

Systematic review and climate change analysis in Los Lagos Region

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1 Climate change analysis

1.1 Climate change in southern Chile

In order to evaluate possible scenarios of climate change in the given area, we used a polygon comprising the Los Lagos and Los Ríos regions in Chile, and compared them using GCM compareR (Fajardo et al., 2020), considering Mean Anual Temperature and Annual Precipitation. The resulting scaled table of comparisson among futures was then use to select models to be used in the project.

We used the simple structure index (ssi) as implemented by the Vegan package (Oksanen et al., 2019; Dolnicar et al., 1999) to test what number of clusters (Between two and eight), was the best way to represent the 32 Compared GCMs. The best representation was five clusters, from each cluster the GCM closest to the centroid of the cluster was selected. The five selected GCMS were cesm1_bgc, gfdl_esm2g, ipsl_cm5a_lr, miroc_esm_chem and mpi_esm_lr and the selected GCMs together with the clusters are shown in figure 1.

1.1.1 Present climate conditions

Once the future GCM models were selected, the 30 seconds resolution maps were downloaded for 2070 and for present conditon from CHELSA (Karger et al., 2020). The present conditions for the Los Lagos and Los Ríos Region are shown in Figure 2, with a close up to the 10 Km buffer surrounding the park in 3. The region is a cold and humid area, with a range in the mean annual temperature from -5.6 to 13 degrees Celsius and a mean of 9.34, and a precipitation range between 858 to 4,537 and a mean of 2,092 mm a year.

The parks concidered in this proyect are in high altitude which leads to even cooler and wetter conditions, with a range from -5.6 to 12.6 and a mean of 8.12 degrees Celsius, and a precipitation range from from 1,176 to 3,410 and a mean of 2,134.51 mm as seen in figure 3.

1.2 Future scenarios

1.2.1 Future temperature

As stated above, the four GCMs chosen to explore and model future scenarios are cesm1_bgc, gfdl_esm2g, ipsl_cm5a_lr, miroc_esm_chem and mpi_esm_lr. Even when this models include relatively wetter models such as cesm1_bgc. On average for the whole region, the temperature will rise from 1.5 to 2.28 depending on the GCM (See figure 4), but in some areas, it the tempearture rise could be as high as 2.6 degrees Celsius.

As seen in figure 5, those changes are even higher within the parks and it's sorrounding areas, which means the effects of climate change might be even greater.

1.2.2 Future precipitation

The change in precipitation is predicted to be much more stark, with some areas decreasing in precipitation up to -696 as seen in figure 6, this is particularly worrisome, since the ecosystems that are prevalent in the area depend on high precipitation.

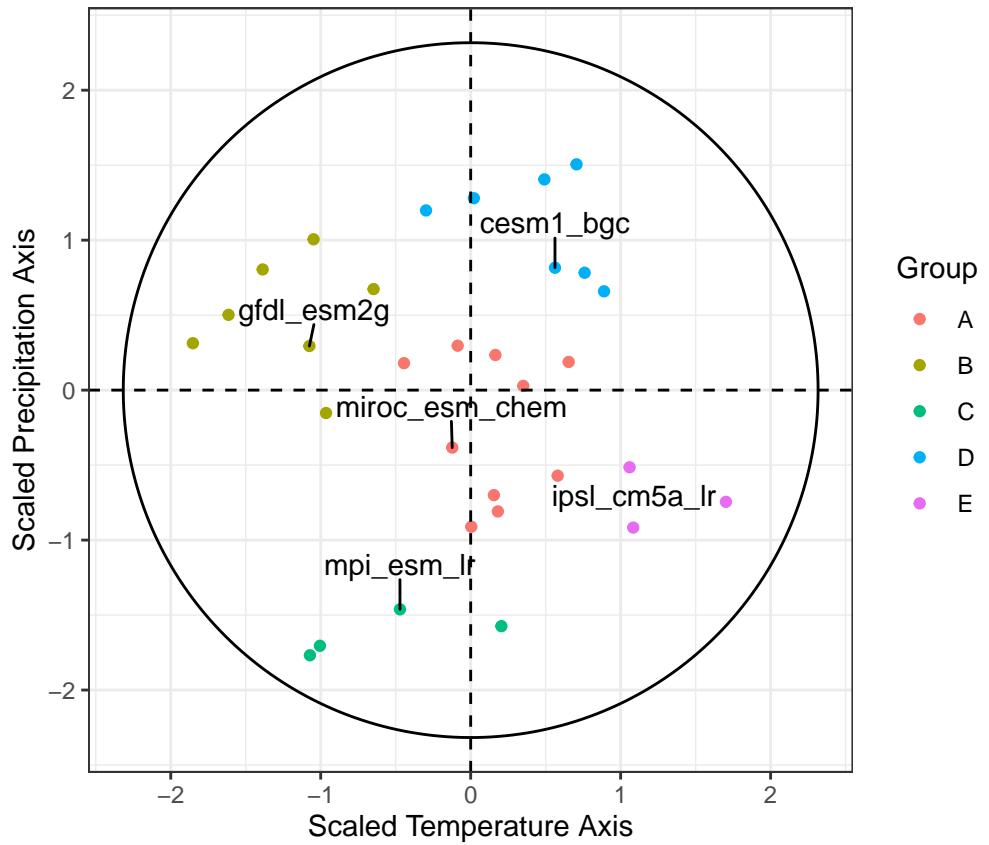


Figure 1: In this graph we can see the scaled temperature and precipitation axis, the center of the graph represent the ensemble of all models, the five groups represent the clusters selected using kmeans for five groups, the selected GCM of each cluster is shown with a label

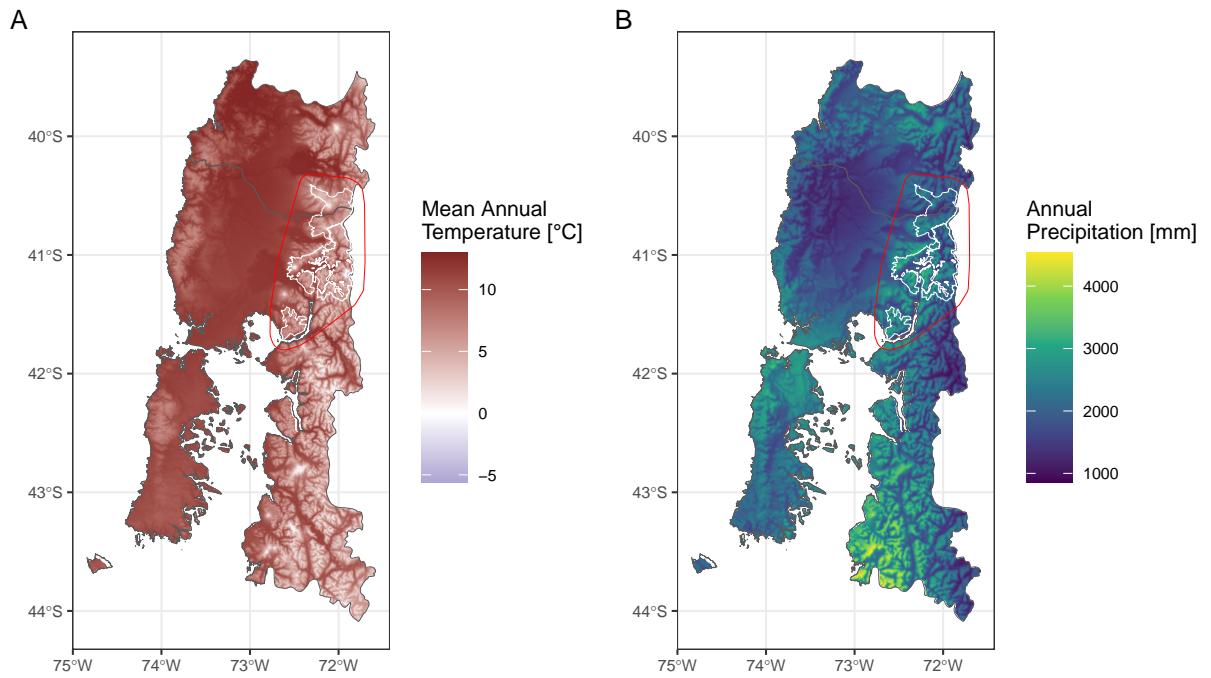


Figure 2: Mean annual temperature in °C (Facet A), and Annual Precipitation mm (Facet B)

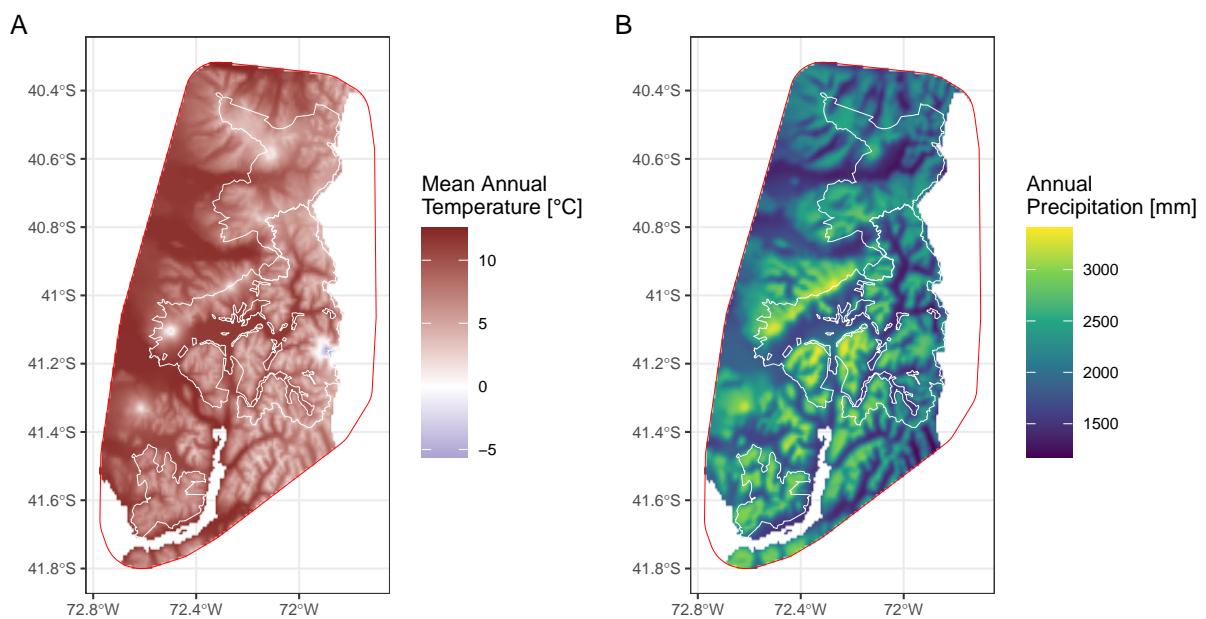


Figure 3: Mean annual temperature in °C (Facet A), and Annual Precipitation mm (Facet B) in the three studied national parks white lines, and a 10 km buffer red line

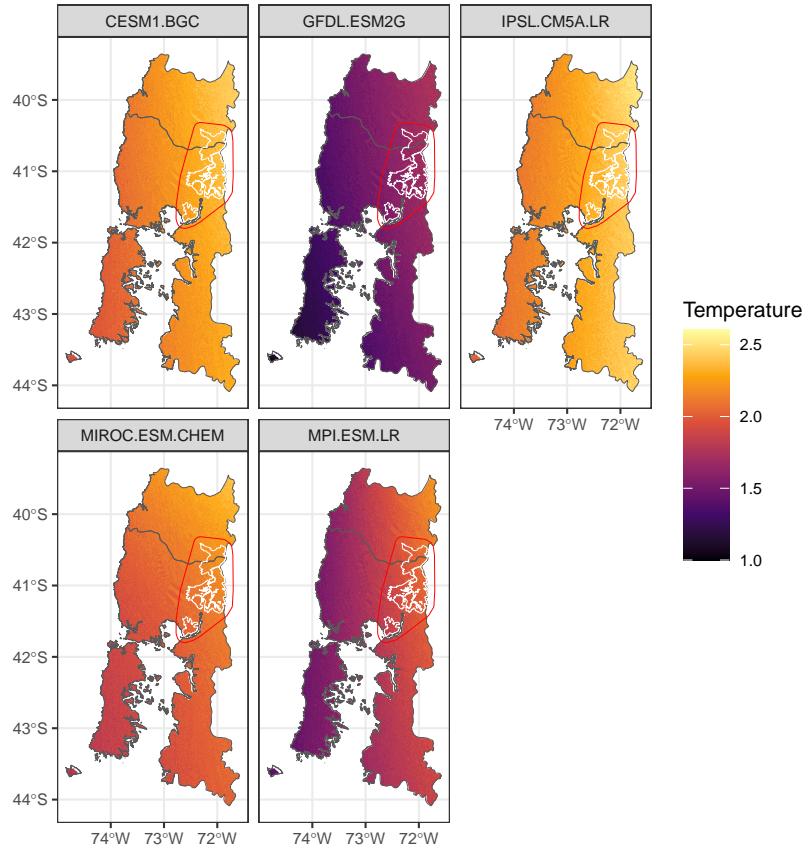


Figure 4: Changes in future mean annual temperature for the five selected GCMs, the red polygon surrounds the area of influence while the white line demarks the limits of the three protected areas in this project

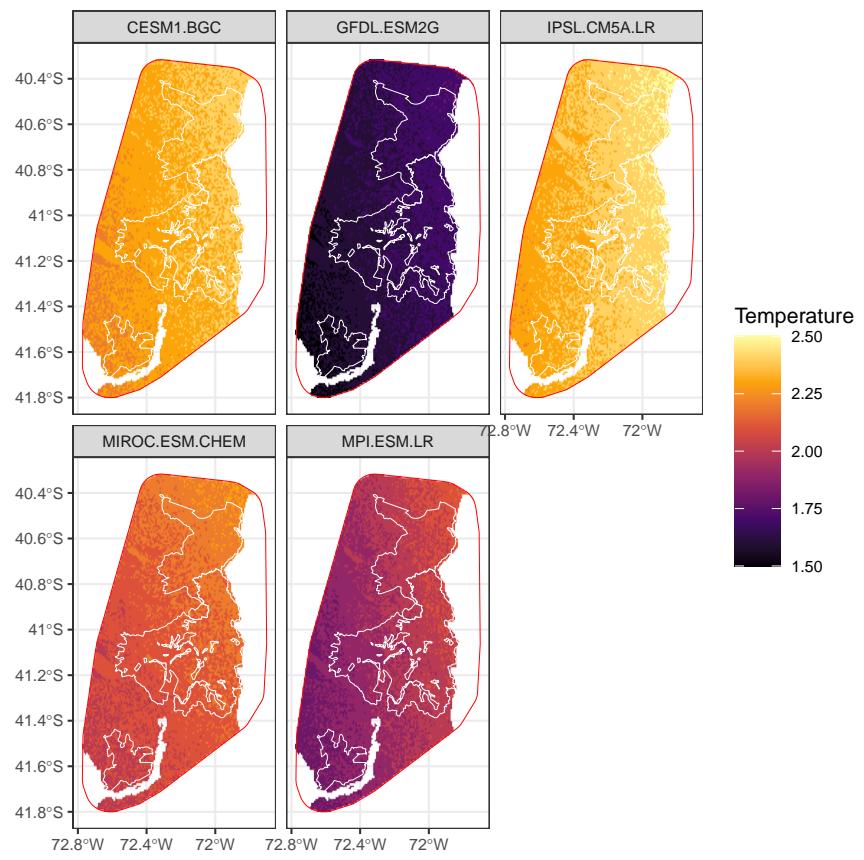


Figure 5: Temperature difference for all five GCMs, for the Close up of the area of influence of the parks

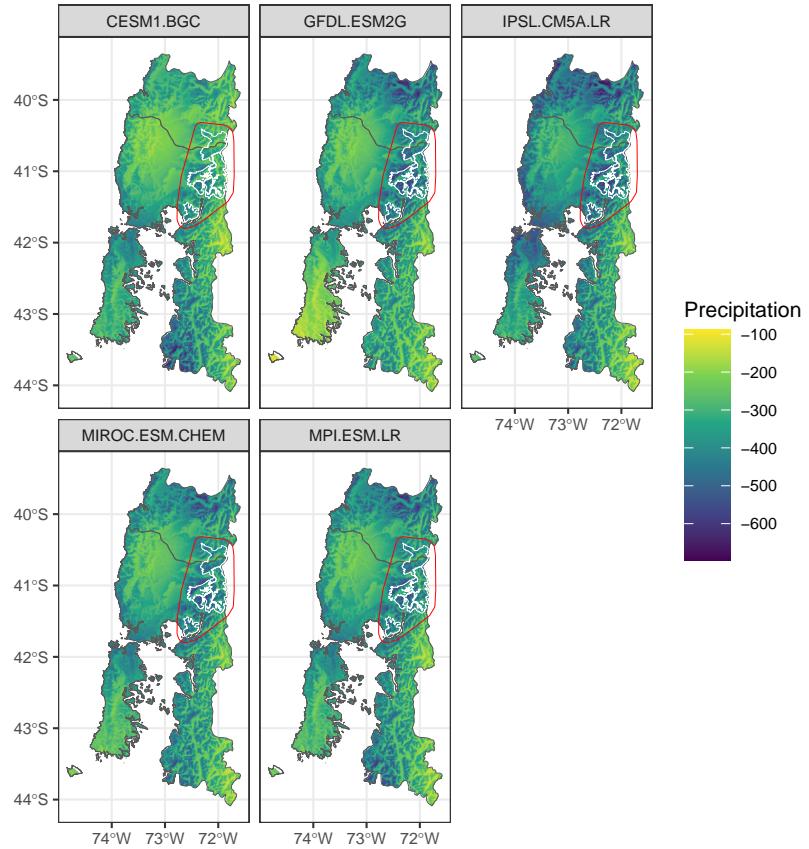


Figure 6: Changes in precipitation for the different GCMs, the red polygon surrounds the area of influence while the white line demarks the limits of the three protected areas in this project

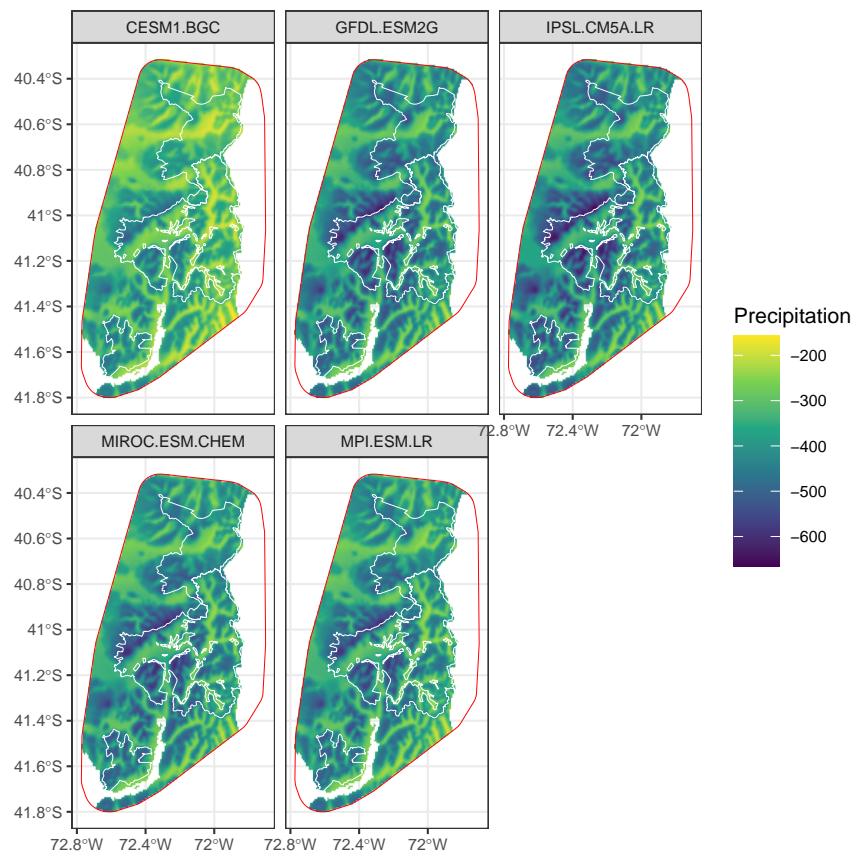


Figure 7: Precipitation difference for all five GCMs, for the Close up of the area of influence of the parks

The range of changes between GCMs will be from -376.99 to -313.23 mean annual precipitation for the whole area. Again the areas where there is going to be a higher drop in precipitation are mostly inside of the national parks, as seen in figure 7, with changes in the mean annual precipitation within the are of -412.29 for the wettest models, and -320.96 for the driest models.

1.3 Vegetational formation

We used Luebert and Pliscoff (2009) to check current vegetational formations as seen in figure 8

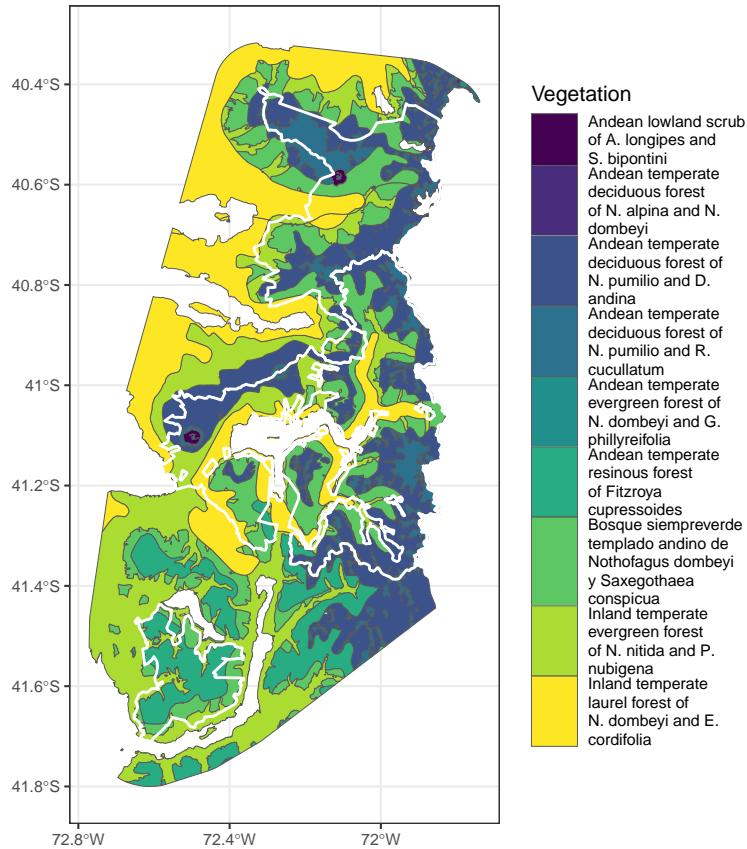


Figure 8: Vegetational formation in the influence area of the parks

1.3.1 Climatic analogues

As seen above, changes in precipitation will be very large. Because of that, we checked which areas in the present have the same climate as it is predicted to be in the future in the national parks and surrounding areas using the analogous R package (Ramírez Villegas et al., 2011). After doing that, the current vegetational formations of those areas were checked in order to determine possible future vegetation in the area. With that we might better understand the biological consequences of climate change.

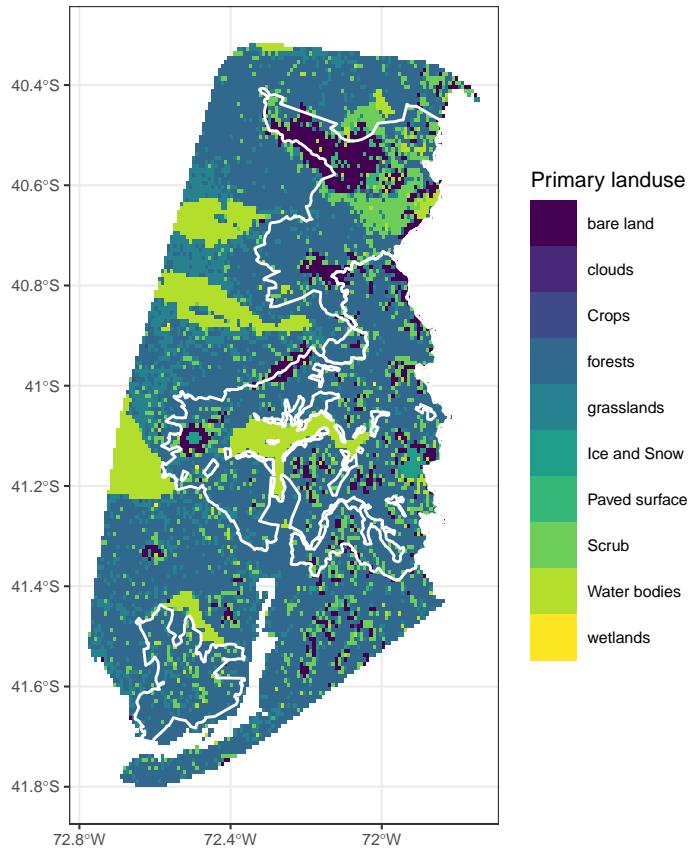
Currently in the National Parks and surrounding areas, the five most common Vegetational formations are Deciduous forest, Evergreen forest, Laurel forest, Low altitude scrubland and Resinous forest, and in the future, new vegetational formations that would be part of the five most common ones would include Absolute desert, Desert scrub, Sclerophyllous forest and Thorny forest. Of the current five most common formations, the ones that would be reduced as far as that they would either be lost completely or become very rare are Evergreen forest, Low altitude scrubland and Resinous forest

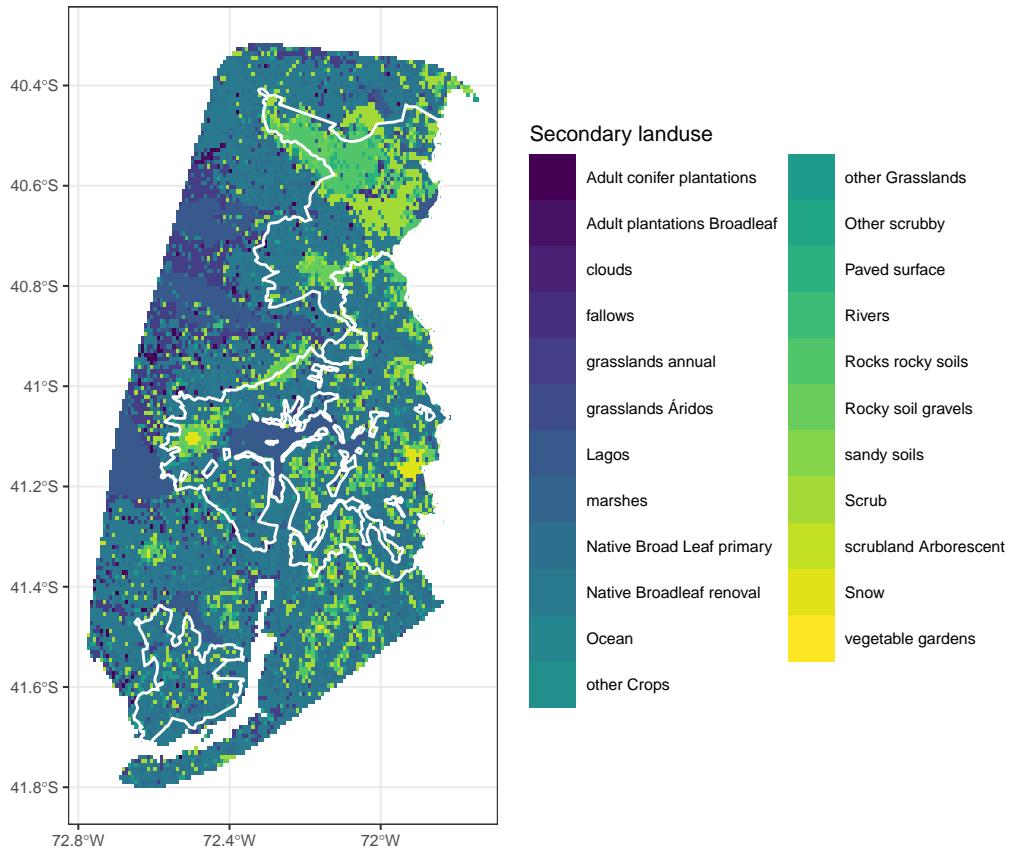
Table 1: Top five vegetation formations for the present and for the different GCMs according to the analogous climate analysis

Rank	Present	CESM1-BGC	GFDL-ESM2G	IPSL-CM5A-LR	MIROC-ESM-CHEM	MPI-ESM-LR
1	Evergreen forest	Deciduous forest	Deciduous forest	Deciduous forest	Deciduous forest	Deciduous forest
2	Deciduous forest	Sclerophyllous forest	Sclerophyllous forest	Sclerophyllous forest	Sclerophyllous forest	Sclerophyllous forest
3	Laurel forest	Desert scrub	Thorny forest	Desert scrub	Desert scrub	Desert scrub
4	Resinous forest	Thorny forest	Desert scrub	Thorny forest	Thorny forest	Thorny forest
5	Low altitude scrubland	Laurel forest	Laurel forest	Absolute desert	Laurel forest	Laurel forest

1.4 Land use

We used Zhao et al. (2016) landuse layer, specifically developed for Chile. This is a 30 meter resolution, that





Primary landuse	Percentage
forests	59.17
grasslands	12.95
Scrub	10.24
Water bodies	9.34
bare land	6.93

Secondary landuse	Percentage
Native Broadleaf renova	39.45
Native Broad Leaf primary	18.63
Scrub	10.11
Lagos	8.97
grasslands annual	8.43
Rocks rocky soils	5.04
other Grasslands	4.40
Rocky soil gravels	1.85

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