

Regional climate change scenarios over southern South America for future climate (2080-2099) using the MM5 Model. Mean, interannual variability and uncertainties

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RESUMEN

Este trabajo evalúa las proyecciones de cambio climático del modelo regional MM5 para fines del siglo veintiuno en el escenario de emisión SRES A2 en el sur de Sudamérica. El modelo proyecta: (i) aumento de la precipitación en el centro de Argentina, Uruguay y sur de Brasil en verano y otoño; (ii) disminución de la precipitación en la mayor parte de la región de estudio en invierno y primavera; (iii) marcada reducción de la precipitación en el centro y sur de Chile durante todo el año. En general, el aumento de temperatura proyectado depende de la época y la región examinada; particularmente es máximo en latitudes tropicales y subtropicales en primavera y en altas latitudes en verano. El modelo MM5 proyecta: (i) aumento de la variabilidad interanual de la precipitación en el centro de Argentina y Uruguay independientemente de la época del año; (ii) una leve disminución de la variabilidad interanual de la temperatura en la mayor parte de territorio argentino para verano e invierno; (iii) un leve aumento de la variabilidad interanual de la temperatura en las estaciones intermedias; con valores mayores en el centro de Chile en otoño y en el centro norte de Argentina en primavera. De la evaluación de cuán robustas son las proyecciones climáticas regionales, puede concluirse que la relación señal-ruido es alta para la temperatura y baja para la precipitación. Por lo tanto, el modelo MM5 es una herramienta de mucha utilidad para la generación y evaluación de escenarios regionales de cambio climático en el sur de Sudamérica, en especial para la temperatura. Esto constituye un punto de partida para realizar estudios relacionados con el impacto del cambio climático.

ABSTRACT

This work focuses on evaluating the climate change projected by the end of the 21st century under the SRES A2 emission scenario over southern South America using the regional model MM5. The model projects: (i) an increase of precipitation over central Argentina, Uruguay and southern Brazil during summer and fall; (ii) a decrease in precipitation over most of the study domain during winter and spring; (iii) an important decrease in precipitation over central and southern Chile, through the year. In general, the projected temperature increase depends on the season and the examined area; particularly, it is highest over tropical and subtropical latitudes in spring and over high latitudes in summer. The MM5 model projects: (i) an increase of the inter-annual precipitation variability of precipitation over central Argentina and Uruguay regardless the season; (ii) a slight decrease in interannual temperature variability over large extents of Argentina for summer and winter; (iii) a slight increase in interannual temperature variability at transition seasons; with highest values

over central Chile in autumn and over north central Argentina in spring. From the reliability assessment of regional climate projections, it can be concluded that signal-to-noise ratio is high for temperature and low for precipitation. Therefore, the MM5 model is a useful tool in the generation of regional climate change scenarios of high resolution over southern South America, particularly for temperature, and is a starting point to perform studies related to impacts of climate change.

Keywords: Regional climate modeling, South America, climate change scenarios, uncertainties.

1. Introduction

Current global climate simulations can reproduce relatively well features of general circulation throughout the world (Mizuta *et al.*, 2006; Randall *et al.*, 2007; Vera and Silvestri, 2009; Blázquez and Núñez, 2012a; Cavalcanti and Shimizu, 2012; Rupp *et al.*, 2013; Grose *et al.*, 2014; Gulizia and Camilloni 2015, among others). Nevertheless, their performance deteriorates when looking at finer temporal and spatial scales, which are needed for impact assessment studies on natural environments such as agriculture (Jones and Thornton, 2003); biogeography and ecology (Diniz Filho *et al.*, 2009; Wiens *et al.*, 2009); hidrology and water resource management (Wood *et al.*, 2004; Kotlarski *et al.*, 2005; Buytaert *et al.*, 2009; Kay *et al.*, 2009; Laghari *et al.*, 2012), and oceanography and glaciology (Good *et al.*, 2009; Holland *et al.*, 2010; Vizcaino *et al.*, 2010).

Due to the need for an analysis of climate change at a regional scale, the use of regional climate models (RCMs) for dynamical downscaling were developed to obtain detailed information from general circulation models (GCMs). The most recent report from the IPCC (Flato *et al.*, 2013) shows that the number of regional climate modeling studies throughout the world has increased during the last decade. Moreover, the research community has made great efforts to improve RCMs, for example increasing their resolution, further developing process descriptions, and adding new components. As a consequence, RCMs are regularly tested to evaluate whether they show improvements over global models (Laprise *et al.*, 2008; Rummukainen, 2010).

Dynamical downscaling experiments of climate change scenarios have become available over Europe (Christensen *et al.*, 2001; Giorgi *et al.*, 2004; Gao *et al.*, 2006; Beniston *et al.*, 2007; Christensen *et al.*, 2007a, b; Krichak *et al.*, 2007; Tapiador and Sánchez, 2008; Pieczka *et al.*, 2011; Brankovic *et al.*, 2012; Gaertner *et al.*, 2012), North America (Pan *et al.*, 2001; Liang *et al.*, 2004), Australia (Whetton *et al.*, 2001), Africa (Arnell *et al.*, 2003; Tadross *et al.*, 2005),

and Asia (Rupa Kumar *et al.*, 2006; Im *et al.*, 2007; Islam *et al.*, 2008; Islam and Rehman, 2009; Islam *et al.*, 2009; Koo *et al.*, 2009; Chotamonsak *et al.*, 2010; Saeed *et al.*, 2013). Particularly over South America, several publications can be mentioned: Marengo and Ambrizzi (2006), Solman *et al.* (2007), Garreau and Falvey (2008), Núñez *et al.* (2008), Marengo *et al.* (2009, 2010), Soares and Marengo (2009), Urrutia and Viulle (2009), Sörensson *et al.* (2010), and Cabré *et al.* (2010, 2014) for SRES A2 and SRES B2; Marengo *et al.* (2012), Franchito *et al.* (2014), and Reboita *et al.* (2014) for A1B; and Teichmann *et al.* (2013), Chou *et al.* (2014) for the new set of scenarios termed as representative concentration pathways (RCPs).

However, due to the complexity of the climate system and modeling processes, relationships between uncertainty types must be considered. Uncertainty in future climate change derives from three main sources: (i) forcing, (ii) model response, and (iii) internal variability (Hawkins and Sutton, 2009). The first arises from incomplete knowledge of external factors influencing the climate system, including future trajectories of anthropogenic emissions of greenhouse gases (GHGs), stratospheric ozone concentrations, land use change, etc. The second occurs because different models may yield different responses to the same external forcing as a result of differences in, for example, physical and numerical formulations. The third is the natural variability of the climate system that occurs in absence of external forcing, also termed “climate noise”, and includes processes intrinsic to the atmosphere, the ocean, and the coupled ocean-atmosphere system (Deser *et al.*, 2012). Consequently, uncertainties studies of global (Giorgi and Mearns, 2002; Hawkins and Sutton, 2009; Hawkins and Sutton, 2011; Räisänen, 2006; Blázquez *et al.*, 2012; Blázquez and Núñez, 2012b; Deser *et al.*, 2012; Torres and Marengo, 2013) and regional climate change projections (Mitchell and Hulme, 1999; Giorgi and Francisco, 2000; Monier *et al.*, 2013; Solman *et al.*, 2013) can be found all over the world.

Therefore, uncertainties associated with RCMs outputs must be taken into account, as they can be quantified and minimized to provide valuable information for decision-making (Foley, 2010). In order to provide a full quantification of uncertainties in the projected regional climate and useful information for impact studies, the broad range of uncertainties should be considered. For the purpose of reducing uncertainties associated with RCMs, several methodologies have been implemented: (i) multimodel ensembles (Doblas-Reyes *et al.*, 2005), which attribute the improvement to the use of different models and increased ensemble size and are useful to characterize intermodel variability; and (ii) perturbed physics ensembles, since an ensemble may also consist of different runs of the same model, each with perturbed versions of the original model physics. This approach is highly useful for quantifying variability within the model. However, the optimal approach would be to use a multimodel perturbed-physics ensemble. For instance, the generation of ensembles performed with different RCMs simulations nested into different GCMs under different emission scenarios is an approach that has been followed by several international efforts, such as the projects PRUDENCE (Déqué *et al.*, 2005; Gao *et al.*, 2006; Christensen *et al.*, 2007a, b), ENSEMBLES (Hewitt, 2005), STARDEX (Haylock *et al.*, 2006) and CORDEX (Jacob *et al.*, 2013) for Europe; CORDEX (Giorgi *et al.*, 2009) for Africa; and CLARIS LPB (Solman *et al.*, 2013) and CORDEX (Solman, 2013) for South America.

This paper presents an analysis of climate change over southern South America based on the MM5 regional model simulation. The regional model was driven by global atmospheric model HadAM3H under the SRES A2 scenario. It is focused on seasonal mean precipitation and near-surface temperature, because these variables are useful for impact studies. Therefore, based on the need of detailed information on climate change projections, it performs an assessment of climate change projections and associated uncertainties over southern South America. This article is organized as follows: section 2 gives a brief summary of some characteristics of the regional model configuration and an update on the new set scenarios from the AR5 (IPCC, 2013). Results are found in section 3; firstly a brief evaluation of uncertainties in future regional climate projections

is found in section 3.1, where the reliability of predictions is explored measured by the signal-to-noise ratio; the spatial distribution of changes of surface variables in mean climate is found in section 3.2; the statistical significance of seasonal changes was determined using Student's *t* test at a confidence level of 95%, and analyzed jointly in this section; furthermore, section 3.3 presents the analysis of changes in the annual cycle, and section 3.4 presents the analysis of spatial distribution of changes in interannual temperature and precipitation variability. Finally, the most important conclusions are summarized in section 4.

2. Model description, experiment design and scenarios

2.1 The regional model MM5

The regional climate simulation was performed using the fifth-generation nonhydrostatic mesoscale model MM5 version 3.6 (the latest version available at the time of starting the experiment design) developed by Penn State University (PSU) and the National Centers for Environmental Prediction/National Centre for Atmospheric Research (NCEP/NCAR) (Grell *et al.*, 1993). Details of the regional model configuration used to perform the continuous 20-year simulation are found in Cabré *et al.* (2014). It includes the Grell convective scheme (Grell *et al.*, 1993), and the planetary boundary layer parameterization is formulated following the MRF scheme by Hong and Pan (1996). Surface processes are represented by the Noah land surface model (Chen and Dudhia, 2001). Moisture tendencies were calculated with an explicit moisture scheme (Hsie *et al.*, 1984). The calculation of radiative heating or cooling in the atmosphere accounts for longwave and shortwave interactions with explicit clouds and clear air. The radiation package calculates long-wave radiation through clouds and water vapor, based on Stephens (1978, 1984) and Garrand (1983); it also accounts for short wave absorption and scattering in clear air, and reflection and absorption in cloud layers Stephens (1984). The nonhydrostatic dynamics allow it to be used effectively to represent phenomena with very few kilometers of resolution. The surface model Noah LSM has been implemented to improve the representation of surface processes. Moreover, the MM5 model requires initial and time evolving boundary conditions for wind components, temperature, geopotential height, relative humidity

and surface pressure. These variables were provided in a six-hourly interval within a relaxation zone in the lateral boundaries.

The regional model was run in a Mercator grid with approximately 40 km resolution in both horizontal directions, with 158 points in the west-east direction and 150 points in the south-north direction, and with 23 sigma levels in the vertical direction. The land sea mask and topography have been derived from the US Navy 10 min resolution dataset. Vegetation and soil properties were obtained from USGS vegetation/land use database.

The integration domain covers southern South America, from 12 to 58° S, 28 to 92° W. The model topography and the study domain are displayed in Figure 1. It can be seen that the selected domain covers the whole of Argentina and is expected to capture the regional climate resulting from the interactions of large-scale and local circulation.

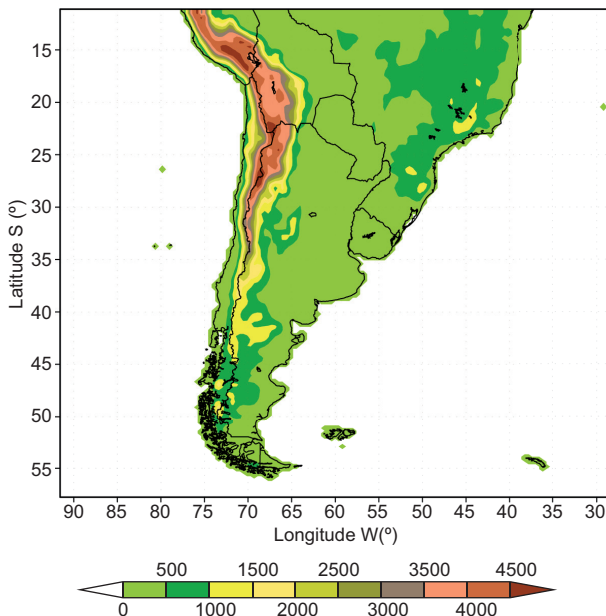


Fig. 1. Model domain and topography of the MM5 regional climate model. Contours are drawn every 500 m.

2.2 Boundary conditions and experiment design

The MM5 model is forced by global atmospheric model HadAM3H. Regarding future simulation (2080-2099), the global model has been forced by sea surface temperature (SST) calculated using the observed SST and modified by the sum of mean change and the trend obtained from the coupled global model

projection. The integration was performed with the coupled HadCM3 model (Gordon *et al.*, 2000).

The experimental design of the regional simulation for present climate (1970-1989) started on January 1, 1969 up to December 1, 1989. A one-year spin-up time was adopted (Christensen, 1999) for surface variables (such as soil moisture and soil temperature for different sub levels between 1 and 2 m) from the land surface model. Likewise, the regional climate simulation for future climate started on January 1, 2079 up to December 1, 2099. As for the present climate simulation, a one-year spin-up time was adopted (Christensen, 1999).

The HadAM3H model has a 1.25° latitude by 1.875° longitude resolution; details of its characteristics can be found in Pope *et al.* (2000). The HadAM3H climate simulations from 1961 to 1990 were initialized with atmospheric and land surface conditions from the Coupled Ocean-Atmosphere Global Climate Model (HadCM3) from the Hadley Centre and forced with observed SST and sea-ice distribution from the HadISST dataset (Rayner *et al.*, 2003). This database shows a mixture of global monthly SST and sea ice concentration at 1° horizontal resolution from 1871.

2.3 Scenarios vs. representative concentration pathways

In the AR4 (IPCC, 2007), emission scenarios were built based on the storylines grouped into a more economically concerned development or a more environmental and sustainable development; also into a more globalized world or a more regionally developing world. In the AR5 (IPCC, 2013), scenarios are based on total anthropogenic radiative forcings at the end of the 21st century. Economic models can take different paths to reach four different radiative forcings that are equivalent to different concentration paths of greenhouse gases (GHGs), the so-called RCPs. The four different scenarios are labeled as: RCP 8.5, RCP 6.0, RCP 4.5, and RCP 2.6, which correspond to radiative forcings of 8.5, 6.0, 4.5, and 2.6 W m^{-2} , respectively. The first RCP is the most pessimistic scenario, resulting in a global average warming at the end of the 21st century of about 4 °C, whereas the last RCP is the most optimistic and corresponds to a global warming of about 1 °C. Radiative forcing in RCP 8.5 corresponds approximately to A2 scenario in AR4 (IPCC, 2007). Particularly,

SRES A2 has a similar trajectory to that of RCP8.5, with both reaching about 8 W m^{-2} by 2100. SRES A2 is also similar to RCP 8.5 in terms of changes in global mean temperature. For instance, for RCP 8.5, global mean surface temperatures for 2081-2100, compared to those for 1986-2005, are likely to be in the range of 2.6 to $4.8 \text{ }^{\circ}\text{C}$, with a multi-model mean increase of $3.7 \text{ }^{\circ}\text{C}$. The projected warming for SRES A2 for the 2090-2099 period, relative to 1980-1999, is given by IPCC AR4 as 2.0 to $5.9 \text{ }^{\circ}\text{C}$, with a best estimate of $3.4 \text{ }^{\circ}\text{C}$ (IPCC, 2013).

3. Results

3.1. Analyses of uncertainties, changes of mean climate vs. interannual variability

This section assesses the uncertainties associated with climate projections of seasonal temperature and precipitation over southern South America by the end of 21st century. The objective is to identify how robust is the projected change compared to interannual variability, which is a measure of the inherent natural variability of the system or noise. Therefore, a key question is how reliable climate models are or how much we can trust on these projections. Identifying which areas and for what variables uncertainties are larger means that the regional simulation may be less reliable. In order to interpret climate statistics correctly, the definitions of climate change, signal-to-noise ratio and statistical significance have been studied for several authors (Hayashi, 1982, and references therein). Accordingly, the significance of projected changes is evaluated by the signal-to-noise ratio defined by the following equation:

$$R_{\text{var}} = \frac{|\Delta \overline{VAR}_{\text{future} - \text{present}}|}{STDV_{\text{var}}} \quad (1)$$

where *var* represents either surface air temperature or precipitation; $\Delta \overline{VAR}_{\text{future} - \text{present}}$ is the signal of climate change (defined as the difference between the climate in the scenario simulation and present climate simulation) and $STDV_{\text{var}}$ is the interannual variability, as a measure of noise (computed as the standard deviation of projections). R_{var} greater than 1 indicates that the signal of climate change is greater than the interannual variability, consequently the signal of climate change is robust, which relates with areas of low uncertainty and likewise to more confident

projections; whereas R_{var} lower than 1 indicates that the signal of climate change is lower than the interannual variability (noise dominates), consequently the signal of climate change is not robust, which relates with areas of high uncertainty.

Figure 2 displays the spatial distribution of signal-to-noise ratio for mean precipitation changes (R_{precip} , left column) and for mean temperature changes (R_{temp} , right column) for all seasons.

The analysis of reliability is summarized for four regions: (1) subtropical latitudes, (2) Uruguay-southern Brazil, (3) east-central Argentina, and (4) central and southern Chile. For precipitation, the magnitude of climate change is lower than the amplitude of interannual variability throughout the year over the entire domain. This suggests that precipitation projections are less reliable all year long over southern South America. However, few regions were identified where the signal was greater than the noise: south-central Argentina in summer and north-western Argentina and Peru in winter. On the other hand, the signal of climate change in temperature seems to show a general opposite behavior. It is greater than interannual variability all over the domain in all seasons, coinciding with areas of more confident projections. Particularly, temperature projections are robust over east-central Argentina and Uruguay-southern Brazil, with highest values of R_{temp} over subtropical latitudes during summer. In winter, temperature projections are reliable over north-western Argentina, northern Chile, western Bolivia and subtropical latitudes. On the other hand, temperature projections are less reliable over east-central Argentina and Uruguay-southern Brazil during winter. The autumn pattern of R_{temp} is similar to that of spring. Particularly, reliability of temperature projections is high over northern Argentina, Uruguay-southern Brazil, and subtropical latitudes, and it is low over the rest of the study domain.

Consequently, the reliability of temperature projections is high, with signal-to-noise ratio greater than 1 over most of the study domain. On the other hand, the reliability of precipitation projections was low. However, the signal-to-noise ratio was close to 1, and even greater than 1 over southeastern Argentina in summer. Figure 3 summarizes the important findings of climate projections reliability over southern South America by the end of the 21st century.

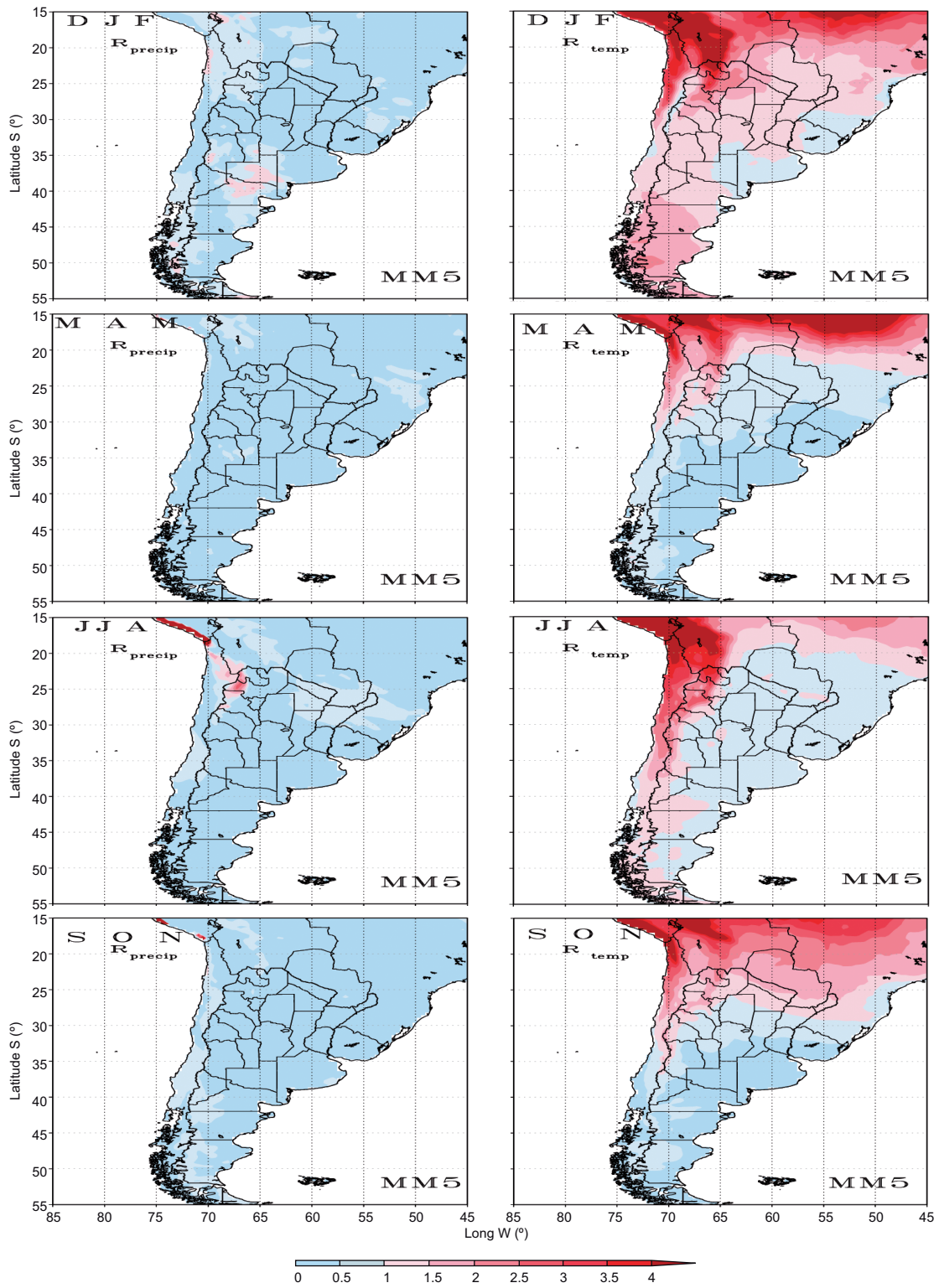


Fig. 2. Signal-to-noise ratio for precipitation (R_{precip} , left column) and temperature (R_{temp} , right column) by the end of the 21st century during summer (DJF), autumn (MAM), winter (JJA) and spring (SON). Dimensionless coefficient.

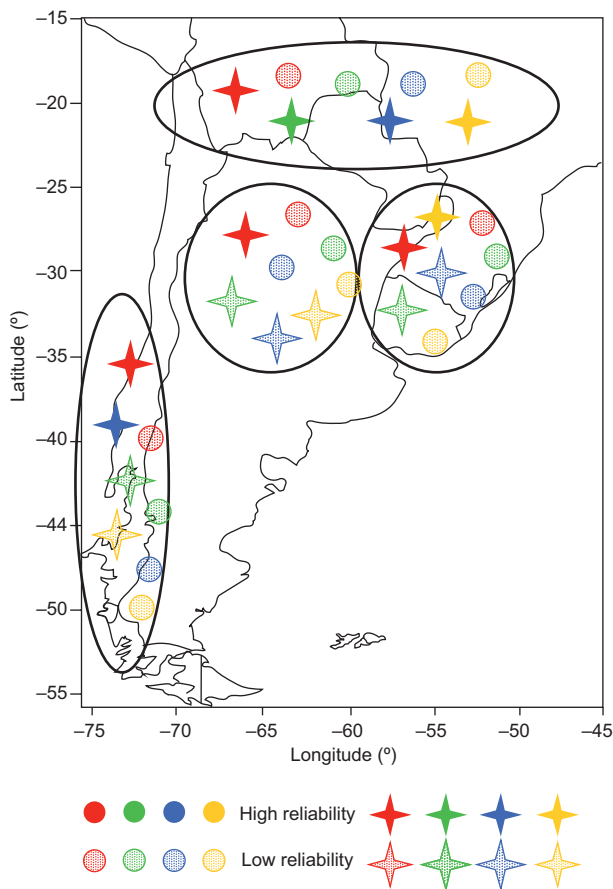


Fig. 3. Schematic graph of the MM5 regional model reliability for climate change on precipitation (circles) and temperature (stars). Red: DJF; green: MAM ; blue: JJA; orange: SON.

3.2 Changes in mean climate

The discussion focuses on the analyses of changes in mean climate over southern South America using the MM5 regional model by the end of the 21st century (2080-2099), relative to 1970-1989 for summer (DJF), autumn (MAM), winter (JJA) and spring (SON).

3.2.1 Spatial distribution

3.2.1.1 Changes in precipitation

Left column in Figure 4 shows the spatial distribution of projected changes in seasonal precipitation over southern South America. Projected precipitation changes are positive or negative depending on the season and the region analyzed. The statistical significance of changes is addressed by means of a Student's *t*-test at a 95% confidence level and is also drawn in the figure with blue and red contours for positive and negative changes, respectively.

During summer, the figure shows a decrease in precipitation over southern Chile, Patagonia and the Bolivian Plateau. Furthermore, a strong precipitation increase is projected over the centre of Argentina and southern Brazil, Paraguay and western Bolivia. This precipitation increase is mainly due to the increase of the cyclonic circulation known as Chaco Low, which together with the southern displacement of the Atlantic Subtropical Anticyclone (Camilloni, 2005; Di Luca *et al.*, 2006) results in a stronger moisture transport towards central Argentina. This pattern of change also agrees with results from the multi-model ensemble mean shown in Meehl *et al.* (2007) for the SRES A1B scenario. In addition, Soares and Marengo (2009), using the HadRM3P RCM have found an intensification of the Low Level Jet (LLJ), which would also explain the increased transport of moisture from the Amazonia into northern and central Argentina. Related to the study of statistical significance, it is worth noting that negative precipitation changes are statistically significant in southern Chile and the Bolivian Plateau, whereas positive precipitation changes are statistically significant in central Argentina, Uruguay-southern Brazil and subtropical latitudes. Those results are in agreement with Núñez *et al.* (2008).

Projected precipitation changes for MAM are similar to those for summer, although with somewhat smaller magnitude. Particularly, a precipitation decrease is apparent over central and southern Chile, the Bolivian Plateau, Argentinean Patagonia, and northern Argentina; whereas a rainfall increase is projected over central Argentina, Uruguay, southern Brazil and Paraguay. In general, these changes are not statistically significant at a 95% confidence level. Only a few small regions were identified as statistically significant, which include positive changes in southern Brazil and negative changes in subtropical latitudes, central Chile and the Bolivian Plateau.

During winter, the projected changes are generally negative (-0.5 mm day^{-1}) over the entire domain, with a maximum precipitation decrease over central Chile. These changes are statistically significant at a 95% confidence level and are in agreement with Núñez *et al.* (2008). On the other hand, a slight precipitation increase is projected over southern of Chile as well as over central Argentina, southern Uruguay and mountainous regions between Chile, Bolivia and Argentina. Changes in the precipitation

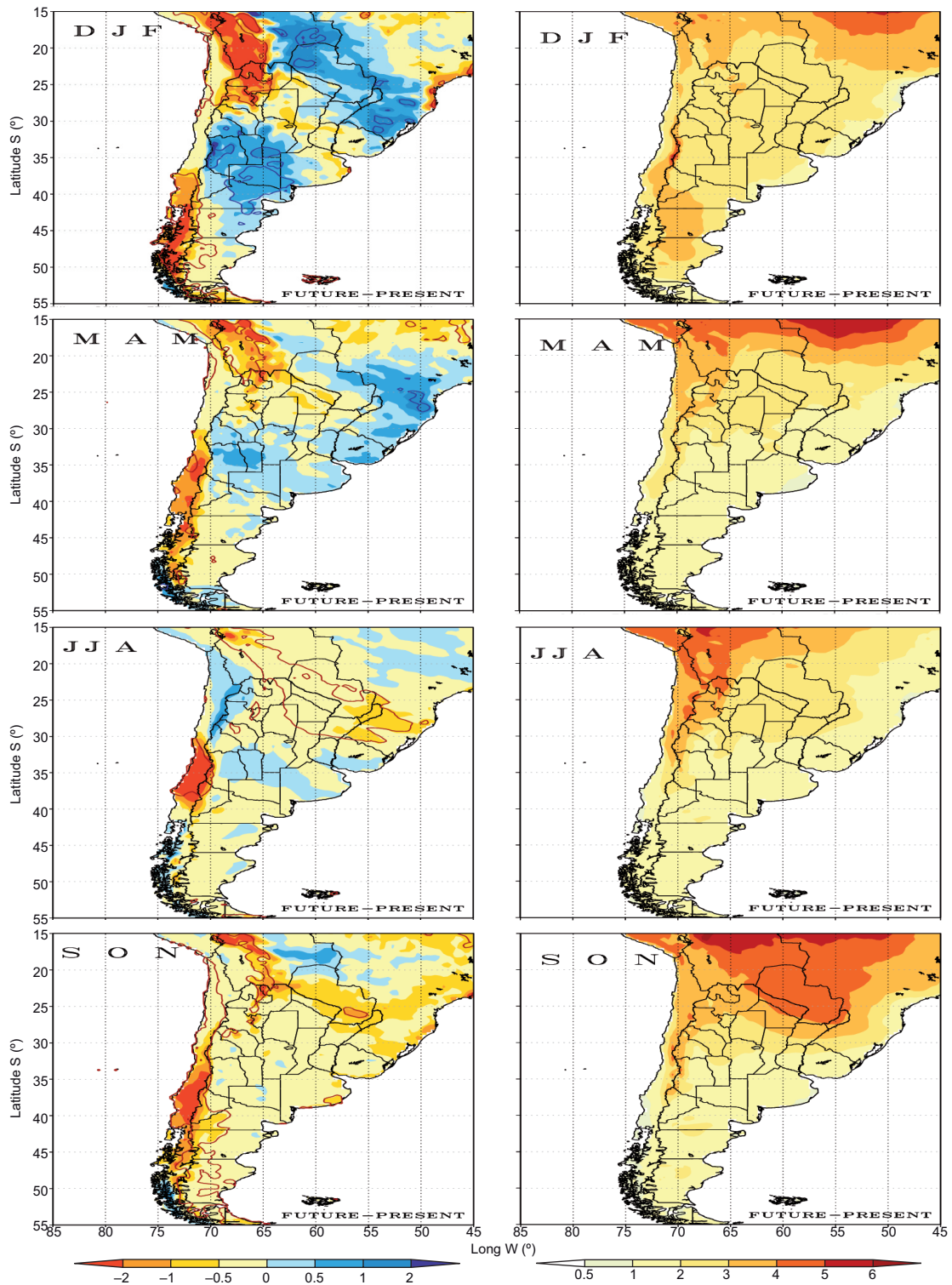


Fig. 4. Spatial distribution of simulated change in precipitation (mm day⁻¹, left column) and temperature (°C, right column) for summer (DJF), autumn (MAM), winter (JJA) and spring (SON), for the period 2080-2099 minus the period 1970-1989. Contoured regions indicate a 95% confidence level.

amount over Chile are consistent with the poleward expansion of the South Pacific subtropical high and the associated poleward shift of the Pacific Ocean storm-track, inducing a significant decrease in precipitation over central Chile and a precipitation increase over southern Chile and southern Argentina (Núñez *et al.*, 2008). In subtropical latitudes, east of the Andes, the figure shows a general pattern of rainfall decrease with maximum precipitation decreased over eastern Argentina, southeastern Paraguay and southern Brazil. The pattern of precipitation change from the regional model agrees with the multi-model ensemble mean shown in Meehl *et al.* (2007) for the SRES A1B scenario.

In general, the model projects a rainfall decrease throughout the domain for spring. Particularly, a maximum rainfall decrease is found over central and southern Chile (around -2 mm day^{-1}) and western Bolivia; whereas a slight rainfall increase is found over eastern Bolivia and northern Paraguay. However, only the precipitation decrease over central and southern Chile and western Bolivia is statistically significant.

Regardless of the season, a rainfall increase is projected over the eastern slope of the Andes (Argentina) and a rainfall decrease over the western slope of the Andes (Chile), particularly in mid-latitudes during summer and fall. In general, a precipitation increase is projected over central Argentina, Uruguay and southern Brazil, especially during summer and autumn; while a precipitation decrease is projected over Paraguay and south of Brazil during spring and over central Chile during winter and spring. Overall, projected precipitation changes seem to occur more likely in summer than in winter. In general, projected changes for summer and winter are in agreement with other studies from different RCMs and scenarios (Núñez *et al.*, 2008; Cabré *et al.*, 2010; Marengo *et al.*, 2010; Krüger *et al.*, 2012; Reboita *et al.*, 2014; Sánchez *et al.*, 2015). Besides, the simulated pattern of precipitation change is also similar to seasonal precipitation change from global climate models under SRES A2 (Torres and Marengo, 2013). Moreover, Chou *et al.* (2014) have found similar patterns of precipitation changes under the RCP 8.5 scenario based on the Eta RCM forced by two GCMs, the HadGEM2-ES and the MIROC5. For example, during summer a rainfall increase is projected over central Argentina, while a rainfall decrease is estimated over southern

Brazil. In winter, the pattern of precipitation change under SRES A2 is similar to the RCP 8.5 at the end of the century. In general, the pattern of change is well represented and differs only slightly in magnitude. In addition, the Argentinean Third National Communication on Climate Change (SAYDS, 2014) to the United Nations Framework Convention on Climate Change (UNFCCC) has been assessed using climatic scenarios RCP 4.5 and RCP 8.5 from the CMIP5 (Taylor *et al.*, 2012) and climatic scenarios SRES A1B from the CMIP3 (Meehl *et al.*, 2007a). Especially, according to the ensemble mean of 42 models from the CMIP5, the mean precipitation change relative to 1981-2005 is a decrease of approximately 10 to 20% over northern and central Patagonia and a similar increase over central Argentina and most of the country, by 2075-2079 under RCP 8.5. Similar results were obtained in this study under the SRES A2 scenario. Furthermore, the statistical significance of results is in agreement with other studies from regional and global models (Núñez *et al.*, 2008; Blázquez *et al.*, 2012) under SRES A2 and SRES A1B.

3.2.1.2 Changes in mean temperature

Spatial distribution of the projected seasonal changes for surface air temperature by the end of the 21st century (2080-2099) is displayed in Figure 4 (right column). The increase in surface air temperature is of the order of 1.5 and 5.5 °C over southern South America by the end of the 21st century. In addition, warming is around 2.5 to 5.5 °C over tropical and subtropical latitudes, and does not exceed 3 °C south of 40° S. Particularly, the maximum temperature increase (between 4 and 5 °C) occurs generally in spring over tropical and subtropical regions and over Argentinean Patagonia during summer (3 °C).

During austral summer the model projects a temperature increase of around 3.5 °C over subtropical areas, 3° C over Argentinean Patagonia, and a maximum warming (4.5 °C) over subtropical latitudes. In winter, the regional model MM5 projects a temperature increase of around 2.5 °C over Paraguay and northern Argentina (similar magnitudes were obtained for summer and fall) and greater than 4 °C over the Andean region between 15 and 30° S. The projected maximum warming over the Andean region is also simulated by PRECIS RCM (CONAMA, 2006) under SRES A2 and by RegCM3 RCM (Reboita *et*

al., 2014), MRI/JMA GCM (Blázquez *et al.*, 2012) and by an ensemble of RCMs (Sánchez *et al.*, 2015) under SRES A1B. Even though the projected temperature increase over southern South America depends on the season and the examined area, it should be noted that the largest warming is generally found over tropical and subtropical latitudes in spring and over the Andes during winter and spring.

For summer and winter, similar patterns of warming were found by Núñez *et al.* (2008) using the MM5 RCM, Krüger *et al.* (2012) using the RegCM3 RCM, Marengo *et al.* (2010) using the RegCM3, Eta CCS and HadRM3P models, Marengo *et al.* (2011) using the Eta/CPTEC model, Reboita *et al.* (2014) using RegCM3, and Sánchez *et al.* (2015) using an ensemble of RCMs. Note a reduction of 2 °C in the rise of temperature over southern Brazil, Paraguay, Bolivia and northern Argentina during JJA, compared to Núñez *et al.* (2008) under the same scenario. Moreover, warming has decreased about 1 °C during DJF. Therefore, this simulation of climate change has greatly improved the spatial distribution of temperature change, reducing the magnitude of projected warming.

According to Torres and Marengo (2013), all of the CMIP3 models project a temperature increase in all of South America that is more intense in its tropical portion and by the end of the century in higher emission scenarios (SRES A1B and SRES A2). For the South American continent as a whole, the range of temperature change is between 2 and 5 °C for the SRES A2 emission scenario. Similar results can be found in Meehl *et al.* (2007b). Moreover, Chou *et al.* (2014) have found similar temperature changes under the RCP 8.5 scenario based on the Eta RCM forced by two GCMs, the HadGEM2-ES and the MIROC5. For example, in summer the projected warming is around 2.5 to 5 °C over central Argentina and around 3 to 3.5 °C over Patagonia. Whereas in winter the pattern of temperature increase is similar to SRES A2, the magnitude of temperature changes under RCP 8.5 is slightly lower than the previous one. In addition, the results from the Argentinean Third National Communication on Climate Change (SAYDS, 2014) to the UNFCCC project a temperature increase for the rest of the century over Argentina, and as it was expected the RCP 8.5 predicts the highest warming. Especially, according to the ensemble mean of 42 CMIP5 models, the mean

temperature change relative to 1981-2005 could reach around 2.5 to 5 °C over Argentina by the end of the century. Regarding the statistical significance of mean temperature changes, a conventional *t*-test at a 95% confidence level was also performed (not shown in the right column of Figure 4). Therefore, the stated changes are statistically significant for all seasons and over the entire domain. Results are in agreement with other studies (Núñez *et al.*, 2008; Blázquez *et al.*, 2012, among others) for SRES A2 and SRES A1B.

3.3 Projected annual cycle

3.3.1 Precipitation

Figure 5 shows the sub-regions selected for evaluating the annual cycle of precipitation. These regions were chosen particularly due to differences in their precipitation regimes, and most of them were defined by Solman *et al.* (2007). Figure 6 shows the annual cycle of monthly precipitation for present (1970-1989) and future climate (2080-2099) under the SRES A2 scenario (left column) and projected precipitation changes (right column) from the MM5 regional model for the sub-regions defined in Figure 5.

There are large seasonal and geographical variations in the changes of the annual cycle of precipitation, as can be seen in Figure 6. An increase in monthly precipitation is projected from April to November (highest in November [around 30%]), except in October; while the opposite behavior occurs from December to March (highest in March [around 40%]) over the southern Andes (SA). For Subtropical Andes (SUA), a rainfall decrease is projected from March to October (around 50%), except in June. The greatest rainfall decrease is projected during the rainy season. Over the Paraguay (PA) region, wetter conditions are projected throughout the year, except for August, September and October. Over Cuyo (CU), the mean precipitation increases (between 80 and 100%) occur during the first half of the year, and this behavior is reversed during the second half of the year (with a maximum deficit in precipitation of around 60% in September).

A maximum decrease in precipitation is projected during September (70%) and a significant increase in June (150%) over the southeast Pampas (SEP). Over the La Plata basin (LPB) region a subtle reduction of precipitation is projected during January, March, April, August, September, November and December,

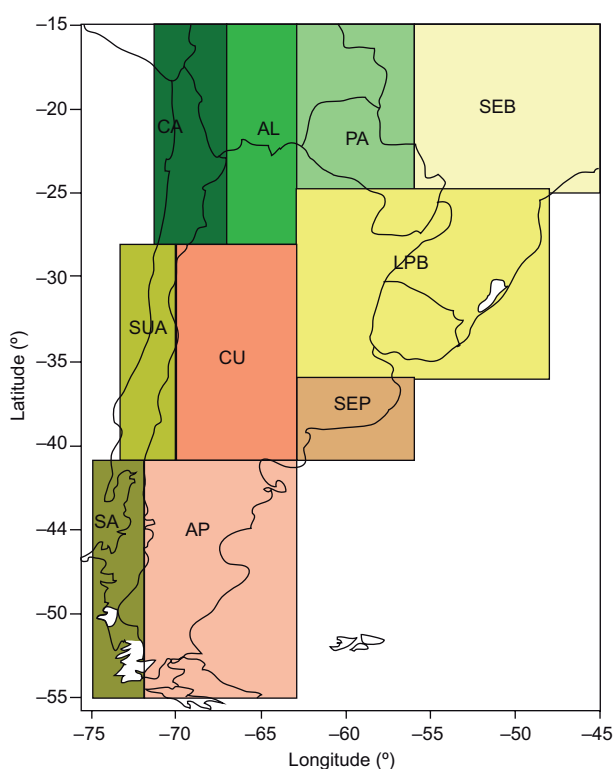


Fig. 5. Sub-regions used for a more detailed analysis of the annual cycle of precipitation: Andes centrales (CA), Altiplano (AL), Paraguay (PA), southeastern Brazil (SEB), subtropical Andes (SUA), Cuyo (CU), La Plata basin (LPB), south-eastern Pampas (SEP), southern Andes (SA), and Argentinean Patagonia (AP).

with increases during February, May, June, July and October. The greatest increase/decrease is found in June/September, around 40 and 60%, respectively. Similar results were obtained by Marengo *et al.* (2009). Over the southeastern Brazil (SEB) region, wetter conditions are projected from March to July and dryer conditions from August to February. Projected changes for the annual cycle of precipitation by the end of the 21st century predict a general rainfall decrease throughout the year, except on winter over the SEB and LPB regions. The maximum deficit is estimated during spring, while a smaller deficit is estimated during fall. Moreover, regions located to the east of the study domain (SEB, LPB, SEP) show an increase in the amplitude of the annual cycle of precipitation during winter and a decrease from spring to autumn.

Over the central Andes (CA) and Altiplano (AL), the regional MM5 model projects a monthly precipitation decrease (on average of 50%) throughout the

year, except for winter. For example, a maximum increase of 70% is projected in June over the CA region. Similar results were obtained by Urrutia and Vuille (2009) under SRES A2 and B2 from PRECIS RCM, and recently by Chou *et al.* (2014) under RCP 8.5, based on the Eta RCM. Broadly speaking, the results obtained are in agreement with previous studies (Nuñez *et al.*, 2008; Marengo *et al.*, 2009; Krüger *et al.*, 2012, among others).

Figure 7 displays a schematic graph of seasonal projected changes in precipitation for the subregions defined in Figure 5.

3.3.2 Mean temperature

Figure 8 shows the selected subregions for evaluating changes in the annual cycle of temperature. These regions were defined in Solman *et al.* (2007). The choice of these subregions was motivated by the analysis of the projected climate change documented in Núñez *et al.* (2008). Figure 9 displays the annual cycle of monthly temperature for present climate (1970-1989) and future climate (2080-2099) under the SRES A2 scenario (left column), and projected temperature changes (right column) from the MM5 RCM for subregions defined in Figure 8.

The projected warming is present throughout the year, however, the maximum increase in temperature is found during autumn and spring in all the analyzed regions, except over the Andes (AN) and Patagonia (PAT).

Over southeastern South America (SESA), temperature increase is lower than 1.5 °C during the first half of the year, while it is between 1.5 and 3 °C during the second half of the year. In particular, Marengo *et al.* (2009) show a pronounced temperature increase during spring and summer (4 °C) with a strong difference in amplitude between winter and summer in regions located over subtropical latitudes. Significant warming over the SESA region during spring under the SRES A2 scenario is in agreement with the results of Núñez *et al.* (2008). Therefore, the projected warming simulated here is of lower magnitude. From this it can be inferred that projections of the present regional climate change show greater consistency with other regional model temperature projections over the SESA region. The magnitude of the projected warming (4 °C) throughout the year over the SESA and Subtropical (ST) regions is also documented by other authors (Marengo *et al.*, 2010;

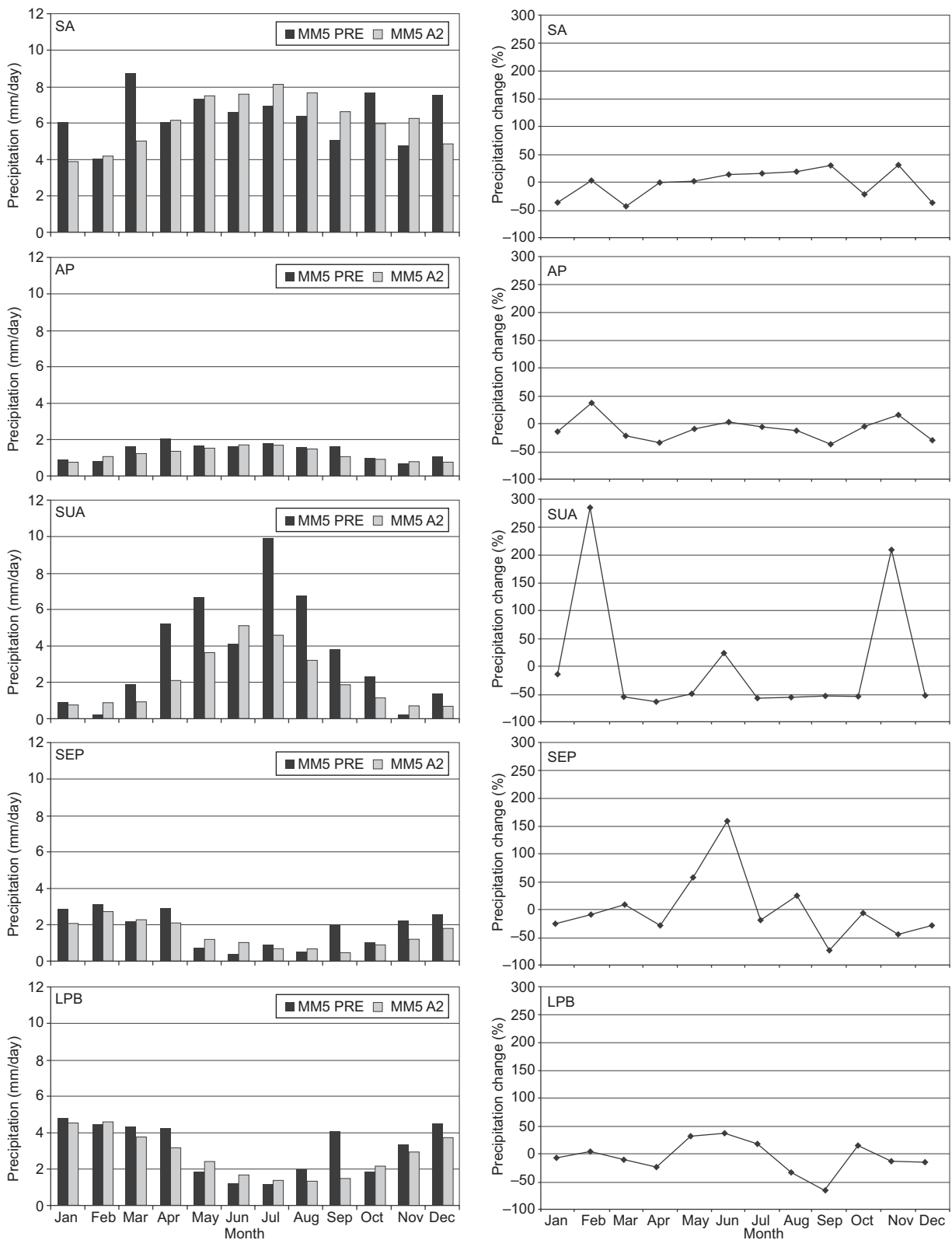


Fig. 6. Left column: Annual cycle of the monthly mean precipitation for present (mm day⁻¹, dark bars) and future climate (light grey bars). Right column: Projected precipitation changes (%). Subregions were defined in Figure 5. (*Continues.*)

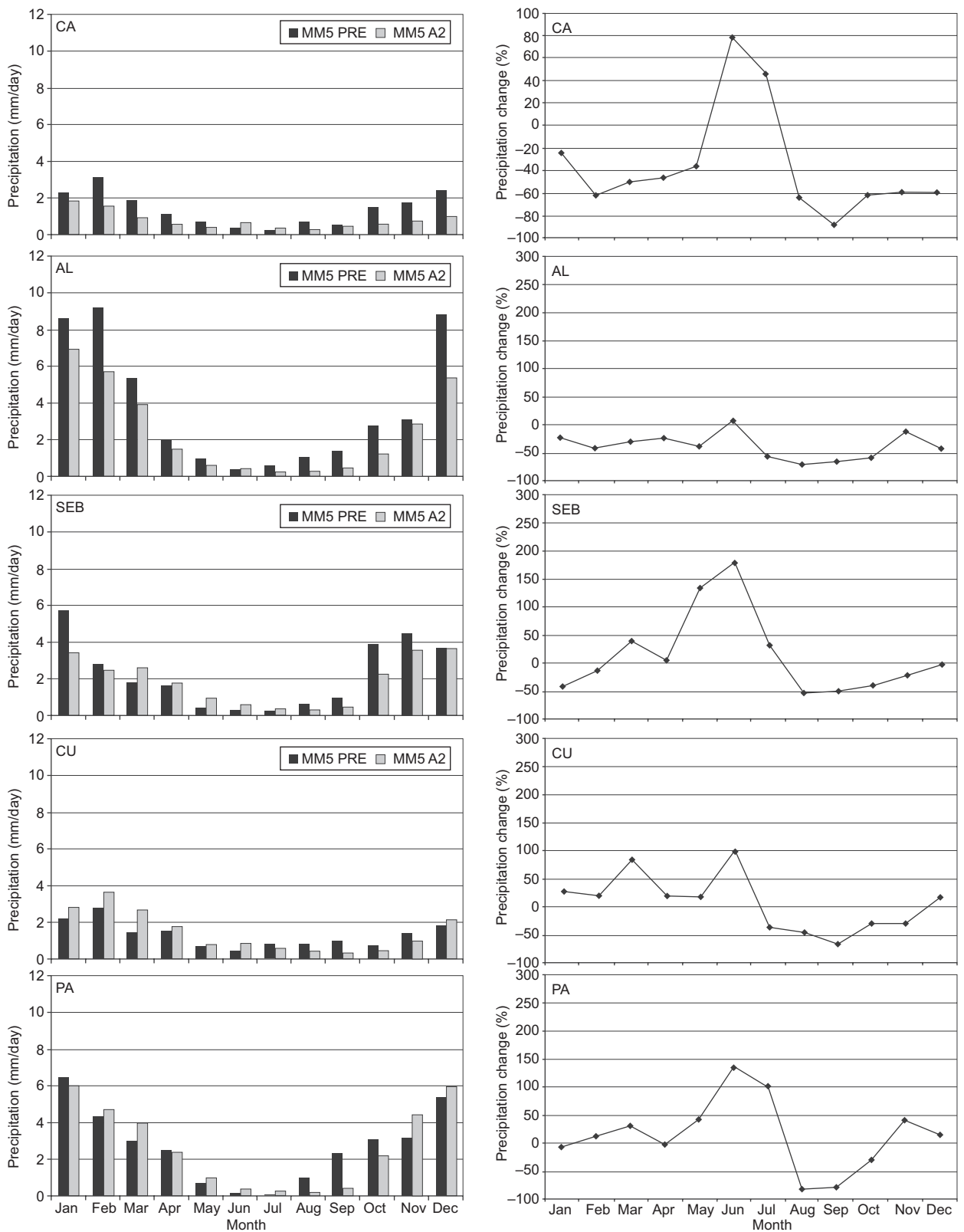


Fig. 6. (Continued.)

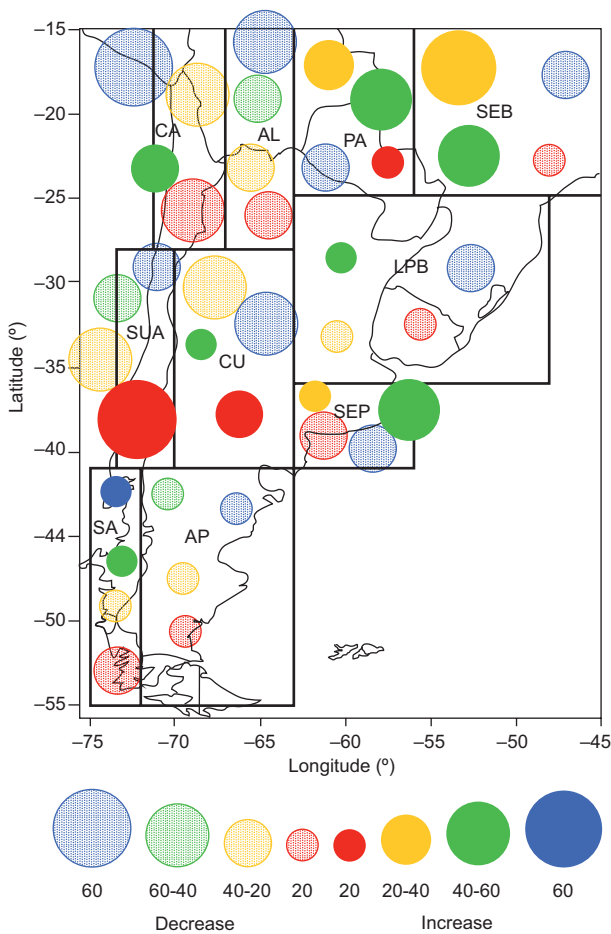


Fig. 7. Schematic graph of the MM5 seasonal projected changes (%) in the subregions defined in Figure 6 (red: DJF; orange: MAM; green: JJA; blue: SON). The size of the circles indicates the magnitude of the overestimation (solid circles) and underestimation (dot circles)

Krüger *et al.*, 2012). Over the central Argentina (CARG) region, the MM5 RCM projects a maximum increase in temperature (2-3 °C) for November and December. Over the ST region, maximum/minimum warming is found in September/June.

On the other hand, the AN and PAT regions show a different behavior compared to previous regions. Over PAT, the projected (maximum/minimum) temperature increase is found in February/May-June. The maximum warming (about 2.5 °C) in AN occurs during July and the minimum (about 2 °C) in February. Similar results over AN are shown in Núñez *et al.* (2008), however the warming obtained here is of lower magnitude. Torres and Marengo (2013) have found similar results over the AN region under the SRES A2 emission scenario with 24 GCMs from the

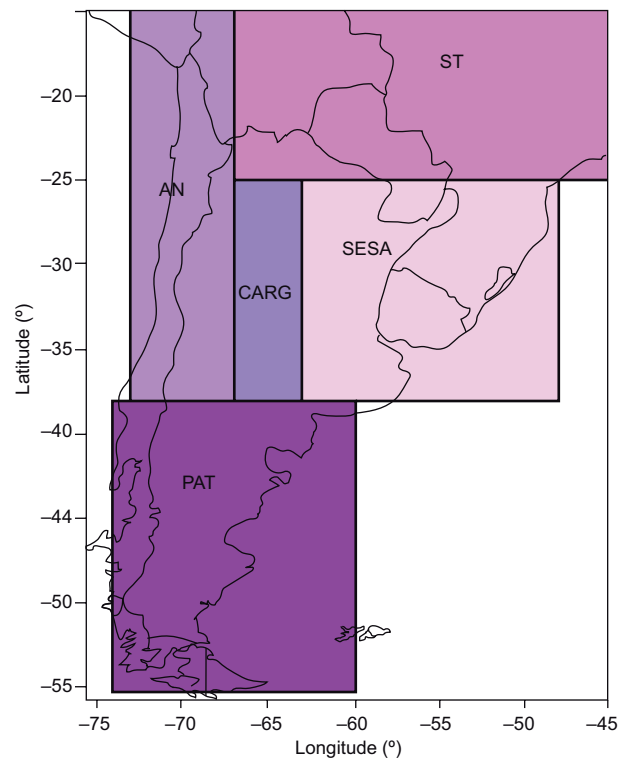


Fig. 8. Subregions defined for a more detailed analysis of the annual cycle of mean temperature. ST: subtropical; AN: Andes; CARG: Central Argentina; SESA: southeastern South America; PAT: Patagonia.

CMIP3 dataset (IPCC, 2007; Solomon *et al.*, 2007). Particularly, as projected by the MM5 regional model, warming during winter and spring is higher compared to summer and autumn.

Regardless of the selected subregion, the projected warming is lower than 4.5 °C in the late 21st century. Particularly, a warming between 2 and 4.5 °C is found over ST, between 1 and 3 °C over CARG, less than 2 °C over PAT, less than 3 °C over SESA, and 2-3 °C over the AN regions. Furthermore, the largest warming is located over central-eastern Argentina, Uruguay, southern Brazil and the subtropical latitudes (ST, CARG and SESA) during autumn and spring, and with a lesser degree during winter. For example, the maximum rising temperature (around 4.5 °C) was found over ST during spring. Even though results discussed previously are in general agreement with Núñez *et al.* (2008), the warming obtained here is of lower magnitude. Similar outcomes were also found by Blázquez *et al.* (2012) and Núñez and Blázquez (2014) from global climate models. As a final remark, in both in the late 21st century and in the late 20th

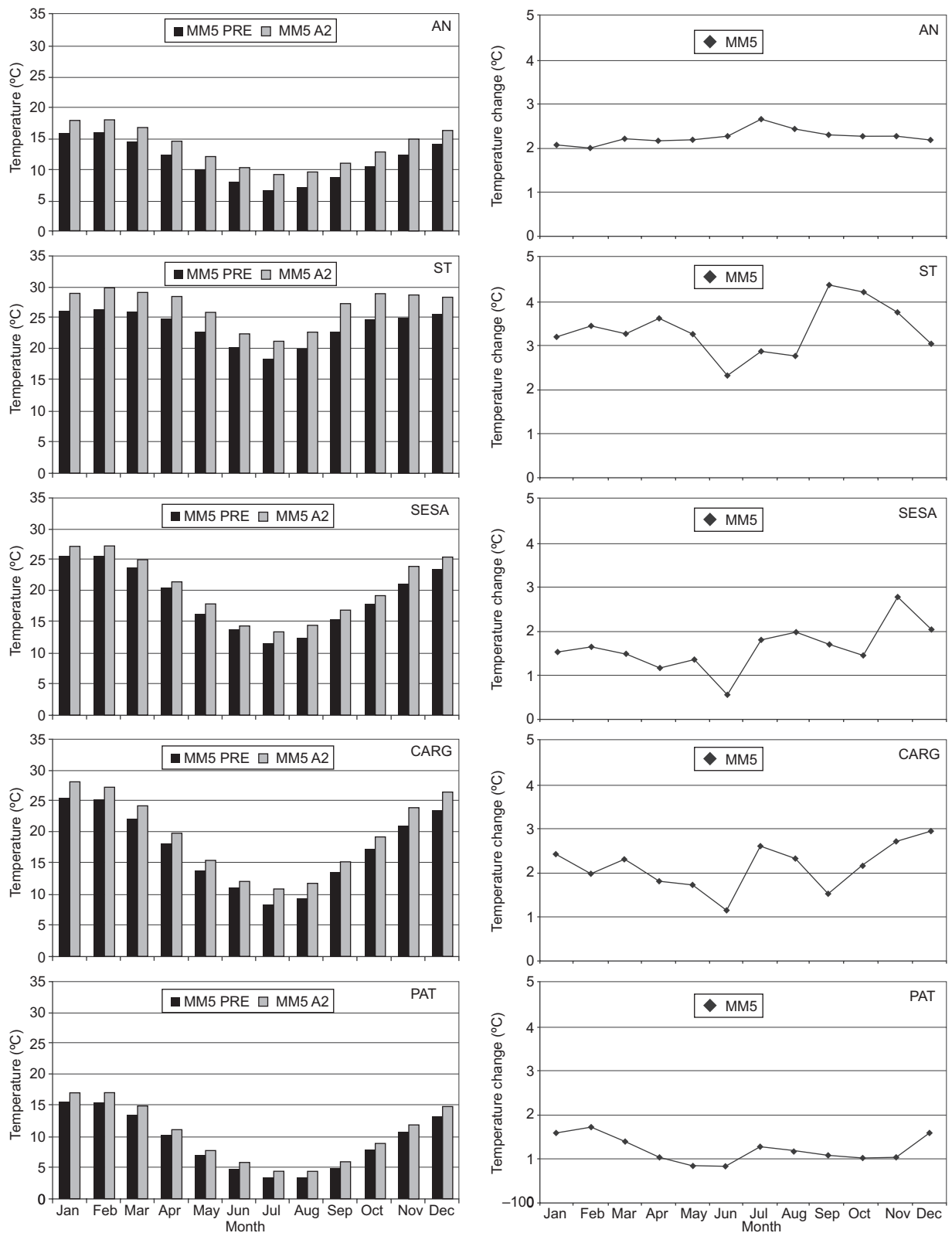


Fig. 9. Annual cycle of monthly mean temperature (°C) for present and future climate (left column) and projected temperature changes (right column). Subregions were defined in Figure 8.

century, the annual cycle of temperature is similar; the only difference is the higher temperature projected for future climate. Clearly, this has implications for the end of 21st century.

Figure 10 displays a diagram of the seasonal projected temperature changes for subregions defined previously.

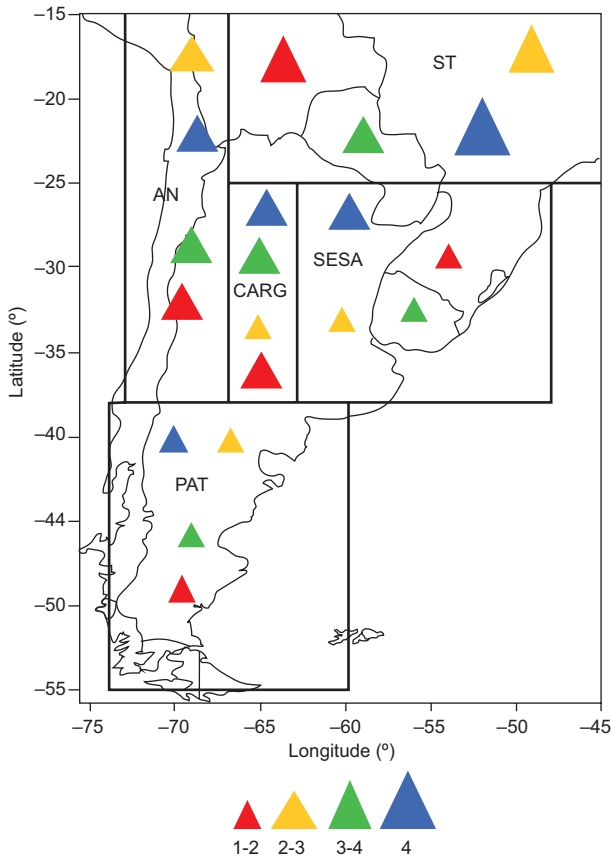


Fig. 10. Schematic graph of the MM5 seasonal projected changes ($^{\circ}\text{C}$) in the subregions defined in Figure 8. Red: DJF; orange: MAM; green: JJA; blue: SON. The size of the triangle indicates the magnitude of the temperature increase.

3.4 Projected changes in interannual variability

3.4.1 Spatial distribution

The analysis of changes in the interannual variability of precipitation and temperature over southern South America is shown in Figures 11 and 12, respectively. Due to the way in which the scenario is constructed with the global model (where the interannual variability of SST corresponds to observed SST modified with an added linear trend), a very large change in

interannual variability from the regional model is not expected. Therefore, regardless of the season, the spatial distribution of the interannual variability of precipitation and temperature for future climate is similar to current climate (1970-1989).

3.4.2 Precipitation

Figure 11 shows the spatial distribution of the interannual variability of precipitation as projected by the MM5 regional model, quantified in terms of the standard deviation (STDV) for future climate for DJF, MAM, JJA and SON (left column), and its differences from present climate (right column).

Changes in the interannual variability of precipitation are about $\pm 1 \text{ mm day}^{-1}$ throughout the study domain. In general, a slight increase in the interannual variability of precipitation is projected during summer, autumn and winter over central Argentina, Uruguay and the south of Brazil. However, larger negative changes (1 to 3 mm day^{-1}) in interannual variability are apparent during autumn, winter and spring over central Chile.

In summer, the model projects two areas of increase in the interannual variability of precipitation. One is located over central-western Argentina and the other over southern Brazil, Uruguay, eastern Argentina and southwestern Bolivia. Values of around 1 to 2 mm day^{-1} are found in small areas located over northern Uruguay and western-central Argentina.

In winter, a decrease in the interannual variability of precipitation is projected over most of the domain, with areas below 1 mm day^{-1} over northern Argentina, Paraguay and Bolivia, as well as over the Argentinean Patagonia, and values between 1 and 2 mm day^{-1} over central and southern Chile. An opposite pattern is projected over subtropical latitudes in eastern Brazil, Argentina and the border region of Chile, Argentina and Bolivia (reaching peaks of around 2 - 3 mm day^{-1}).

3.4.3 Mean temperature

Figure 12 shows the spatial distribution of the interannual variability of temperature for the late 21st century under the SRES A2 scenario (left column) and its difference from the current climate (right column). Regardless of the season, the spatial distribution of the projected interannual variability of temperature, basically shows a similar pattern as current climate (1970-1989). In general, a decrease in the interannual variability of temperature is projected over central

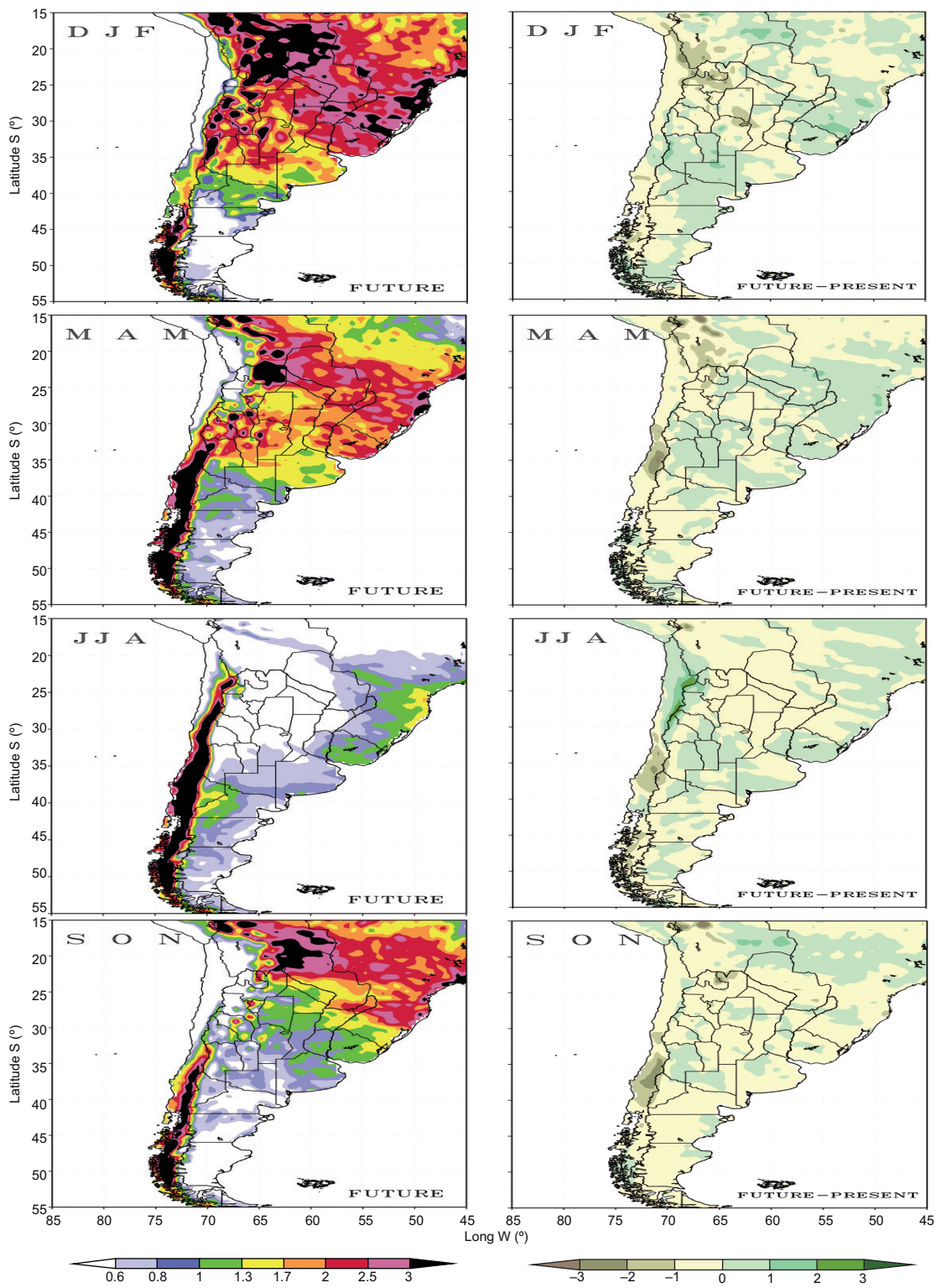


Fig. 11. Spatial distribution of the standard deviation (mm d^{-1}) for simulated precipitation for DJF, MAM, JJA, SON for future climate (left column) and changes related to present climate, difference between future and present climate periods (right column).

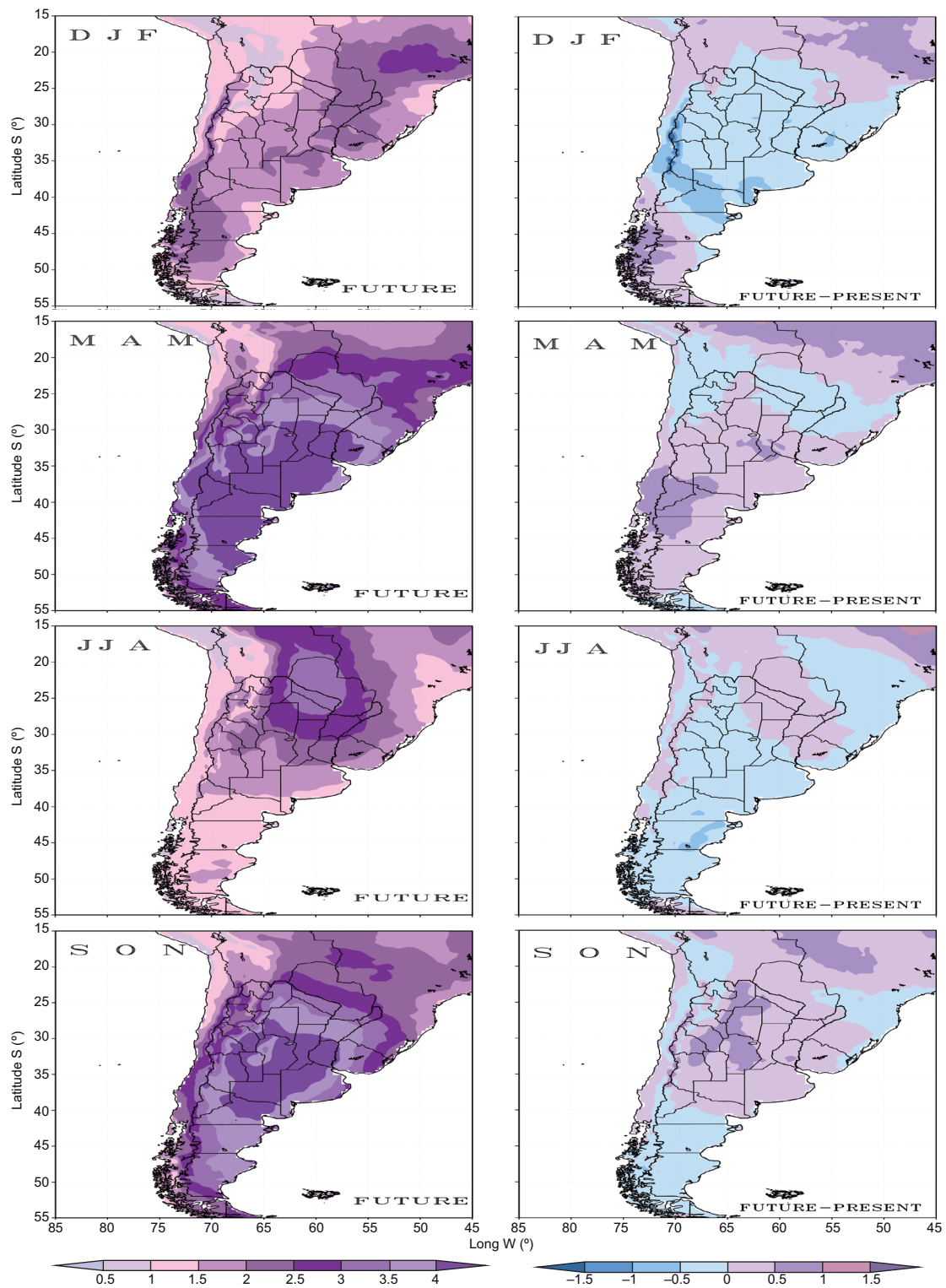


Fig. 12. Spatial distribution of the standard deviation (°C) for simulated temperature for DJF, MAM, JJA and SON for future climate (left column) and changes related to present climate, difference between future and present periods (right column).

and northern Argentina and Uruguay for summer and winter, while the opposite pattern is projected for autumn and spring.

During summer, a slight decrease of the interannual variability of temperature (around 0.5 °C) is projected over central and northern Argentina and Uruguay, with a maximum decrease projected over the north of the Argentinean Patagonia and central Chile. A small increase of the interannual variability of temperature (about 0.5 °C) over the south of the Argentinean Patagonia is apparent and another maximum increase (about 1 °C) is also apparent over the southwest tip of the continent. Moreover, another area of maximum interannual variability of temperature is projected over subtropical latitudes, with values not exceeding 1 °C.

In winter, the regional model MM5 projects a slight decrease in the interannual variability of temperature over central Argentina, southern Brazil, southern Chile and the Argentinean Patagonia; whereas a slight increase in the interannual variability of temperature (around 0.5 °C) is projected over northeastern Argentina, Paraguay, Uruguay and southern Bolivia.

In brief, a slight decrease in the interannual variability of temperature is projected for summer and winter over a large extent of Argentina. On the other hand, the MM5 model projects a slight increase in the interannual variability of temperature over central Argentina and the subtropical latitudes during transition seasons. The highest increases in interannual variability (around 1 °C) are found over central Chile during fall and over north central Argentina during spring.

Similar results concerning future changes in the interannual variability of precipitation and temperature were found over South America from GCMs by Giorgi and Bi (2005a, b); Giorgi and Diffengauhb (2008); Blázquez *et al.*, (2012), among others.

4. Summary and conclusions

Results reported here were based on one realization of the MM5 RCM forced by a single driving HadAM3H GCM. The analysis was focused on evaluating the spatial distribution of seasonal temperature and precipitation changes in terms of: (i) mean climate, (ii) interannual variability and (iii) the annual cycle of monthly temperature and precipitation changes for selected subregions. In addition,

in order to discuss the climate model robustness, a brief evaluation of uncertainties in regional model projections was also included. It has been found that the MM5 model agrees with high-resolution global projections and other regional projections, suggesting that it is an efficient tool to generate high-resolution scenarios of climate change over southern South America. Main conclusions can be summarized as follows:

1. The projected precipitation changes are positive or negative depending on the season and the region analyzed. A rainfall increase is projected over central Argentina, Uruguay and southern Brazil, especially during summer; whereas highest negative values are found over the Bolivian Plateau and southern Chile. The highest negative changes were found over central Chile in winter. In autumn, negative changes were found over southern and central Chile and highest positive changes were found over southern Brazil, Uruguay and central Argentina. For spring, the MM5 model projects a precipitation decrease over most of the study domain, however highest negative values were found at central and southern Chile. The statistical analyses conclude that projected precipitation changes conclude that variations in precipitation are significant (95% of confidence level) during summer over southern and central Chile; during winter over northern Argentina, Paraguay and southern Brazil, and central Argentina; whereas projected precipitation changes for intermediate seasons are not statistically significant over the whole study domain, except for spring over central and southern Chile.

2. The projected temperature increase over southern South America depends on the season and the examined area. Particularly, it ranges from 1.5 to 5.5 °C, being highest over subtropical latitudes. Seasonally, the warming is greatest during spring. The regional model projects a maximum temperature increase (between 4 and 5 °C) over tropical and subtropical latitudes mainly during spring, and a large warming (3 °C) over high latitudes (south of 40° S) during summer. Warming found over northeastern Argentina and Paraguay is similar for winter, summer and autumn, and it is greatest during spring. Regarding temperature, the projected changes are statistically

significant in all seasons (at a 95% confidence level) throughout the domain considered.

3. Concerning the projected annual cycle of precipitation, the MM5 model projects a rainfall increase (20-60%) during winter and spring, regardless of the selected region. Moreover, a precipitation decrease (lower than 60%) is projected for summer and autumn. A strong precipitation increase (higher than 60%) is projected during summer over SUA and during autumn over SEB. In general, a precipitation increase (above 60%) is projected over SUA during summer; over SEB during autumn, and over CA during spring. Concerning temperature, regardless of the selected region, a warming between 3 and 4 °C is projected for summer, winter and spring. The maximum warming (higher than 4 °C) is projected over the ST region during spring.

4. Regardless of the season, the MM5 RCM projects a low increase in the interannual variability of precipitation by the end of the 21st century under the SRES A2 scenario, compared to current climate. However, some seasonal characteristics are worth noting. A slight increase in the interannual variability of precipitation is projected over central Argentina and Uruguay for summer and winter. Especially for temperature, a slight decrease in the interannual variability is projected for summer and winter over a large extent of Argentina. However, the MM5 model projects a slight increase in the interannual variability of temperature during transition seasons, with highest values over central Chile in MAM and north central Argentina in SON.

5. The assessment of uncertainty in climate change projections revealed that temperature is a more reliable parameter than precipitation over southern South America by the end of the 21st century. Regardless the season, temperature projections are reliable over the entire study domain, particularly in subtropical latitudes, whereas the level of reliability for temperature projections over Uruguay-southern Brazil and east-central Argentina depends on seasonality. On the other hand, regardless the season projections of precipitation show less reliability in all evaluated areas. Hence, results obtained represent an important contribution to the characterization of uncertainty and reliability of regional climate

simulations over southern South America. To conclude, in spite of the different levels of reliability found for temperature and precipitation changes, the MM5 RCM can indeed provide valuable information for preliminary assessment of impact studies.

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