

# IPCC Climate Scenarios: how our decisions are shaping our future

Course: 6950 - Computer based tools and applications

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## 1 Introduction

The IPCC (Intergovernmental Panel on Climate Change) was created by the United Nations Environment Programme on 1988, dedicated to provide policymakers with regular scientific assessments on climate change, critical to prepare for potential climate risks and to develop mitigation and adaptation strategies for companies and governments all over the world. The IPCC does not conduct its own research, rather it identifies where there is agreement on the scientific community about climate change, reporting it neutrally but staying policy-relevant (IPCC, 2019).

A climate scenario is an assumption about political changes and their effect on climate change. There are multiple climate scenarios described after the Shared Socio-economic Pathways (SSP), which include plausible socio-political and technological scenarios and its consequences on every variable playing a role on climate change. Under the Coupled Model Intercomparison Project Phase 6 (CMIP6) and through Copernicus Climate Data Store, multiple climate variables are available for free download as gridded data for historic or SSP experiments (Government of Canada, 2016).

This project will be focusing on the reduction of snow depth in Newfoundland and Labrador, Canada, since climate change and generalized global warming have imposed increasing stress on the accumulation of snow, crucial for climate regulation.

## 2 Methodology

The download of the data can be made through two methods: direct download from the webpage, selecting the appropriate parameters, which are the ones in the list below, or through an API call.

- Temporal resolution: either daily, monthly, yearly or fixed. Depends on the dataset.
- Experiment: one of the multiple available climate scenarios.
- Variable: the desired climate variable.
- Level: atmospheric height in hPa, may not be available for all variables.

- Model: the computational model used to obtain the variable. Not all models contain all variables.
- Year: year range for the data, from 1850 to 2300.
- Month: month range for the data.
- Day: day range for the data.
- Geographical area: bounding geographical coordinates of interest.

The methodology used in this project is the API call, since it is a programmatic way to easily get data from the database. The webpage offers the exact call we need to make to the API in order to download specific data, and following the pattern we can program a class that allows us to enter the parameters in the form of strings or vectors for ranges and get back the right data. Once downloaded, the data comes in NetCDF format, which is a very efficient format to carry large amounts of data in a relatively small file size.

The variables obtained through the API call are as follow:

- Near-surface air temperature
- Snow depth

All of the data is downloaded for the climate scenarios called Nationally Determined Contributions (NDC, SSP2-4.5) and Current Policies (SSP3-7.0). The scenario names NDC and Current Policies are based in the nomenclature by the Network for Greening the Financial System (NGFS, 2025), and the codes SSP2-4.5 and SSP3-7.0 are named under the Shared Socioeconomic Pathways as defined in the Intergovernmental Panel on Climate Change sixth assessment in 2023 (IPCC, 2023). These two scenarios are chosen because they are the two more likely scenarios to become reality.

To load such data, we aid ourselves with the Python library **Xarray**, which is designed to handle large datasets, often with special functions for climate science, while keeping compatibility with numerical and data handling libraries like Numpy and Pandas. Xarray follows more closely the conventions used by Pandas, using as a base object the so-called **DataArray**, which behaves similar to a Pandas **DataFrame**. The main difference is that in DataArrays, each point is associated with a set of coordinates, either spatial or time coordinates, so we can retrieve the data for a specific geographical point with precision.

It is worth noting that many of the available models are inherently different in how they organize the data inside the files, especially the coordinate naming conventions. For example, the model CESM2 (USA) uses names such as **lon** and **lat** for the longitude and latitude coordinates, and vectors ranging from -90 to 90 for latitude and from 0 to 360 for longitude, which are easy to work with. Other models uses an "ij" indexing system, which contains a meshgrid for the coordinates that makes

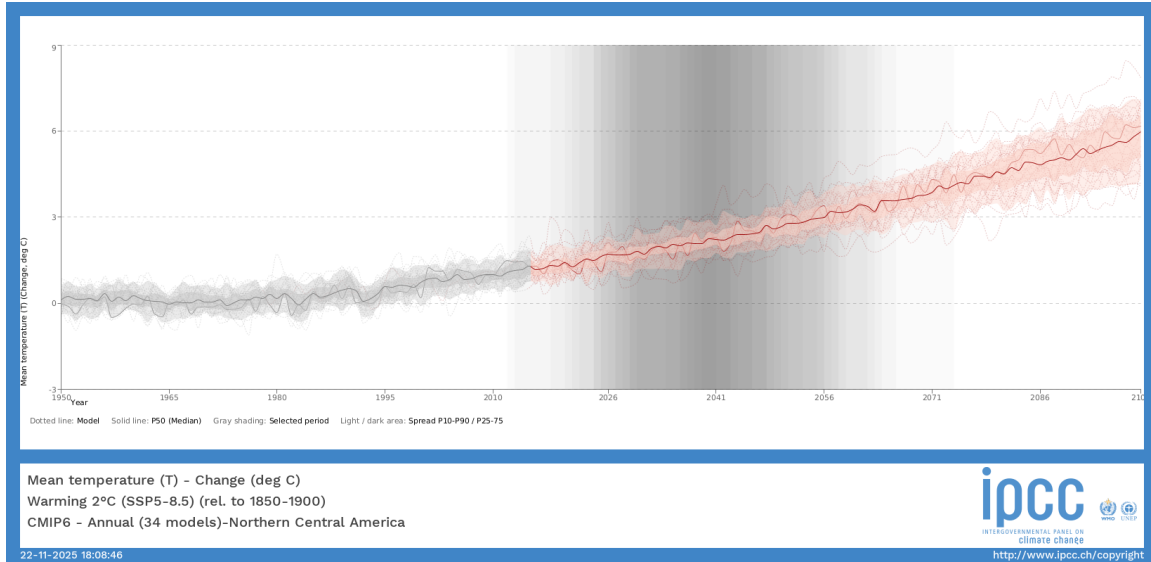


Figure 1: Results for all 34 available models on the CMIP Yearly Mean Temperature. Each line represents a model

it more difficult to access. That is why we select the model CESM2 (USA) for all variables.

Different models give slightly different results for the same variable, and choosing one near the average value is crucial. The model CESM2 (USA) is one of these, visible in Figure 1 in a slightly bolder line other than the mean value.

The final data thus is composed of near-surface air temperature and snow depth for the entire world, with a spatial resolution of 1 degree by 1 degree per pixel, monthly for the years 1950-2015 for the historical records and 2015-2100 for the projections. Despite being 2025 as of the writing of this report, the experiments run by the CMIP6 were created during 2016 and has been the last cutoff of modelling.

There are many ways to visualize this data, but being 2D data, one of the most intuitive is create maps, overlayed on the outline of the countries to provide a frame of reference. The data for the world boundaries can be obtained as an ESRI Shapefile from public datasets (Huwise, 2019) and loaded using the Python library Geopandas, which is especially designed to handle geographical data. The Shapefile format is made of a table of attributes, so it can be thought of a DataFrame with location and geometry attributes, perfectly readable by a library like Geopandas, and loaded to a map to draw the lines of the countries boundaries, which will be the primary basemap.

To create a comparison of the changes across the years on the climate variables we are using, we need to get rid of possible outliers in the data by taking the mean over the winter season. We choose the winter because snow concentration only happens in this season, and including other seasons could bias the results. To select only the winter season, we select the months December, January and February, discard the

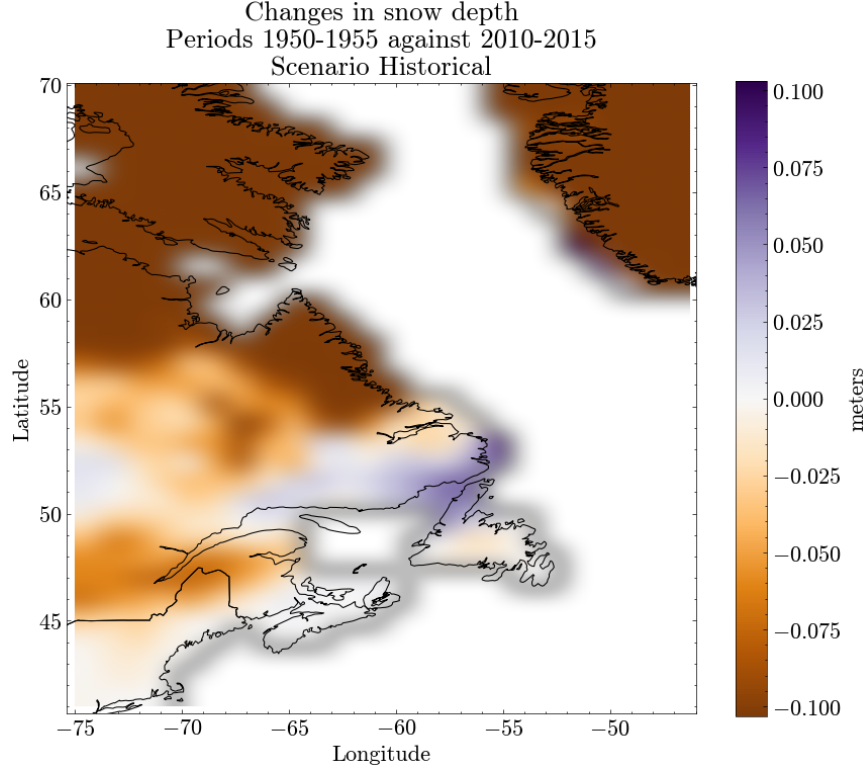


Figure 2: Historical 5-year average snow depth changes from 1950 to 2015

rest and then take the 5 year average, creating multiple 5-year periods that are easier to compare to one another and eliminating useless data in the process. We perform this operation for all the datasets, historical and projected.

To ease the creation of maps for multiple datasets, we create a class that allows us to pass the data and create figures from it afterwards by calling a simple function. The class is called `PlotObject` and takes an Xarray as input, and the variables and appropriate colormaps or ranges are automatically extracted from the input array.

Apart from creating maps, we can look at the timeseries of the yearly average concentration of snow in the whole area of interest, comparing how it has changed over the years and how it is projected to change under the different climate scenarios. We create line plots that show

### 3 Results

### References

Government of Canada. (2016). CMIP6 and shared socio-economic pathways overview. <https://climate-scenarios.canada.ca/?page=cmip6-overview-notes>

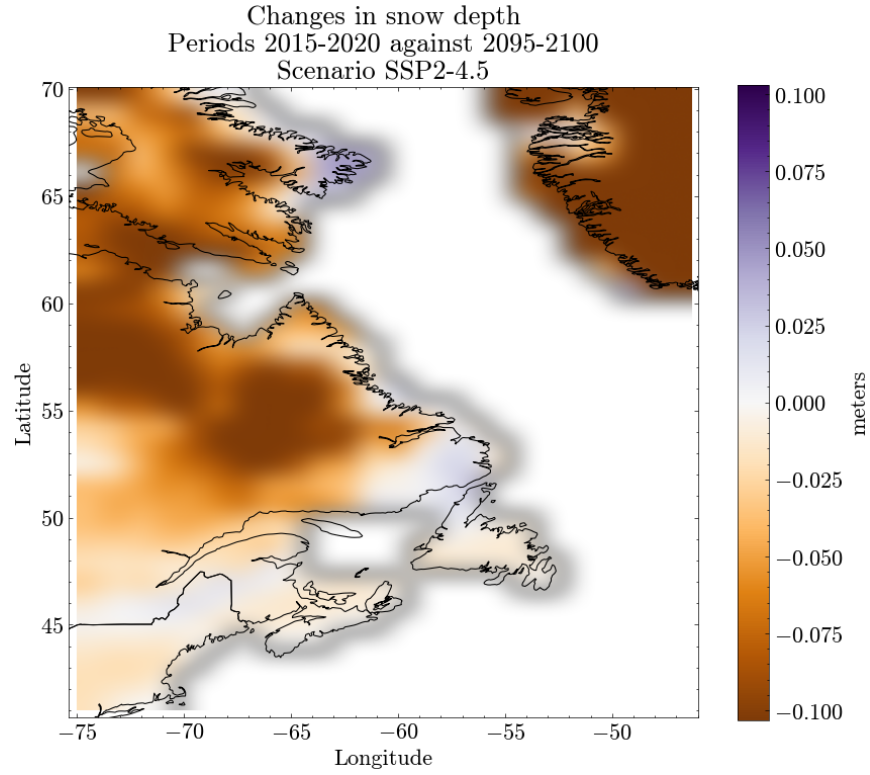


Figure 3: Historical 5-year average snow depth changes towards 2100 under the climate scenario SSP2-4.5

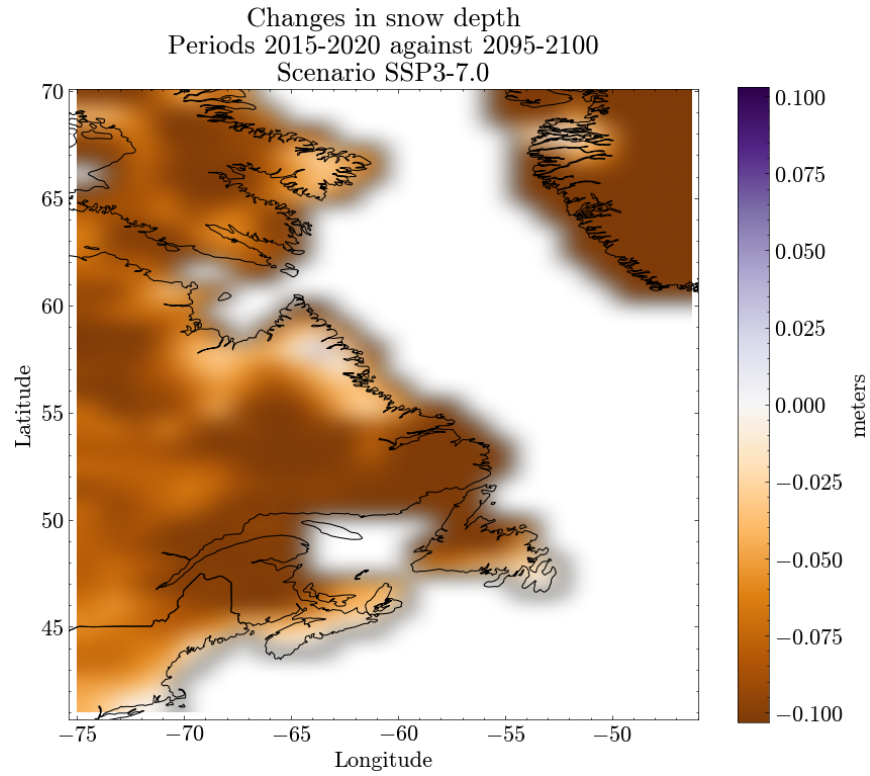


Figure 4: 5-year average snow depth changes towards 2100 under the climate scenario SSP3-7.0

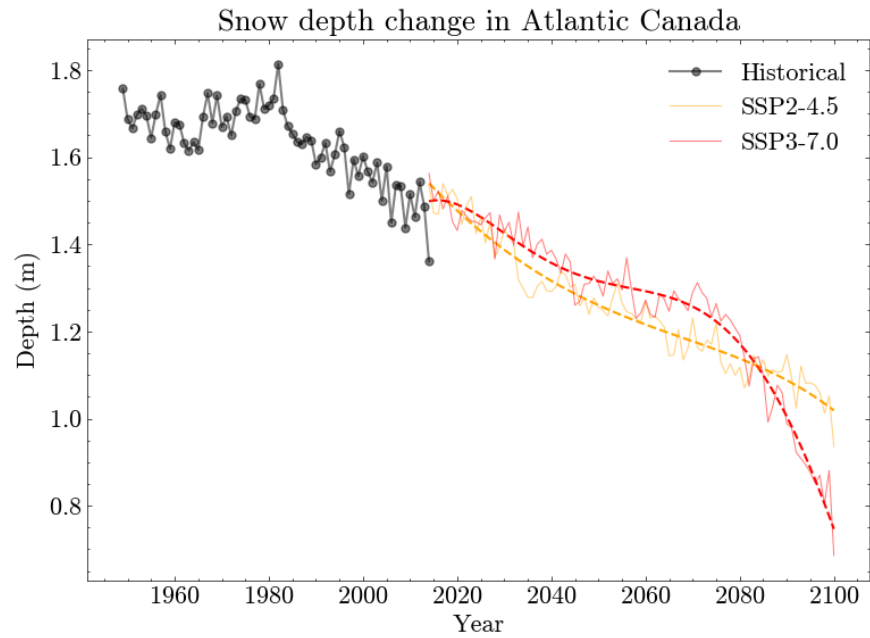


Figure 5: Yearly average snow depth in Atlantic Canada, showing the projections of two different climate scenarios.

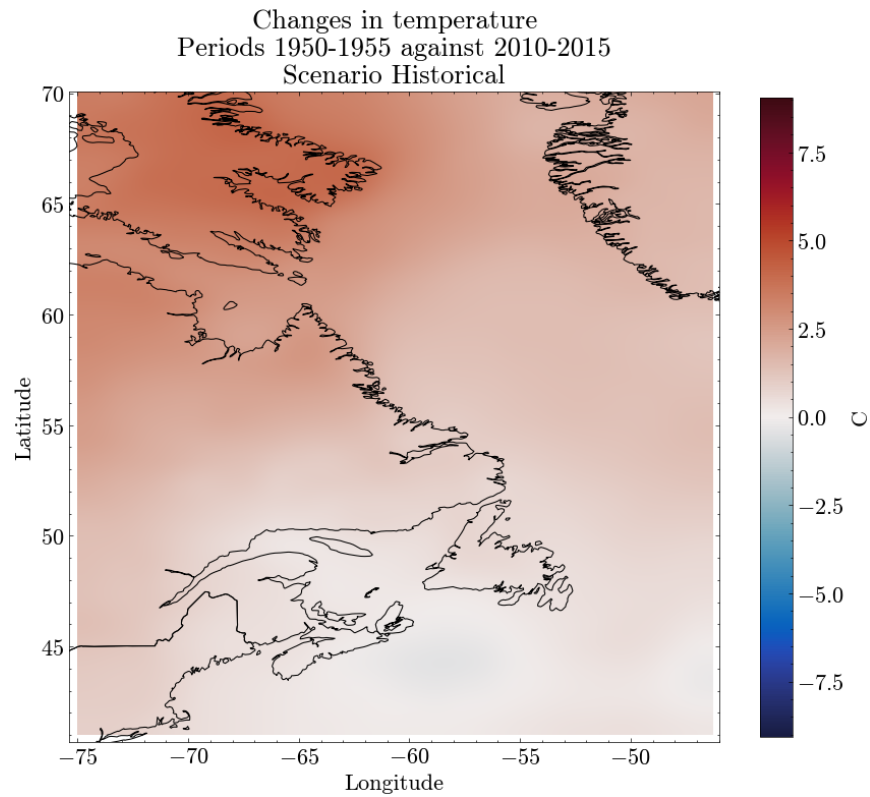


Figure 6: Historical 5-year average temperature changes from 1950 to 2015

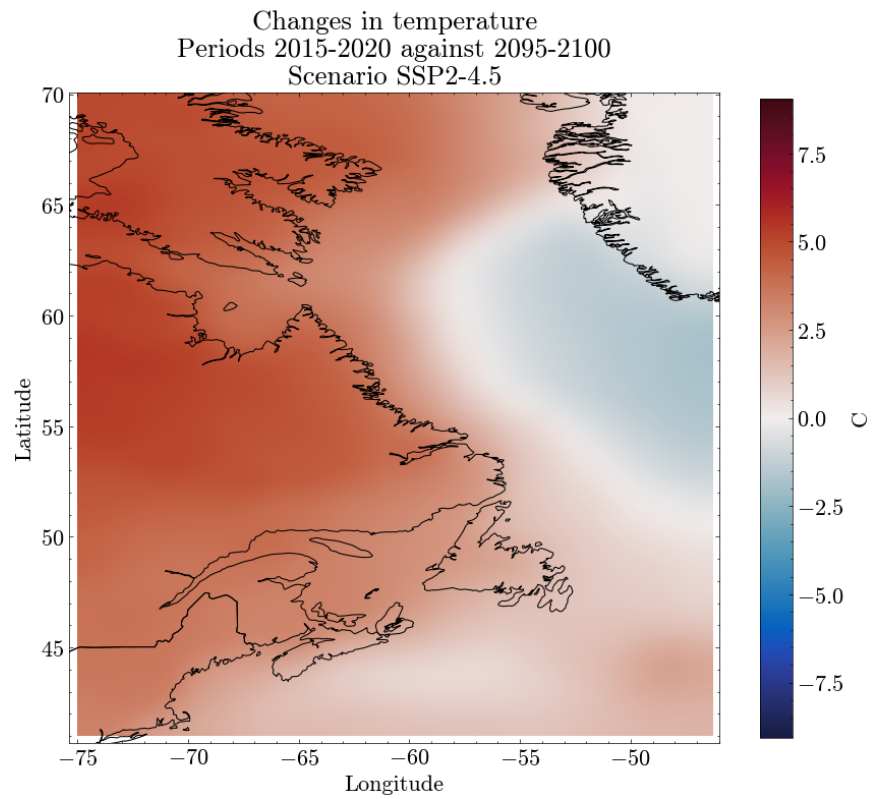


Figure 7: Historical 5-year average temperature changes towards 2100 under the climate scenario SSP2-4.5

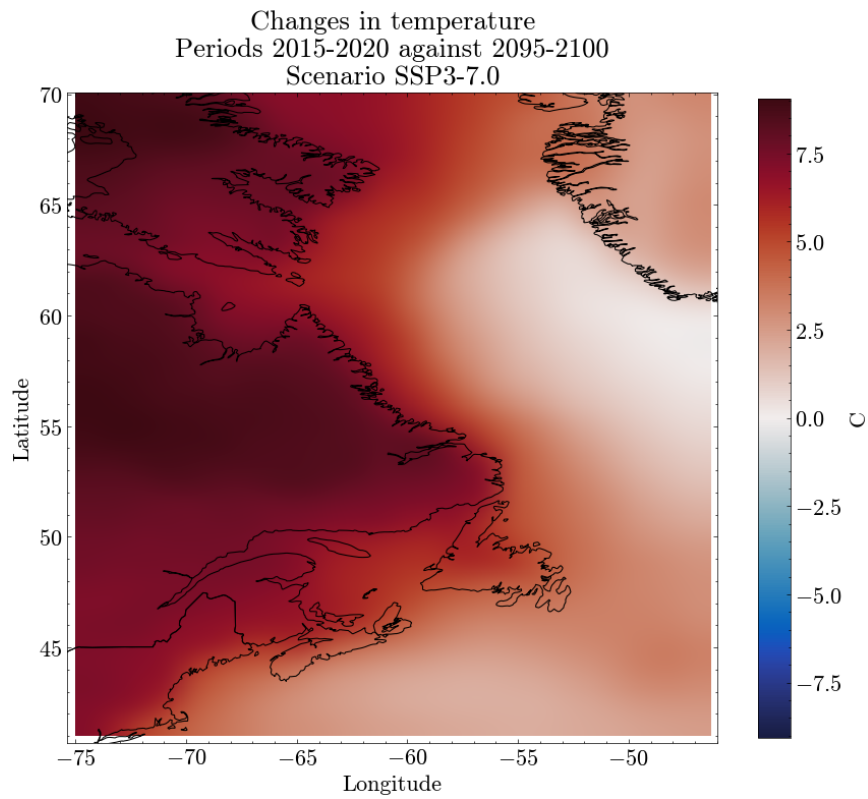


Figure 8: 5-year average temperature changes towards 2100 under the climate scenario SSP3-7.0

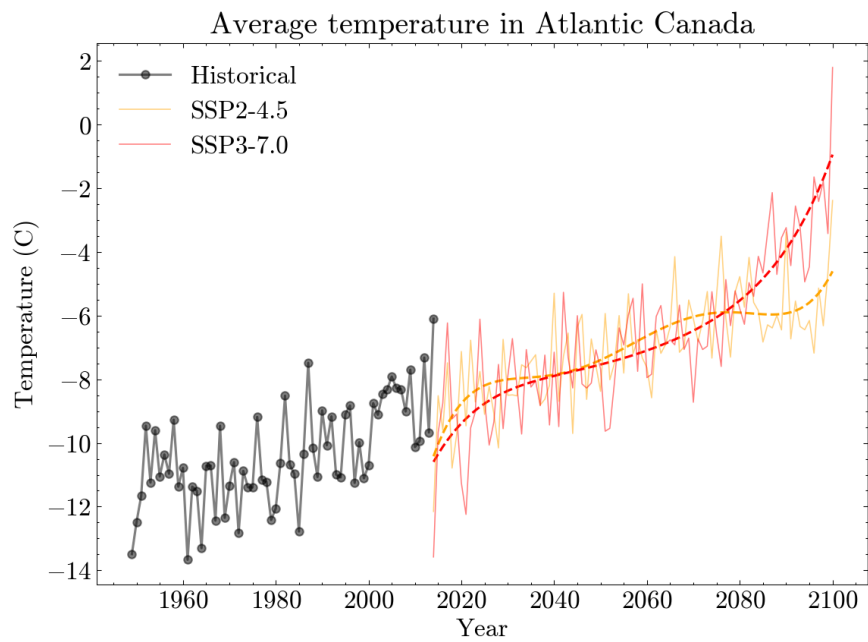


Figure 9: Yearly average temperature in Atlantic Canada, showing the projections of two different climate scenarios.



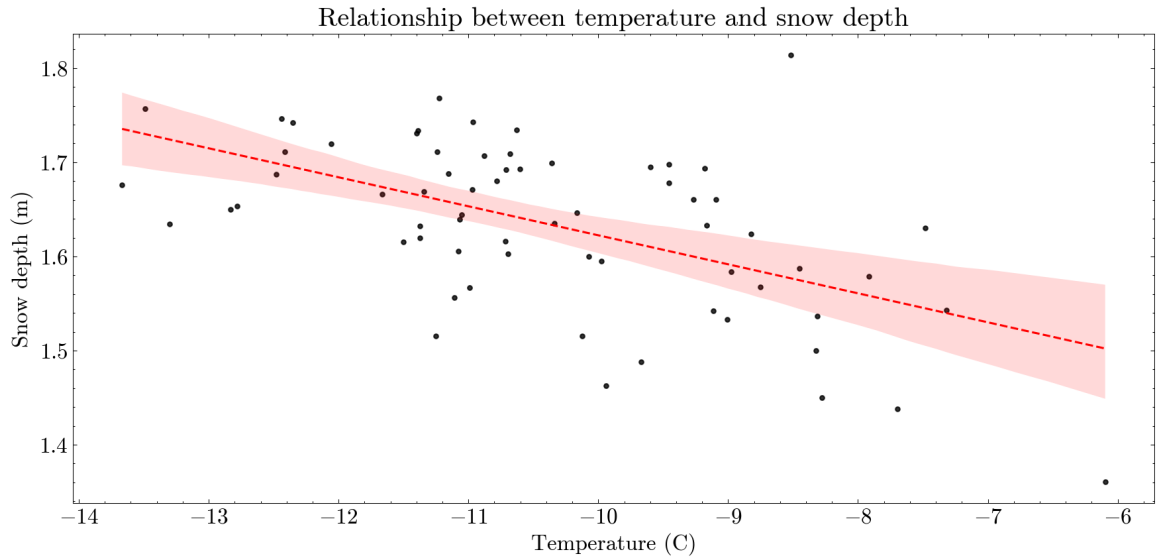


Figure 10: Linear regression plot of temperature vs snow depth. Notice the inverse relationship.

Huwise. (2019). World Administrative Boundaries - Countries and Territories. <https://public.opendatasoft.com/explore/assets/world-administrative-boundaries/>  
 IPCC. (2019). Intergovernmental panel on climate change. <https://www.ipcc.ch/>  
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