitation and their main spatial patterns for the whole country. The percentage of explained variance in the relationship between seasonally aggregated indices and the leading modes of precipitation confirmed that most of the interannual and winter precipitation variability in Chile is linked to the seasonal aggregation of El Nino Southern Oscillation (ENSO). The leading modes of summer, autumn, and spring precipitation were mostly linked to seasonal aggregations of the Madden and Julian Oscillation (MJO), and the Antarctic Oscillation (AAO).

KEY WORDS precipitation in Chile; annual and seasonal variability; EOF analysis; climate indices

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1. Introduction

Rainfall variability can be defined as the degree to which rainfall amounts vary spatially or temporally over a wide range of scales. Knowledge and understanding of such variability may lead to improve risk management practices in agricultural, industrial, and other human activities (Meinke et al., 2005; Trenberth et al., 2007; IPCC, 2012). This information can also help in predicting changes in the precipitation patterns associated with global warming (Hartmann et al., 2013).

Seasonal variability over the South American continent is evident in long-term monthly means of precipitation accumulation (e.g. Marengo et al., 2001; Garreau et al., 2009; Bombardi et al., 2014). During the austral winter, precipitation accumulation reveals numerous maxima located in the tropical regions along the east–west-oriented intertropical convergence zone (ITCZ) (Andrews et al., 2003; Marshall et al., 2014), while the central part of the continent experiences its dry season (Garreau et al., 2009). During the austral summer, an extensive area of heavy precipitation extends from the southern half of the Amazon Basin to northern Argentina. This southward shift of convection returns gradually to northern South America during austral fall (Garreau et al., 2009). This migration, defined by scientists as the South American monsoon system (SAMS), is considered the most important climatic feature in tropical and subtropical regions of South America (Zhou and Lau, 1998; Vera et al., 2006; Marengo et al., 2012; Carvalho et al., 2012). The extratropical regions of South America exhibit a marked zonal asymmetry, with wetter conditions to the west of the Andes cordillera. In contrast, forced subsidence over the eastern side of the Andes produces very dry conditions in Argentina’s Patagonia (Garreau et al., 2009). Mid-latitude storms tend to drift eastward along latitudinal bands known as storm tracks (Garreau and Aceituno, 2007). South of 40°S, low-level westerlies prevail year-round over the oceans and the continent, in connection with a mean poleward decrease in sea level pressure (SLP) and sea surface temperature (SST) (Figure 1). The mid-latitude westerlies extend through the entire troposphere, reaching a maximum speed (subtropical jet stream) in the upper troposphere (Figure 1(b) and (d)). This jet...
stream is commonly located between the 400 and 100 hPa levels (7–16 km) (Archer and Caldeira, 2008). During the austral winter, the low-level westerlies and the subtropical jet stream move equatorward. Thus, the area affected by mid-latitude precipitation over western South America expands up to about 30°S in austral winter (Figure 1(b) and (d)), and then retracts to the south of 40°S during austral summer (Figure 1(a) and (c)) (Garreaud et al., 2009).

All these climate features of South America play an important role in modulating the climate patterns in Chile. However, the exceptional topographic variation of the country is other crucial factor to take into account when analysing precipitation variability at different temporal scales. For instance, Chile is narrow (ranging from 90 to 445 km), constrained between the Andes and the coast of Pacific Ocean. In addition, its north–south extent, of
more than 4000 km, contributes to a wide variety of climates. The overall pattern shows that annual precipitation accumulation increases southward as well as windward of the Andes due to orographic effect (Garreaud et al., 2009; Pizarro et al., 2012; Quintana and Aceituno, 2012; Valdés-Pineda et al., 2014; Viale and Garreaud, 2015). However, the activity of the El Nino Southern Oscillation (ENSO), the Pacific Decadal Oscillation (PDO), the Antarctic Oscillation (AAO), and the Madden and Julian Oscillation (MJO) (for details, see Appendix S1, Supporting Information), in connection with the proximity to the South East Pacific Anticyclone (SEPA), the Antarctic Circumpolar Current (ACC), and the cold Humboldt Current (Peruvian Current), creates three general annual precipitation cycles in the country, northern, central, and austral (Rutllant and Fuenzalida, 1991; Vargas et al., 2000; Montecinos and Aceituno, 2003; Carrasco, 2006; Garreaud et al., 2009; Barrett et al., 2012; Quintana and Aceituno, 2012; Villalba et al., 2012; Ancapichún and Garcés-Vargas, 2015). In northern Chile (17°30′–30°S), an extremely hyper-arid climate is observed in the Central Valley (below 2300 m.a.s.l.), characterized by a lack of precipitation (less than 1 mm year−1), very low humidity, and the general absence of clouds (Houston, 2006). During the austral summer, wet episodes tend to occur throughout the Western Cordillera and Altiplano (plateaus with elevations greater than 3000 m). This occurs because strong upper-level easterly winds enhance moisture transport from Amazonia, creating saturation during uplift within deep convection cells (Garreaud et al., 2003). This climate feature is popularly known as ‘Bolivian Winter’; however, as previously mentioned, it is scientifically identified as the SAMS. The character of SAMS produces a rain shadow effect over the Western Cordillera and Atacama Desert, which is evidenced by the rapid decline of mean annual precipitation from over 300 mm year−1 at 5000 m.a.s.l. to less than 20 mm year−1 at 2300 m.a.s.l. (see also Vuille et al., 1998; Garreaud, 1999; Vuille, 1999; Garreaud, 2000; Garreaud and Aceituno, 2001; Garreaud et al., 2003; Houston and Hartley, 2003; Houston, 2006; Minvielle and Garreaud, 2011).

Central Chile (30°–40°S) has a well-defined annual cycle characterized by a peak of precipitation accumulation in the austral winter, and much lower values in the austral summer (Valdés-Pineda et al., 2015). This strong seasonality of precipitation results from the winter retreat of SEPA, which allows stronger low-level westerlies carrying frontal systems to the continent (for details, see Montecinos and Aceituno, 2003; Falvey and Garreaud, 2007). The annual accumulation increases southward from around 100 mm at 30°S to nearly 2000 mm at 40°S (Quintana and Aceituno, 2012), and is concentrated from May to September (Montecinos and Aceituno, 2003). Precipitation amounts also increase eastward due to the orographic effect (i.e. an altitudinal increase); in fact, the windward slopes of the Andes between 32° and 35°S can experience up to twice the precipitation observed over the western lowlands (Falvey and Garreaud, 2007; Viale and Garreaud, 2015).

In the austral region (40°–56°30′S), frontal cyclones drifting eastward along the oceanic storm tracks are responsible for most of the precipitation and day-to-day weather variations (Quintana and Aceituno, 2012; Garreaud et al., 2014). In general, precipitation exhibits a marked zonal asymmetry, with very wet (dry) conditions to the west (east) of the Andes cordillera (Garreaud et al., 2009). This pattern is consistent with strong westerlies throughout the year, which result in precipitation reaching more than 4000 mm year−1 and occurring in all four seasons along the western Andes (Carrasco et al., 2002; Villarroel, 2013), while producing a dramatic decrease to about 200 mm year−1 in Argentina’s Patagonia due to rain shadow east of the Andes (Aravena and Luckman, 2009; Garreaud et al., 2009).

Although there is already in-depth knowledge about the ocean–atmospheric processes affecting Chile, this study aims to detect how the leading modes of annual and seasonal precipitation (DJF for summer, MAM for spring, JJA for winter, and SON for autumn) are related to gridded data and different seasonally aggregated climate indices, i.e. the Southern Oscillation Index (SOI), the bivariate ENSO index (BEST), multivariate ENSO index (MEI), AAO, PDO, and MJO, among others. Specifically, this study examines how much interannual and seasonal variability can be explained by the leading modes of precipitation accumulation, whether any significant spatial patterns can be identified, and what seasonally aggregated climate indices can better explain the variability of the annual and seasonal leading modes of precipitation. The main aspects of the methodology, results, and discussion are in the main text, and further details about the methods and results of this study are presented in the Appendix S1.

2. Methodology

2.1. Study area, data set, and quality control

Continental Chile is located on the East Pacific Coast of South America between latitudes 17°30′S and 56°30′S (more than 4000 km), and is divided into 15 administrative regions. The geography of the country is dominated by steep mountainous terrains, with only around 20% of the territory being flat (Figure 2(a)). The territory includes 101 main hydrologic catchments, with more than 300 rivers flowing mainly westward into the ocean (Figure 2(b)). Chile is also characterized by a wide variety of landscapes. From a geomorphological point of view, it is possible to distinguish four major geographical units along the country: Andes Mountains (East), Intermediate Depression, Coastal Mountains, and Coastal Plains (West) (Valdés-Pineda et al., 2014).

The rain gauges data set was provided by the National Water Resources Directorate of Chile (DGA), and is available at http://www.dga.cl/Paginas/default.aspx. This public institution manages more than 3000 gauge stations in the country to measure different hydro-meteorologic variables. From this collection, a total of 671 operating rain gauges were used to determine their length.
of records and missed data (Figure 2(c)). A systematic data quality control process was conducted for each rain gauge, taking into account only those years with at least 10 months of data. In addition, for any period of analysis, only those rain gauges with at least 90% of annual/seasonal data were selected, and only for those records that contained at least 30 years of data. The annual/seasonal time series for each defined period were sorted from north to south and then assembled in matrix form \((A_{ij} = \text{accumulation for time } i \text{ at station } j)\). Then, K-means cluster analysis was used to determine the level of correlation between the series of every pair of stations as a criterion for clustering (i.e. see Spath, 1985). These levels of correlations and different clusters of stations were finally used to fill in missing data through the K nearest-neighbour method \((K_{\text{NN}})\), considering the ten nearest-neighbours with weights inversely proportional to the distances expressed by \(K_{\text{NN}} = \frac{1}{1-\rho}\), where \(\rho\) is the correlation coefficient from the neighbouring stations (Fix and Hodges, 1989). After this, the final consolidated data set contained records spanning between 1893 and 2012 (Figure 2(d), Table 1). For instance, the shortest selected period was 1979–2012 (34 years), which represented the largest spatial coverage.
and along the subtropical east coast of the continent variability in central Chile is linked to the PDO (Rutllant and
observed during extreme ENSO phases (Rutllant and
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either during austral summer – at the mature stage of
variability in the subtropical and extratropical regions
Maya (2003; Villalba et al. 2007). Regional and
seasonal accumulation for the year
accumulation patterns, such as those previously described in
Figure 3) also share significant relationships to those
observed during extreme ENSO phases (Rutllant and
Fuenzalida, 1991; Accentoun and Garreaud, 1995; Gar-
reaud and Battisti, 1999; Montecinos and Accentoun, 2003;
Garreaud et al., 2009). The interdecadal precipitation
variability in central Chile is linked to the PDO (Montec-
inos and Accentoun, 2003; Garreaud et al., 2009; Quintana
and Accentoun, 2012), and the intra-seasonal variability is
mostly modulated by the propagation of the MJO (Car-
sasco, 2006; Barrett et al., 2012). The AAO modulates
precipitation anomalies in southern Chile (largest at 40°S)
and along the subtropical east coast of the continent
(Garreaud et al., 2009; Villalba et al., 2012).

In keeping with study objectives, 13 Seasonal Climate
Indices were calculated from the original monthly records of
the SOI, MEI, Oceanic Nino Index (ONI), Trade Winds
Index (TWI), Niño 3.4 (N3.4), Trans Nino Index (TN1),
AAO, BEST, and PDO. For a detailed description of the
coupled ocean-atmosphere indices used in this research,
refer to Section I of the Appendix S1.) The daily values
of MJO were first aggregated at a monthly scale and
then at a seasonal scale. MJO is associated with ten particular
phases that represent different conditions as the MJO
oscillation progresses eastward from the Indian Ocean
through the Pacific Ocean (Madden and Julian, 1971, 1972;
Zhang, 2005; Barrett et al., 2012). Accordingly,
we only selected the phases 6, 7, 8, and 9 because they
represent enhanced convection conditions over the South
American continent. The seasonal averaged indices were
calculated as:

\[
S_{CI} = \frac{\sum_{i=1}^{n} CI_{ij}}{n}
\]

where, \(S_{CI}\) is the Seasonal Climate Index for a
single year \(k\). \(CI_{ij}\) is the Climate Index of the month \(i\)
(where \(i=1 \ldots 3\)), within the season \(j\) (where \(j=1 \ldots 4\),
and \(j_1 = DJF; j_2 = MAM; j_3 = JJA; \) and \(j_4 = SON\)). \(n\) is the
number of months within each season (\(n = 3\)).

2.3. Spatial relations and seasonal ratio index

As a way of accounting for any spatial effect in the accumu-
lation patterns, such as those previously described in
the Section 1, the relationship between annual/seasonal
precipitation and the geographic position (latitude, lon-
gitude, and elevation) of each rain gauge was initially
mapped and analysed. The spatio-temporal distribution
of seasonal precipitation accumulation (mm) compare to
annual precipitation accumulation (mm) was additionally
computed for each year and for each station using the sea-
sonal ratio index (SR) as: \(SR_i = \frac{S_{Acc}}{A_{Acc}}\), where \(S_{Acc}\) is
the seasonal accumulation for the year \(i\) (mm), and \(A_{Acc}\) is
the annual accumulation for the year \(i\). The SR for any
year can be represented as decimal numbers (or percentage)
ranging between 0 and 1 (0–100%). The length of the
\(S_{CI}\) time series varied: the longest was from 1854 to
present for the SOI (which was mainly used to detect pat-
terns of longest rain gauges series); the shortest were for
MJO and AAO oscillations, from 1978 and 1979 onward,
respectively.

2.4. Annual and seasonal SPI calculation

At each rain gauge, the annual and seasonal series of
precipitation accumulation were standardized by

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Table 1. Seasonal time-lagged correlations between PC1Annual and SCI for the period 1979–2012.

<table>
<thead>
<tr>
<th>Climatic index</th>
<th>AAO</th>
<th>SOI</th>
<th>PDO</th>
<th>MEI</th>
<th>ONI</th>
<th>TWI</th>
<th>N34</th>
<th>TNI</th>
<th>MJO6</th>
<th>MJO7</th>
<th>MJO8</th>
<th>MJO9</th>
<th>BEST</th>
</tr>
</thead>
<tbody>
<tr>
<td>SON(0) Correlation r</td>
<td>-0.17</td>
<td>-0.50</td>
<td>0.46</td>
<td>0.58</td>
<td>0.58</td>
<td>-0.30</td>
<td>0.56</td>
<td>-0.06</td>
<td>-0.52</td>
<td>-0.48</td>
<td>-0.36</td>
<td>0.14</td>
<td>0.56</td>
</tr>
<tr>
<td>r-Test</td>
<td>-0.99</td>
<td>-3.30</td>
<td>2.92</td>
<td>4.07</td>
<td>3.99</td>
<td>-1.77</td>
<td>3.86</td>
<td>-0.36</td>
<td>-3.41</td>
<td>-3.12</td>
<td>-2.20</td>
<td>0.80</td>
<td>3.85</td>
</tr>
<tr>
<td>JJA(0) Correlation r</td>
<td>-0.16</td>
<td>-0.63</td>
<td>0.19</td>
<td>0.59</td>
<td>0.61</td>
<td>0.00</td>
<td>0.58</td>
<td>-0.07</td>
<td>-0.60</td>
<td>-0.37</td>
<td>0.32</td>
<td>0.59</td>
<td>0.66</td>
</tr>
<tr>
<td>r-Test</td>
<td>-0.93</td>
<td>-4.65</td>
<td>1.09</td>
<td>4.19</td>
<td>4.39</td>
<td>0.02</td>
<td>4.06</td>
<td>-0.41</td>
<td>-4.19</td>
<td>-2.27</td>
<td>1.93</td>
<td>4.13</td>
<td>4.91</td>
</tr>
<tr>
<td>MAM(0) Correlation r</td>
<td>-0.26</td>
<td>-0.21</td>
<td>0.04</td>
<td>0.08</td>
<td>0.10</td>
<td>-0.02</td>
<td>0.10</td>
<td>-0.08</td>
<td>-0.21</td>
<td>-0.16</td>
<td>-0.07</td>
<td>0.13</td>
<td>0.22</td>
</tr>
<tr>
<td>r-Test</td>
<td>-1.55</td>
<td>-1.23</td>
<td>0.24</td>
<td>0.45</td>
<td>0.54</td>
<td>-0.09</td>
<td>0.58</td>
<td>-0.47</td>
<td>-1.23</td>
<td>-0.92</td>
<td>-0.41</td>
<td>0.74</td>
<td>1.25</td>
</tr>
<tr>
<td>DJF(0) Correlation r</td>
<td>-0.14</td>
<td>0.16</td>
<td>-0.02</td>
<td>-0.22</td>
<td>-0.18</td>
<td>0.31</td>
<td>-0.17</td>
<td>-0.20</td>
<td>-0.13</td>
<td>0.05</td>
<td>0.25</td>
<td>0.38</td>
<td>-0.15</td>
</tr>
<tr>
<td>r-Test</td>
<td>-0.82</td>
<td>0.93</td>
<td>-0.10</td>
<td>-1.28</td>
<td>-1.01</td>
<td>1.83</td>
<td>-0.99</td>
<td>-1.12</td>
<td>-0.73</td>
<td>0.27</td>
<td>1.47</td>
<td>2.34</td>
<td>-0.87</td>
</tr>
<tr>
<td>SON(0-1) Correlation r</td>
<td>0.17</td>
<td>0.16</td>
<td>-0.16</td>
<td>-0.24</td>
<td>-0.26</td>
<td>0.39</td>
<td>-0.27</td>
<td>-0.28</td>
<td>0.19</td>
<td>0.36</td>
<td>0.51</td>
<td>0.44</td>
<td>-0.19</td>
</tr>
<tr>
<td>r-Test</td>
<td>0.96</td>
<td>0.91</td>
<td>-0.93</td>
<td>-1.40</td>
<td>-1.50</td>
<td>2.39</td>
<td>-1.56</td>
<td>-1.67</td>
<td>1.11</td>
<td>2.20</td>
<td>3.31</td>
<td>2.80</td>
<td>-1.11</td>
</tr>
<tr>
<td>JJA(0-1) Correlation r</td>
<td>-0.04</td>
<td>0.35</td>
<td>-0.16</td>
<td>-0.24</td>
<td>-0.27</td>
<td>0.24</td>
<td>-0.29</td>
<td>-0.16</td>
<td>0.20</td>
<td>0.44</td>
<td>0.40</td>
<td>0.10</td>
<td>-0.28</td>
</tr>
<tr>
<td>r-Test</td>
<td>-0.20</td>
<td>2.12</td>
<td>-0.91</td>
<td>-1.42</td>
<td>-1.57</td>
<td>1.38</td>
<td>-1.71</td>
<td>-0.90</td>
<td>1.14</td>
<td>2.78</td>
<td>2.50</td>
<td>0.57</td>
<td>-1.67</td>
</tr>
<tr>
<td>MAM(0-1) Correlation r</td>
<td>-0.15</td>
<td>0.09</td>
<td>0.05</td>
<td>-0.05</td>
<td>-0.11</td>
<td>0.00</td>
<td>-0.13</td>
<td>-0.01</td>
<td>0.00</td>
<td>0.07</td>
<td>0.14</td>
<td>0.19</td>
<td>-0.11</td>
</tr>
<tr>
<td>r-Test</td>
<td>-0.85</td>
<td>0.50</td>
<td>0.30</td>
<td>-0.28</td>
<td>-0.62</td>
<td>0.00</td>
<td>-0.76</td>
<td>-0.07</td>
<td>0.00</td>
<td>0.38</td>
<td>0.78</td>
<td>1.09</td>
<td>-0.64</td>
</tr>
<tr>
<td>DJF(0-1) Correlation r</td>
<td>-0.21</td>
<td>-0.16</td>
<td>-0.02</td>
<td>0.05</td>
<td>-0.01</td>
<td>0.04</td>
<td>-0.01</td>
<td>0.07</td>
<td>-0.16</td>
<td>-0.29</td>
<td>-0.36</td>
<td>-0.20</td>
<td>0.09</td>
</tr>
<tr>
<td>r-Test</td>
<td>-1.23</td>
<td>-0.89</td>
<td>-0.12</td>
<td>0.29</td>
<td>-0.03</td>
<td>0.22</td>
<td>-0.05</td>
<td>0.40</td>
<td>-0.93</td>
<td>-1.72</td>
<td>-2.20</td>
<td>-1.16</td>
<td>0.49</td>
</tr>
</tbody>
</table>
using the 12-month Standardized Precipitation Index (SPI), which was calculated following the procedure of McKee et al. (1993). Conceptually, the SPI represents a z-score, which is the number of standard deviations above or below the zero-mean (see Section III of the Appendix S1). This index is the current technique used by the National Oceanic and Atmospheric Administration (NOAA) to monitor droughts in United States (http://www.cpc.ncep.noaa.gov/products/Drought/Monitoring/spi.shtml). Over the last decade, SPI has increasingly been used for assessment of drought intensity in many countries (Vicente-Serrano et al., 2004; Wilhite et al., 2005; Wu et al., 2006; Karavitis et al., 2011; Mwangi et al., 2014; Bonaccorso et al., 2015). When compared to other drought indices, the SPI has demonstrated favourable performance (Guttman, 1998; Keyantash and Dracup, 2002; Zarch et al., 2015). Vicente-Serrano et al. (2010) made a modification to create the Standardized Precipitation Evapotranspiration Index (SPEI), which is based on precipitation and temperature data. Bázrafshan et al. (2014) and Beguería et al. (2014) established the Multivariate Standardized Precipitation Index (MSPI), with which it is possible to aggregate a variety of SPI time series into new time series. In Chile, Verbiest et al. (2010) implemented the SPI Index approach in the Coquimbo Region (29°54’S latitude 70°15’W longitude) of Chile by calculating the index at seasonal scales, with the objective of designing a drought early warning system for risk management support. Currently, SPI is also used by the Chilean National Weather Service (www.meteochile.gob.cl/prediccion_estacional.php) and by the Drought Monitor System of the Ministry of Agriculture of Chile (http://www.climatedatalibrary.cl/UNEA/maproom/).

The SPI classification (see Section IV of the Appendix S1) proposed by the National Drought Mitigation Center of the University of Nebraska-Lincoln (http://drought.unl.edu/) consists of seven classes ranging from negative values for dry conditions (−2.0 ≤ SPI ≤ −1.0), to positive values for wet conditions (1.0 ≤ SPI ≤ 2.0). This classification was used in this study to evaluate the probability of observing very and extremely wet conditions P(SPI ≥ 1.5), or observing very and extremely dry conditions P(SPI ≤ −1.5), using data of El Niño and La Niña events recorded between 1979 and 2012 (238 rain gauges).

2.5. EOF analysis for annual and seasonal precipitation
Several empirical orthogonal functions (EOF) analyses or also called principal component analysis (PCA) have been conducted in South America (Chile) during the last few decades. Using infrared data from geostationary satellites (0.5° × 0.5° of spatial resolution), Garreaud and Wallace (1997) applied EOF analyses to seasonal mean 3-h anomalies of the cold cloud coverage (defined for each grid as the fraction of temporal samples that exhibited cloud-top temperatures colder than 235 K). Searching for dominant modes of the variability of convective cloudiness on synoptic timescales, Garreaud and Wallace (1998) applied EOF analysis to the daily anomalies of the Convective Index over South America. Montecinos and Aceituno (2003) analysed the seasonality of the ENSO rainfall in central Chile and associated circulation anomalies through a seasonal rainfall index defined as the first principal component (PC1) of standard deviation–normalized rainfall series. Aravena and Luckman (2009) used PCA and singular spectrum analysis (SSA) to analyse annual spatial and temporal patterns from a network of 23 rain gauges located in southern South America (40°S southward). Montecinos et al. (2011) used the first PCA of standardized rainfall anomalies during 24 non-ENSO winters in the period 1958–2000, calculated from records of 63 ground stations (30°–42°S), in order to analyse the non-ENSO interannual

Figure 3. Standardized time series (left panels) and scatter plots (right panels) of correlations between ENSO indices SOI, MEI, BEST, and precipitation in Santiago de Chile between 1948 and 2012.
Figure 4. Methodological steps used in this study to describe annual and seasonal precipitation patterns.

rainfall variability in central Chile during austral winter. Zamboni et al. (2012) used EOF analysis to determine seasonal variations and links between the interannual precipitation variability of South America and the South Pacific. Most of these studies were focused on analysing modes of interannual variability; therefore, a seasonal resolution would allow for the establishment of currently unidentified spatio-temporal patterns that could explain the seasonal precipitation variability in the country. Accordingly, in this study, we calculated the leading modes of annual and seasonal precipitation, and their degree of correlation to different climatic oscillations and other gridded data sets. (For a detailed description of EOF analyses carried out in this study, refer to Section V of the Appendix S1.) For instance, to study how the main modes of annual and seasonal precipitation variability (rain gauges) are associated with zonal winds, SLP, and SST, we projected each gridded field onto the PC1 (of annual and seasonal precipitation) in order to obtain the global signature (but mainly focused on the Pacific Basin) associated with each leading mode.

2.6. Significant leading modes of precipitation variability

In order to decide which EOF/PCA to retain (or discard), North et al. (1982) proposed a test to define typical errors between two neighbouring eigenvalues $\lambda$ (or eigenvectors). Accordingly, the North test was used to evaluate the statistical significance of each leading mode of annual/seasonal precipitation at the 95% level, assuming each year is independent. The statistical significance was calculated with to the following equation:

$$\Delta \lambda_i = \lambda_i \sqrt{\frac{2}{n}}$$  \hspace{1cm} (2)

where, $\Delta \lambda_i$ is the confidence interval at the 95% level of significance for the explained variance. $\lambda_i$ is the $i$th eigenvalue. $n$ is the sample size.

Then the first ten eigenvalues with error bars (scaled as percentage of explained variance) were plotted for annual/seasonal precipitation patterns in order to determine their significant modes of annual and seasonal precipitation patterns.

2.7. Climate oscillations and their relationship to leading modes of precipitation

Each significant leading mode of annual and seasonal precipitation (PC) was correlated to precipitation anomalies at each rain gauge, and also using gridded data from NOAA (see Section 3.7). The statistical significances were evaluated using the $t$-test ($\alpha = 0.05$). This analysis allowed for the determination of the different point- and gridded-based spatial patterns that dominate annual and seasonal precipitation variability in Chile. In addition, seeking to determine which seasonally aggregated climate indices were significantly related to these spatial patterns, a seasonal time-lagged correlation coefficient was also calculated for the relationship between each significant PC ($S_{pc1} = PC_1, PC_2, PC_3, \ldots, PC_n$) and each seasonally aggregated index ($S_{cli} = S_{CI1}, S_{CI2}, S_{CI3}, \ldots, S_{CIk}$), considering up to 1-year back of correlations (four seasons). The significances of these correlations were also
calculated using a $t$-test ($\alpha = 0.05$). The point of applying the time-lagged correlation analysis was to identify which seasonally aggregated index was significantly influencing annual/seasonal precipitation variability in Chile. The methodological steps used in this study are summarized in Figure 4.

3. Results

3.1. Spatial patterns of precipitation accumulation

The spatial distribution of average annual and austral summer precipitation accumulation in Chile shows a marked latitudinal decrease in Northern regions with an evident North-to-South gradient (Figure 5(a) and (b)). This latitudinal gradient goes from nearly 400 mm year$^{-1}$ over the Altiplano ($\sim 18^\circ$S) to about 20 mm year$^{-1}$ in the Atacama Desert Region ($\sim 27^\circ$S). This is due to the fact that the most-northern regions receive a larger contribution of precipitation during the austral summer, caused by strong convection and daily moisture movement from the Amazonia during SAMS (see also Fuenzalida and Ruttlant, 1987; Horel et al., 1989; Garreaud and Wallace, 1997; Vuille et al., 1998; Garreaud, 1999; Vuille, 1999; Garreaud, 2000; Garreaud and Aceituno, 2001; Garreaud et al., 2003; Houston and Hartley, 2003; Houston, 2006; Minvielle and Garreaud, 2011). A second latitudinal gradient is observed from 27$^\circ$S southward, along Central Chile (Falvey and Garreaud, 2007), as a result of the North–south movement of westerlies (see Rahn and Garreaud, 2014). This pattern reaches its annual and seasonal maxima between 40$^\circ$ and 45$^\circ$S (see Figure 6), where precipitation accumulation can increase up to more than 150 times. For instance, more than 3000 mm year$^{-1}$ (on average) can be registered in Northern Patagonia; this total is typically distributed as $\sim 500$ mm in summer; $\sim 800$ mm in autumn; $\sim 1000$ mm in winter; and $\sim 700$ mm in spring. However, some rain gauges can even register more than 4000 mm year$^{-1}$ along the western Andes (as reported by Carrasco et al., 2002; Villarroel, 2013; Viale and Garreaud, 2015; Valdés-Pineda et al., 2015). This large amount of precipitation is due to the fact that south of 40$^\circ$S the low-level westerly flow prevails year-round and most of the rainfall is produced by deep stratiform clouds that develop along warm and cold fronts (Garreaud et al., 2009). A third latitudinal gradient reveals a decrease...
southward, reaching minimum annual accumulations of \( \sim 250 \) mm in those rain gauges located between 51\(^\circ\) and 53\(^\circ\)S (Figure 5(a)). Because the low-level westerlies expand equatorward during austral winter, but weaken predominantly in Patagonia south of 50\(^\circ\)S (see Figure 5(b) and (d)), the larger contribution to annual precipitation is mostly observed during austral autumn (up to 1.8 times larger) rather than austral winter (Figure 5(c) and (d)). This latter season even registers similar amounts to those recorded during the austral summer and spring; i.e. seasonally averaged precipitation accumulation ranging between \( \sim 50 \) and \( \sim 150 \) mm (Figure 5(b) and (e)). These results also are consistent with the analysis of Carrasco et al. (2008) on the basis of average monthly distribution of precipitation in Chile.

The described North-to-South (down-up-down) latitudinal gradient of annual and seasonal precipitation accumulation can be clearly shown through scatter plots developed for the whole country (upper plots in Figure 6). For instance, a linear decay is observed in Northern Chile that is stronger during austral summer (\( r = 0.74 \)) but weakens during austral winter (\( r = 0.55 \)), and remains more or less constant during autumn and spring. Then an exponential growth is evident along the Central part of the country; it is stronger at annual scale (\( r = 0.95 \)) but also noticeable at seasonal scale (\( r \geq 0.90 \)). In the Southern regions, a linear decrease is more evident during austral winter (\( r = 0.84 \)), but also significant during the other three seasons (\( r \geq 0.80 \)). On the other hand, larger variability was noticed for the relationships between precipitation accumulation and West longitude or altitude (middle and lower plots of Figure 6). This variability can be attributed to the Andean orography, which varies intensely along the country and generates different local conditions of orographic enhancement (OE) that cannot be clearly detected when the data are arranged at a national scale. Accordingly, the Andean orographic influence on precipitation accumulation is revised in detail in the following section.

3.2. Orographic effects on precipitation accumulation
The average height of the Chilean Andes surpasses 4000 m.a.s.l. in the subtropical latitudes (18\(^\circ\)–35\(^\circ\)S), and the average width of the range is less than 200 km (Figure 7(a) and (b)). The average height rapidly decreases southward, dropping to less than 1500 m.a.s.l. in the extratropical latitudes (40\(^\circ\)–55\(^\circ\)S); in that area the width can increase up to about 400 km, forming an elevated plateau (Garreaud et al., 2009; Viale and Garreaud, 2015). In Northern Chile, most of the annual precipitation accumulation originates from mid- and upper-level easterly winds that transport moist air from Amazonia to feed convective storms over the eastern Andes during the austral summer (Minvielle and Garreaud, 2011). As a consequence, drier conditions are observed on the western side of the Andes, where precipitation accumulation can decrease more than 20 times due to rain shadow. This altitudinal gradient is highly linear between 2000 and 4500 m.a.s.l. (17–21\(^\circ\)S) at annual and seasonal scales, being stronger during austral summer and autumn but also significant during austral winter and spring (Figure 7(a)).

In Central Chile, westerlies carrying frontal systems to the continent are blocked by the Andes, generating an OE which is evident at annual and seasonal scales.
Figure 7. Topographic transects (a–e), for different subtropical and extratropical latitudes in Chile (left-map). These transects are represented on the right plots as elevation profiles (black). The left axis represents the elevation in metres above sea level (m.a.s.l.), while the upper axis is annual or seasonal accumulation in millimetres (mm), represented as scatter plots for those rain gauges distributed in the following bands: (a) 17°–21°S; (b) 29°–31°S; (c) 34°–37°S; (d) 44°–45°S; and (e) 51°–53°S. Black arrows represent mean annual 925-hPa zonal and meridional wind (m s\(^{-1}\)) components (U\(_{925}\); V\(_{925}\)).

(see Falvey and Garreaud, 2007; Viale and Garreaud, 2015). The enhancement rates (ER) associated with annual accumulation can reach to more than four times between 0 and 2000 m.a.s.l. The seasonal ER ranges from three during austral winter to more than five during autumn and spring (Figure 7(b) and (c)). During austral summer, precipitation accumulation can surprisingly increase more than 15 times in the cross-mountain direction at about 30°S (Figure 7(b)). This ER is much more significant south of 35°S (ER > 20) since a latitudinal increase is observed in summer precipitation (Figure 7(c)).

In Southern Chile, the monthly precipitation variability is strongly negatively (positively) correlated with 850-hPa zonal wind speed in eastern (western) Patagonia. This means that a year (or season) with stronger than average low or mid-level westerly flow features increased precipitation to the west of the Andes and decreased precipitation over the lowlands to the east due...
to rain shadow (Garreaud et al., 2013). This pattern is clearly detected from the group of stations distributed around 45°S (0–100 m.a.s.l.), and those located around 52°S (0–250 m.a.s.l.) where a strong (linear) altitudinal decrease is observed because a larger proportion of precipitation is registered over the Pacific Ocean at annual and seasonal scales (Figure 7(d) and (e)).

3.3. Seasonal ratio index

The averaged $SR_i$ calculated for those rain gauges distributed along the band 18°–25° reveals that between 60 and 80% of the annual accumulation occurs during the austral summer. Along the regions located between 25° and 45°S, summer precipitation consistently contributes less than 20%; however, this percentage is slightly exceeded in stations located in the band 50°–55°S (Figure 8(a)). The autumn contribution is concentrated between 20 and 40% for most of the country (Figure 8(b)). During the austral winter, precipitation accumulation is commonly less than 20% of annual precipitation for most of the stations located in northern regions ($\leq 25^\circ$S). An increase in winter contribution is then observed between 25° and 30°S, reaching a peak of about 80% at this latter latitude. Then the winter contribution gradually decreases to less than 20% in some cases in southern rain gauges along the band 50°–55°S (Figure 8(c)). Lower spring contribution ($SR_i \leq 20\%$) is observed in those rain gauges located north of 40°S, and higher spring contribution ($\sim 20\%$) was observed in stations located south of 40°S (Figure 8(d)). Temporal trends of $SR_i$ calculated for all rain gauges (Mann–Kendall test) revealed significant positive (negative) changes during austral summer (spring) between 1979 and 2012 for some rain gauges in Northern Chile (Figure 8(e) and (h)). These changes are more noticeable in Central and Southern Chile for those rain gauges located around 40°S, which showed mostly significant negative (positive) trends during austral autumn (winter) (Figure 8(f) and (g)). Seasonal changes have recently been reported for United States by Allen and Sheridan (2015), who analysed temperature data concluding that late starts of autumn and winter have been observed while earlier onsets of spring and summer have taken place between 1948 and 2012.

3.4. SPI and ENSO conditions

Results for northern Chile have to be taken with caution, because the large number of years with zero precipitation in the records causes some wet years to easily exceed very and extremely wet conditions (SPI $\geq 1.5$) as defined by the SPI classification (Figure 9(a)–(c)). In central and southern Chile, SPI can adequately represent the drought conditions. For instance, the ‘Great Drought’ occurring in central Chile between 1968 and 1969 is clearly represented by records of more than 100 years, showing negative values of SPI larger than extremely dry conditions (Figure 9(d)–(f)). The recent 2007–2008 drought
event that affected most of the country is also evident in the records of southern rain gauges, where extremely dry conditions are clearly identified (Figure 9(g)–(j)).

Recent trends and ENSO influence on droughts in Chile between 29° and 32°S demonstrate that extreme dry events (calculated by SPEI) show negative temporal trends particularly during the austral spring and summer, which are strongly associated with a trend towards increasing drought frequency (Meza, 2013). Regional Frequency Analysis coupled with L-moments has additionally showed that droughts with 40% of the normal precipitation are associated with return periods of 4 years at around 29°S to 22 years at about 35°S (Núñez et al., 2011). In accordance with previous research, this temporal pattern of above average precipitation (droughts) in central Chile should be closely related to El Niño (La Niña) conditions (see Figure 9), which are linked to weakening (strengthening) of the subtropical anticyclone allowing for the shift (blocking) of frontal storms to more northern locations. When assessing the probability of observing extreme hydrologic conditions (calculated by SPI) during strong and very strong ENSO phases, our results indicate that there is a greater probability (10–30%) of observing very wet and extremely wet conditions in Central Chile during El Niño years (Figure 10(a)). This pattern can also be observed in those rain gauges located at around 25°S due to the northward displacement of westerlies during El Niño conditions, i.e. the Antofagasta region as reported by Vargas et al. (2000, 2007). On the other hand, the lowest probability of observing very dry and severely dry conditions (SPI ≤ −1.5) during El Niño years is concentrated in central Chile (P ≤ 5.0%), with greater probabilities concentrated in southern Chile (Figure 10(b)). During La Niña years, the probability of observing very wet and extremely wet conditions is greater at around 20°S and also in the rain gauges located south of 45°S (Figure 10(c)). Intense dryness is mainly observed in central Chile during La Niña years, where the probability of very and severely dry conditions (SPI < −1.5) is greater than that observed for the rest of the country (P ≥ 20%) (Figure 10(d)). Although these results are consistent with previous studies, it is worth mentioning that there is a potential impact associated with the limitations of Standardized Drought Indices (SDI), which are derived by the influence of multi-decadal climate variability (Núñez et al., 2014). Therefore, the relationship between a future national drought policy and whichever indices are used to make it effective will have to account for this variability.

3.5. Leading modes of annual and seasonal precipitation

Strong leading modes of annual and seasonal precipitation variability were detected at all temporal scales (1979–2012). For instance, the first principal component of annual precipitation (PC1_annual) for the whole country is significant, explaining 55% of the total annual variance. According to the approach proposed by North et al. (1982), the second and third leading modes of annual precipitation are not well separated from the subsequent modes; therefore, they are not significant at the 95% confidence level (Figure 11). Similar results were found by Pezoa (2003), who used PCA to study the spatial patterns of precipitation from 57 Chilean stations between 33° and


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Precipitation and Ocean-Atmosphere Circulation Patterns in Chile

Figure 10. (a) and (c) Probability (%) of having very to extremely wet conditions $P(\text{SPI} > 1.5)$; (b) and (d) show the probability of having very to extremely dry conditions $P(\text{SPI} < -1.5)$, associated with El Niño (a and b) and La Niña (c and d) conditions (1979–2012).

53°S (1962–2001). The first principal component (PC1) of mean annual precipitation explained 61% of the total variance and was associated with stations north of 40°S. A second PC (12%) was associated with stations around 45°S, and a third PC (5%) was associated with the Punta Arenas station (~53°S). Results of PCA for southern South America showed very strong spatial patterns for the annual rainfall records (1961–2000), depicting six significant PCs that contained more than 78% of the total variance, and a PC1 explaining 21% of the total variance (Aravena and Luckman, 2009).

The leading modes of summer precipitation showed significant components with explained variances of 46.2, 20.1, and 8.7%, respectively (Figure 12(a)). Similar results were found for autumn and winter precipitation patterns (Figure 12(b) and (c)), demonstrating at least...
Figure 11. Eigenvalue spectrum calculated through North Test (left panel), and time series (right panel) of the first three leading modes (principal components PC1, PC2, and PC3) of annual precipitation for 238 raingauges (1979–2012).

Figure 12. Eigenvalue spectrum of the covariance matrix showing the ten eigenvalues (%), i.e. explained variance of the first ten EOF/PC (modes) calculated for summer (DJF), autumn (MAM), winter (JJA), and spring (SON) precipitation accumulation (238 stations for the 1979–2012 period).

Three significant modes in each season. During the austral spring, precipitation patterns were similar to those in the annual analysis since only the first leading mode of precipitation (PC1_{Spring}) was found to be significant. This mode also represented the smallest portion of explained variance (Figure 12(d)). Montecinos and Aceituno (2003) reported larger explained variances by calculating the PC1 of monthly standardized records of rainfall (1958–1999). The authors found explained variances of 82% (winter), 84% (late spring), and 78% (summer), for a total of 49
rain gauges distributed in three different clusters in central Chile. However, during non-ENSO winter seasons, the explained variance increases, because the PC1 of standardized rainfall anomalies during those conditions can explain up to 91% of the total precipitation variability (Montecinos et al., 2011).

### 3.6. Anomalies and leading modes of precipitation

Correlation maps between annual (seasonal) anomalies at each rain gauge and significant leading modes of precipitation revealed varied results. For instance, the first leading mode of annual precipitation (PC1Annual) shows positive and significant correlations between 20° and 40°S; however, a strong significant PC1Annual pattern is mainly concentrated in Central Chile (see also Pezoa, 2003), where the largest significant (positive) correlations can be observed between annual anomalies of precipitation and PC1Annual. Only one rain gauge located in Southern Chile (south of 50°S) showed a significant relationship for PC1Annual, which is negatively correlated to annual precipitation anomalies (Figure 13(a)). These most southern regions of Chile have been described by Aravena and Luckman (2009) who found four regional patterns of annual precipitation (1950–2000): Northwestern Patagonia (41°–44°S), Central Patagonia (45°–47°S), Patagonia plain-Atlantic (43°–50°S), and Southern Patagonia (51°–53°S).

At a seasonal scale, the leading mode of summer precipitation (PC1Summer) revealed mostly negative and significant correlations in Central Chile (Figure 13(b)). On the contrary, the first leading mode of autumn precipitation (PC1Autumn) showed positive and significant links in central Chile, with the strongest activity (r ≥ 0.85) around 35°S (Figure 13(c)). The significant mode of winter precipitation (PC1Winter) showed similar patterns to those observed at an annual scale: positive and significant correlations concentrated mainly between 20° and 40°S but strongest between 31° and 35°S (i.e. Montecinos et al., 2000a, 2000b), and negative correlations concentrated between 45° and 55°S (Figure 13(d)). The relationship between spring anomalies and PC1Spring proved to be significant in Central Chile, and also (but to a lesser extent) for some rain gauges located in the Northern (~20°S) and Southern regions (~53°S) (Figure 13(d)).

### 3.7. Gridded fields and leading modes of precipitation

Gridded data sets from the NOAA Earth System Research Laboratory were additionally used to compare and associate our previous results. [The gridded data sets are: (1) University of Delaware monthly precipitation (V3.01), 0.5°×0.5° (latitude/longitude grid) provided by the NOAA/OAR/ESRL PSD. (2) Monthly U-winds (850 hPa) and monthly sea level pressure (SLP), 2.5°×2.5°, provided by NCEP/NCAR Reanalysis (see Kalnay et al., 1996). (3) Monthly sea surface temperature (COBE-SST2) data, 1.0°×1.0°, provided by the NOAA/OAR/ESRL PSD (see Hirahara et al., 2014).] In this regard, the leading modes of precipitation in the...
Figure 14. Significant correlation maps calculated using t-test, between PC1\textsubscript{Annual}, PC1\textsubscript{Seasonal}, and the precipitation anomalies.

Chilean domain (15°–60°S and 65°–80°W) denoted similar spatial patterns to those observed for the rain gauges, but not at all scales. For instance, the strongest significant correlations between annual precipitation anomalies and the PC1\textsubscript{Annual}, PC1\textsubscript{Winter}, and PC1\textsubscript{Spring} were concentrated between 35° and 40°S (Figure 14(a), (d), and (e)). During austral summer, there is a southward displacement of this pattern (to about 40°S), that is not so evident at the rain gauge-scale; it is probably related to the strengthening of SEPA, which blocks the passing of frontal storms in more northern locations (Figure 14(b) and (c)).

Examining the relation between PC1 and gridded fields of zonal winds at 850 hPa, SLP and SST revealed that high correlations are mostly restricted to Southern South America and the adjacent Pacific Ocean (Figure 14); however, substantial differences between summer and winter seasons are also evident. For instance, there are strong opposite correlations in different zonally elongated bands of U-winds and SLP extending across South America which are noticeable at an annual scale (Figure 14, upper maps). These bands denote a spatial pattern of U-winds that is (strongly) negatively correlated in central and southern Chile during austral winter, but positively correlated during austral summer (Figure 15, left maps). Similar patterns are observed for the seasonality of SLP and SST (Figure 15, middle and right maps), which also exhibit opposite correlations between mid- and high latitudes of Chile. This confirms that precipitation variability is clearly associated with pressure and temperature changes over the equatorial and austral Pacific Basin that can also be observed at interannual and seasonal scales. These results are in agreement with those obtained by Garreaud et al. (2013), who showed that the leading mode of the 850-hPa zonal winds calculated for Southern South America (35°–65°S, 90°–60°W) indicates two zonal bands spanning the whole Southern Hemisphere (SH). The results of this analysis demonstrate that the use of gridded data is an effective means of analysing climatic patterns in Chile; however, the complex geography of the country still
requires high-resolution products that can both address the narrow distances from the Andes to the Pacific Ocean and record information over the extreme South regions of the country.

3.8. Climate indices and leading modes of precipitation

Analysing the ability of each seasonally aggregated coupled ocean–atmosphere index to predict precipitation variability, it is evident that the PC1 _Annual_ variability is best explained by ENSO indices, especially by the negative phases (El Niño years) of the winter SOI (SOIWinter) \((r = -0.61)\), and even more so \((r = 0.66)\) by the phases of winter Bivariate ENSO Index (BEST Winter) (Figure 16(a) and (b)). Despite this, other ENSO indices also show significant correlations with the precipitation amounts, i.e. the mean annual values of MEI as reported by Garreaud _et al._ (2009). The phases of seasonally aggregated MJO are strongly correlated to PC1 _Annual_, but this relationship is not as strong as that observed for ENSO Indices. In general, the best relationships between climate indices and PC1 _Annual_ were observed for the winter season (same year); the exceptions are those indexes related to MJO, which showed significant correlation with summer (same year) and spring seasons (1-year back) (Table 1).

The significant modes of austral summer and spring precipitation (PC1 _Summer_ and PC1 _Spring_) were shown to be primarily related to the phases of the seasonally aggregated MJO (spring MJO8 and winter MJO6) (Figure 16(c) and (f)). The autumn mode of precipitation (PC1 _Autumn_) was best explained by the phases of the AAO Autumn...
The changes in the annual and seasonal rainfall regimes in Chile were described in the context of the evolution of major regional and large-scale factors controlling rainfall variability along the extratropical west coast of South America. The PCA method was used to describe and analyse the main significant modes of annual and seasonal precipitation variability in Chile, and to reveal the influence of different climate indices and gridded ocean–atmospheric fields. In general, the approach showed that the explained variance of the first significant modes of annual and seasonal precipitation ranged between 43% during spring (PC1_Spring) and 55% at an annual scale (PC1_Annual). Additionally, the explained variance of second significant mode of precipitation ranged between 13.6% in the fall (PC2_Autumn) and 20% in the summer (PC2_Summer). Regarding annual and spring precipitation patterns, no significant correlations were observed for PC2. Finally, the explained variance of third significant mode of precipitation ranged between 6.3% in winter (PC3_Winter) and 8.7% in austral summer (PC3_Summer); again the annual and spring precipitation patterns were not observed to be significant. The observed spatial (point-based) precipitation patterns were concentrated in Central Chile, especially for the first leading modes of annual and seasonal precipitation. The precipitation patterns in the Northern and Southern regions of Chile were mainly explained by the second and third leading modes (PC2 and PC3) of annual and seasonal precipitation. The use of gridded data at global scale denoted several spatial patterns extending across South America (U-winds, SLP, and SST), which are noticeable at an annual and seasonal scale. In this context, precipitation variability is clearly associated with pressure and temperature changes over the equatorial and austral Pacific Basin that can clearly be observed at an interannual scale (Garreaud et al., 2013), but also at a seasonal scale.

The use of climatic indices revealed that the SOI_Winter explained most of the interannual and winter variability of precipitation (see also Rutllant and Fuenzalida, 1991; Aceituno and Garreaud, 1995; Garreaud and Batisti, 1999; Montecinos and Aceituno, 2003; Garreaud et al., 2009). The immediate effects of ENSO modulation in the variability of precipitation can be explained by the propagation of Rossby waves caused by heat anomalies in the tropical central Pacific that modify the normal atmospheric circulation, generating changes in the storm track that impact Chile (Pizarro et al., 2002). The close relationship between winter and annual accumulation is in agreement with the results of SR1 for central Chile, because in the subtropical domain along the West coast of the continent between 40 and 80% of precipitation falls during the austral winter. This phenomenon is primarily associated with extratropical systems that reach this region when the subtropical anticyclone in the southeast Pacific and the mid-latitude band of migratory low pressure systems are

Figure 16. Scatter plots for the best correlations between different climate indices and the first significant mode of annual and seasonal precipitation (PC1_Annual and PC1_Seasonal). Significant correlations were obtained from t-test with a critical region equal to t(n−2, α = 0.05) = ±2.037 for n = 34 years with two degrees of freedom.

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Table 2. Summary of the results obtained for the leading modes of annual/seasonal precipitation variability in Chile, and their relationships to precipitation anomalies and seasonally aggregated climate indices.

<table>
<thead>
<tr>
<th>Anomalies</th>
<th>Pattern</th>
<th>Significant (leading) modes of annual or seasonal precipitation</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>PC$_1$ Annual explained 55.3% of variance with significant correlations between 20° and 40°S; however, the most substantial spatial pattern of SMP-annual was found to be mainly concentrated in central Chile (30°–40°S).</td>
</tr>
<tr>
<td>Annual</td>
<td>Anomalies</td>
<td>PC$_2$ Annual explained 9.3% of variance however the mode is not significant.</td>
</tr>
<tr>
<td></td>
<td>Climate indices</td>
<td>PC$_3$ Annual explained 7.9% of variance however the mode is not significant.</td>
</tr>
<tr>
<td>Climate indices</td>
<td>Winter SOI (negative phase) and BEST (positive phase) indices corresponding to the same year (JJA$_t$), with a strong and significant spatial influence mainly concentrated in central Chile (30°–35°S).</td>
<td>No significant mode was observed. Therefore, no correlations with climatic indices were calculated.</td>
</tr>
<tr>
<td>DJF (summer)</td>
<td>Anomalies</td>
<td>PC$_2$ Summer explained 20% of variance with similar results to the first mode; however, this time the spatially significant influence was mainly observed between 33° and 40°S.</td>
</tr>
<tr>
<td>Climate indices</td>
<td>Spring MJO (phases 7 and 8) (SON$_t$) was related to the First SMP-DJF, especially significant in regions 35°S northward.</td>
<td>No significant mode was observed. Therefore, no correlations with climatic indices were calculated.</td>
</tr>
<tr>
<td>MAM (autumn)</td>
<td>Anomalies</td>
<td>PC$_3$ Summer explained 8.7% of variance, and it was significantly correlated in northern regions above 25°S.</td>
</tr>
<tr>
<td>Climate indices</td>
<td>Autumn activity (MAM$_t$) of the same year for the Antarctic Oscillation Index (AAO) (negative phase). Significant spatial influence was distributed from 30° to 40°S with the largest values observed between 35° and 40°S.</td>
<td>Spring N3.4 (negative phase) (SON$_{t-1}$) was associated with the third SMP-DJF mainly observed over northern regions between 17° and 25°S. Also some spatial influence was detected around 33°S. PC$<em>3$ Autumn explained 6.5% of variance. It also showed a dipole pattern strongly marked in northern Chile (about 25°S northward and with some significances observed about 45°S and south of 50°S as well. Autumn MJO7 Index (phase 7, positive) of the same season (MAM$</em>{t-1}$) was found to be significantly correlated with the Third SMP-MAM. The spatial influence is specifically concentrated over regions located 25°S northward.</td>
</tr>
</tbody>
</table>
Table 2. continued

<table>
<thead>
<tr>
<th>Pattern</th>
<th>Anomalies</th>
<th>Climate indices</th>
<th>Anomalies</th>
</tr>
</thead>
<tbody>
<tr>
<td>JJA (winter)</td>
<td>Anomalies</td>
<td>Winter SOI (negative phase) and BEST precipitation (phase 6).</td>
<td>Anomalies</td>
</tr>
<tr>
<td>PC1</td>
<td>Winter explained 52.3% of variance in the area between 27° and 40°S.</td>
<td>Winter SOI (negative phase) and BEST precipitation (phase 6).</td>
<td>Anomalies</td>
</tr>
<tr>
<td>PC2</td>
<td>Winter explained 14.5% of variance. A dipole pattern was observed to be positive around 32°S and negative phase of the Autumn AAO (negative phase). The spatial pattern was not entirely defined by the Autumn AAO (negative phase). Only those stations located about 37°S showed significant correlations.</td>
<td>Winter SOI (negative phase) and BEST precipitation (phase 6).</td>
<td>Anomalies</td>
</tr>
<tr>
<td>PC3</td>
<td>Winter explained 6.3% of variance.</td>
<td>Winter SOI (negative phase) and BEST precipitation (phase 6).</td>
<td>Anomalies</td>
</tr>
</tbody>
</table>

A dipole pattern was observed to be significantly correlated to those stations located in regions 25°S northward. No significant mode was observed. Therefore, no correlations against climatic indices were carried out.

PC1 explained 14.4% of variance; however, the mode is not significant.

PC2 explained 16.7% of variance; however, the mode is not significant.

PC3 explained 14.5% of variance. A dipole pattern was observed to be positive around 37°S and negative around 35°S.

PC4 explained 10.6% of variance. No significant mode was observed. Therefore, no correlations against climatic indices were carried out.

PC5 explained 8.2% of variance. No significant mode was observed. Therefore, no correlations against climatic indices were carried out.

PC6 explained 6.3% of variance. It showed to be significantly correlated to those stations located in regions 25°S northward. No significant mode was observed. Therefore, no correlations against climatic indices were carried out.

at their northernmost position (Montecinos and Aceituno, 2003; Garreaud et al., 2009).

The AAO Autumn was found to be the main source of autumn precipitation variability in Central Chile. The AAO plays a significant role in modulating the interannual rainfall variability in all of central Chile, especially during the austral winter semester (April–September). This is because this pattern is partially connected to changes in the meridional SLP gradient at mid- and high latitudes in the southeast Pacific (Quintana and Aceituno, 2012). Villalba et al. (2012) mentioned that negative (positive) precipitation anomalies in northern Patagonia (38°–42°S) are related to positive (negative) phases of the AAO. Moreover, no significant results were observed for the AAO in the southern regions (south of 45°S). This is probably due to the lack of point-based data in the intermediate regions (40° and 50°S), which may be an important source of possible misleading results.

The MJO Spring explained most of the summer precipitation variability (phases 7 and 8), and also most of the spring variability (phase 6). In a composites analysis of precipitation and rainfall intensity in Central-South Chile (30°–45°S) in relation to the real-time multivariate MJO index, Barrett et al. (2012) showed that above normal precipitation and greater frequency of heavy precipitation occurred during phases 8, 1, and 2; the study also showed that below normal precipitation takes place in phases 3–7. In other words, when the deep convective clouds and rainfall are located in the Pacific Basin, the atmospheric circulation can demonstrate an ENSO-like pattern. Therefore, during this phase, precipitation and heavy rainfall events are more likely to occur in the extratropical region, such as the Chilean territory facing the southeastern Pacific Ocean (Carrasco, 2006; Barrett et al., 2012).

5. Conclusions

Because Chile’s regions span from subtropical to extratropical latitudes, there is a strong latitudinal influence in precipitation accumulation which is observed at both annual and seasonal scales. In addition, a narrow distance between the Pacific Ocean and the Andes Mountain range depict an important orographic effect which depends on local orographic conditions and the source of precipitation. In general, Northern regions are affected by rain shadow conditions in the western Andean slopes; however, Central regions can denote important ER in the cross-mountain direction. Southern regions receive large contribution of precipitation from westerlies that generate rain shadow conditions in the eastern Andean slopes.

The SRt showed that most of the contribution of seasonal precipitation to annual accumulation occurs in Central Chile, especially for winter accumulation. This situation was completely different in Northern Chile where most of the annual precipitation is contributed by summer accumulation. We also found significant evidence of temporal changes in the SRt during autumn and winter seasons.
These trends are exclusively concentrated in central and southern regions of Chile.

According to our results, the SPI can satisfactorily explain the intensity of droughts in Chile. However, it is not a valid index in arid regions, such as those regions located above 25°S (Atacama Desert) in Northern Chile, where precipitation often varies from zero to several inches over relatively short periods. The probability analysis demonstrated that, primarily for central Chile, very dry and extremely dry conditions are more likely during La Niña years. On the other hand, very wet and extremely wet conditions were mostly observed during El Niño years.

The annual and seasonal rainfall variability is mostly explained by the first leading mode of precipitation (PC1), especially for Central Chile. The second and third modes of precipitation were found to be significant only for autumn, winter, and spring accumulation. The leading modes of annual/winter precipitation deployed significant relationships with SOI\textsubscript{Winter}, confirming that most of the interannual and winter precipitation variability in Chile can be explained by ENSO phenomena. Significant relationships also were found between seasonally aggregated MJO indices and spring/summer precipitation. This suggests that the seasonal spring/summer precipitation variability in central Chile could be satisfactorily predicted with seasonally averaged MJO indices, especially using phases 6, 7, and 8. Finally, fall precipitation was better explained by the fall AAO, demonstrating that this low-frequency variability mode (in its aggregated version) also accounts for part of the seasonal precipitation in central Chile. The significant spatial patterns of these relationships are in concordance with the analysis of gridded data based on U-winds, SLP, and SST.

The ocean–atmosphere processes that affect precipitation variability in Chile are very complex because there are several factors that influence seasonal and annual accumulation. Despite the detailed annual and seasonal analysis conducted here, more detailed country-based studies are still required to develop a comprehensive understanding of the climatic patterns in a country which is geographically very difficult to study. For example, future studies will be required to define how ocean–atmosphere processes affect the spatio-temporal variability of precipitation accumulation at a monthly or daily scale; and also determine how global warming will disturb the future availability and distribution of water resources in Chile.

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Supporting Information

The following supporting information is available as part of the online article:
Appendix S1. Supplementary Optional Material.

References

Fuenzalida HP, Rutllant J. 1987. Origen del vapor de agua que precipita sobre el Altiplano de Chile. In Proceeding of II Congreso Inter-American de Meteorologia, American Meteorological Society, Buenos Aires, 6.3.1–6.3.4.

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