

Session 2 Code

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Install packages

```
#install.packages("terra")
#install.packages("dplyr")
#install.packages("sf")
#install.packages("ggplot2")
#install.packages("dismo")
#install.packages("rasterVis")
#install.packages("reshape")

## Import packages
library(terra)
library(dplyr)
library(sf)
library(ggplot2)
library(dismo)
library(rasterVis)
library(reshape)
library(RColorBrewer)
```

Part 1: Temperature Data (one model)

Set working directory

```
#setwd("C:/R Code/Training/ICCB/")  
getwd() # get work directory
```

```
[1] "/Users/scottforrest/Library/CloudStorage/OneDrive-QueenslandUniversityofTechnology/PhD
```

```
dir() # list folders in the work directory
```

```
[1] "data"  
[2] "ICCB2025_Session2_ClimateProjections.pdf"  
[3] "README.md"  
[4] "scripts"  
[5] "session_2_code_files"  
[6] "session_2_code.html"  
[7] "session_2_code.qmd"  
[8] "session_2_code.rmarkdown"  
[9] "session_2_home.qmd"
```

```
dir("data/annual/") # List files in subdirectory
```

```
[1] "pr_ACCESS-ESM1-5_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[2] "pr_ACCESS-ESM1-5_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[3] "pr_ACCESS-ESM1-5_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[4] "pr_EC-Earth3_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[5] "pr_EC-Earth3_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[6] "pr_EC-Earth3_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[7] "pr_GFDL-ESM4_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[8] "pr_GFDL-ESM4_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[9] "pr_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[10] "tas_ACCESS-ESM1-5_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[11] "tas_ACCESS-ESM1-5_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[12] "tas_ACCESS-ESM1-5_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[13] "tas_EC-Earth3_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[14] "tas_EC-Earth3_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[15] "tas_EC-Earth3_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[16] "tas_GFDL-ESM4_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[17] "tas_GFDL-ESM4_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"  
[18] "tas_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc"
```

```
files=dir("data/annual/")
files[1]
```

Load file and query data (working with one model)

```

class      : SpatRaster
dimensions : 205, 176, 120  (nrow, ncol, nlyr)
resolution : 0.1, 0.1  (x, y)
extent      : 137.45, 155.05, -29.45, -8.95  (xmin, xmax, ymin, ymax)
coord. ref. : lon/lat WGS 84 (CRS84) (OGC:CRS84)
source      : tas_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc
varname     : tas_annual (Seasonal average of Near-Surface Air Temperature (Annual))
names       : tas_annual_1, tas_annual_2, tas_annual_3, tas_annual_4, tas_annual_5, tas_annual_6
unit        :          degC,          degC,          degC,          degC,          degC,
time (days) : 1981-01-01 to 2100-01-01 (2 steps)

```

```
xFromCol(tas)
```

3

```
yFromRow(tas)
```

```
[1] -9.0 -9.1 -9.2 -9.3 -9.4 -9.5 -9.6 -9.7 -9.8 -9.9 -10.0 -10.1
[13] -10.2 -10.3 -10.4 -10.5 -10.6 -10.7 -10.8 -10.9 -11.0 -11.1 -11.2 -11.3
[25] -11.4 -11.5 -11.6 -11.7 -11.8 -11.9 -12.0 -12.1 -12.2 -12.3 -12.4 -12.5
[37] -12.6 -12.7 -12.8 -12.9 -13.0 -13.1 -13.2 -13.3 -13.4 -13.5 -13.6 -13.7
[49] -13.8 -13.9 -14.0 -14.1 -14.2 -14.3 -14.4 -14.5 -14.6 -14.7 -14.8 -14.9
[61] -15.0 -15.1 -15.2 -15.3 -15.4 -15.5 -15.6 -15.7 -15.8 -15.9 -16.0 -16.1
[73] -16.2 -16.3 -16.4 -16.5 -16.6 -16.7 -16.8 -16.9 -17.0 -17.1 -17.2 -17.3
[85] -17.4 -17.5 -17.6 -17.7 -17.8 -17.9 -18.0 -18.1 -18.2 -18.3 -18.4 -18.5
[97] -18.6 -18.7 -18.8 -18.9 -19.0 -19.1 -19.2 -19.3 -19.4 -19.5 -19.6 -19.7
[109] -19.8 -19.9 -20.0 -20.1 -20.2 -20.3 -20.4 -20.5 -20.6 -20.7 -20.8 -20.9
[121] -21.0 -21.1 -21.2 -21.3 -21.4 -21.5 -21.6 -21.7 -21.8 -21.9 -22.0 -22.1
[133] -22.2 -22.3 -22.4 -22.5 -22.6 -22.7 -22.8 -22.9 -23.0 -23.1 -23.2 -23.3
[145] -23.4 -23.5 -23.6 -23.7 -23.8 -23.9 -24.0 -24.1 -24.2 -24.3 -24.4 -24.5
[157] -24.6 -24.7 -24.8 -24.9 -25.0 -25.1 -25.2 -25.3 -25.4 -25.5 -25.6 -25.7
[169] -25.8 -25.9 -26.0 -26.1 -26.2 -26.3 -26.4 -26.5 -26.6 -26.7 -26.8 -26.9
[181] -27.0 -27.1 -27.2 -27.3 -27.4 -27.5 -27.6 -27.7 -27.8 -27.9 -28.0 -28.1
[193] -28.2 -28.3 -28.4 -28.5 -28.6 -28.7 -28.8 -28.9 -29.0 -29.1 -29.2 -29.3
[205] -29.4
```

Adding missing year values to the data

```
dates = seq(as.Date("1981-01-01"), as.Date("2100-12-01"), by="year")
names(tas) = dates # fixing the time data in the NetCDF
tas
```

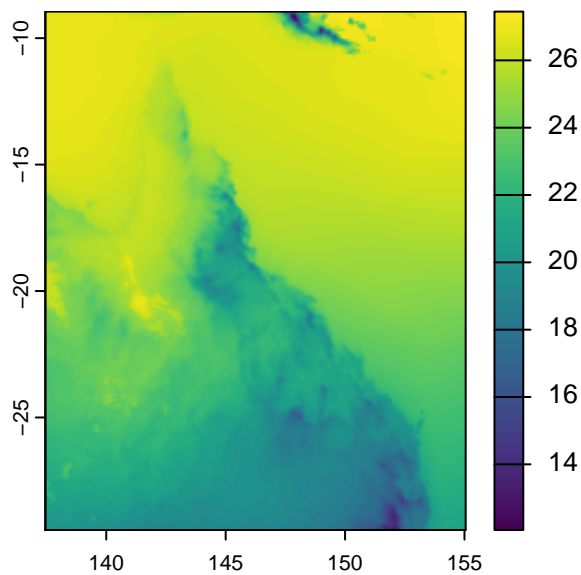
```
class      : SpatRaster
dimensions : 205, 176, 120 (nrow, ncol, nlyr)
resolution : 0.1, 0.1 (x, y)
extent     : 137.45, 155.05, -29.45, -8.95 (xmin, xmax, ymin, ymax)
coord. ref.: lon/lat WGS 84(CRS84) (OGC:CRS84)
source     : tas_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_sem_1981-2100.nc
varname    : tas_annual (Seasonal average of Near-Surface Air Temperature (Annual))
names      : 1981-01-01, 1982-01-01, 1983-01-01, 1984-01-01, 1985-01-01, 1986-01-01, ...
unit       : degC, degC, degC, degC, degC, degC, ...
time (days): 1981-01-01 to 2100-01-01 (2 steps)
```

Sub-setting and plotting the data

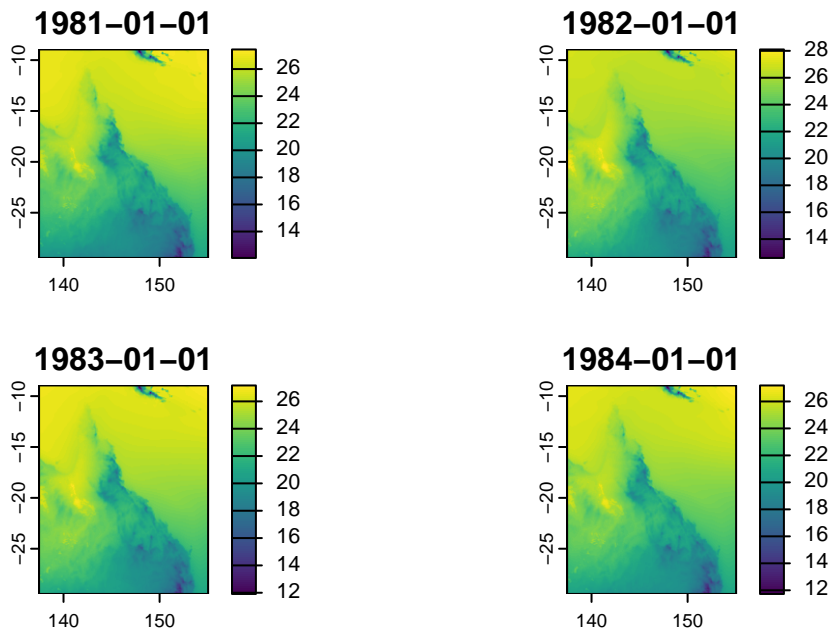
```
tas[[1]]
```

```
class      : SpatRaster
dimensions : 205, 176, 1  (nrow, ncol, nlyr)
resolution : 0.1, 0.1  (x, y)
extent     : 137.45, 155.05, -29.45, -8.95  (xmin, xmax, ymin, ymax)
coord. ref.: lon/lat WGS 84 (CRS84) (OGC:CRS84)
source     : tas_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_au-10i_10km_sem_1981-2100.nc
varname    : tas_annual (Seasonal average of Near-Surface Air Temperature (Annual))
name       : 1981-01-01
unit       :      degC
time (days) : 1981-01-01
```

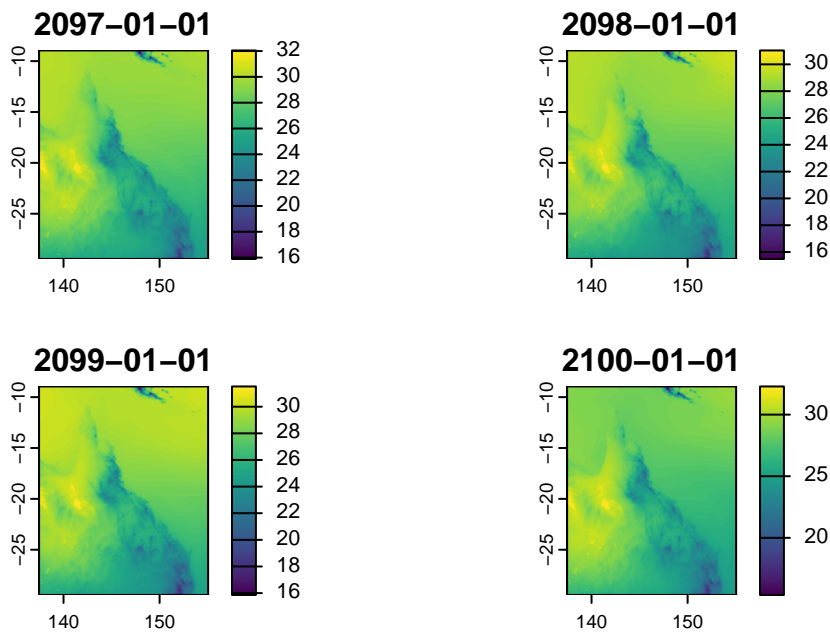
```
plot(tas[[1]])
```



```
plot(tas[[1:4]])
```

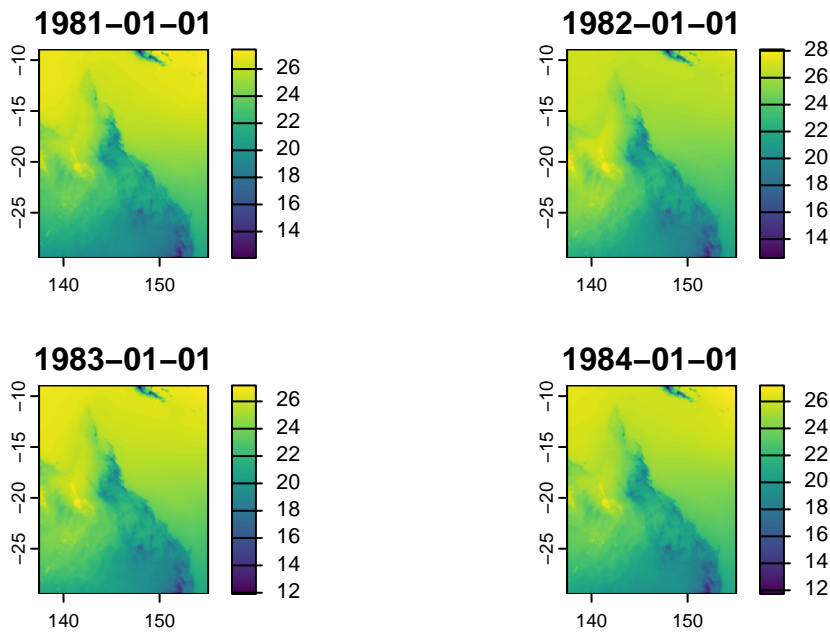


```
plot(tas[[117:120]])
```

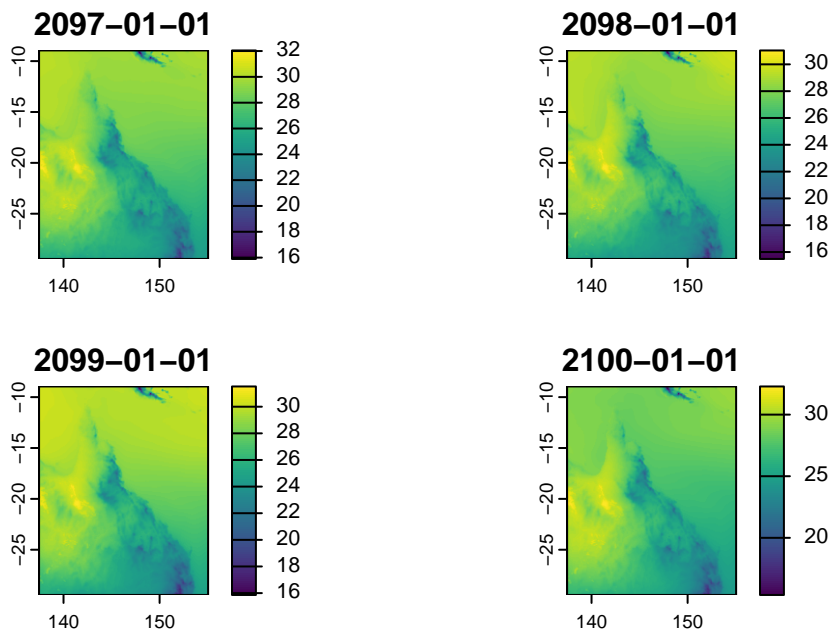


Can also sub-set the data according to the date array

```
plot(tas[[dates >= as.Date("1981-01-01") & dates <= as.Date("1984-01-01")]])
```



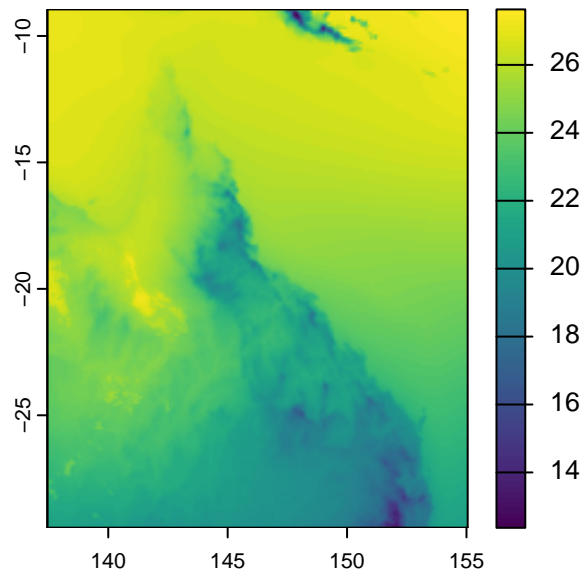
```
plot(tas[[dates >= as.Date("2097-01-01") & dates <= as.Date("2100-01-01")]])
```



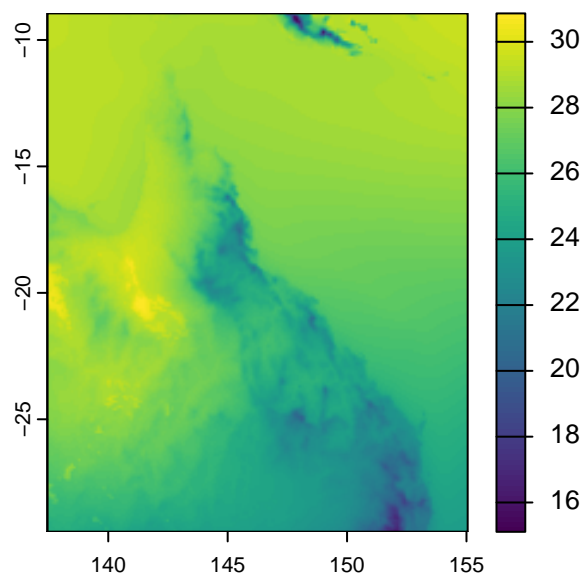
```
tas_base = mean(tas[[1:30]]) # Calculating climatology for baseline (1981-2010)
tas_fut = mean(tas[[91:120]]) # Calculating climatology for future (2071-2100)
```

Q: Can you cut the historical data and future based on the dates?

```
plot(tas_base)
```

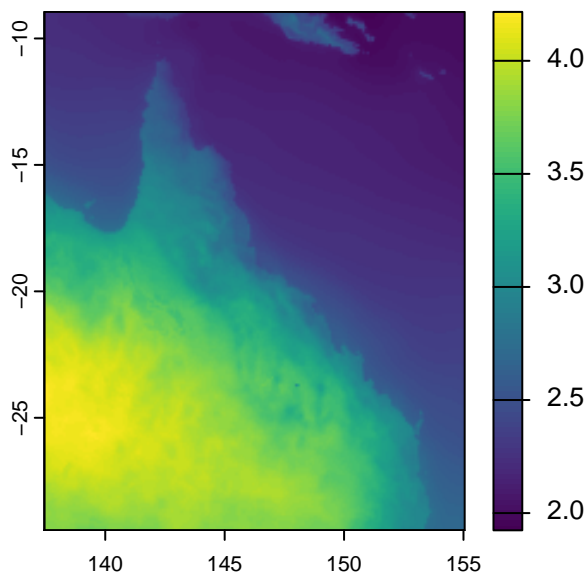


```
plot(tas_fut)
```



Change in future temperature (future - base)

```
tas_dif = tas_fut - tas_base  
plot(tas_dif)
```

Q. Can you make this plot nicer? Add a title and change the colours

Hint: You can set plot titles using 'main'

Control the colours using `col = brewer.pal(11, 'PaletteName')` (see <https://colorbrewer2.org/> for colour options)

You can plot multiple figures in one plot using `par(mfrow = c(nrows, ncols))`

Also check the instructions from Session 1!

Extracting out point data (timeseries)

```
tas[50,50] # extracts data from the 50th lat and 50th lon position
```

	1981-01-01	1982-01-01	1983-01-01	1984-01-01	1985-01-01	1986-01-01	1987-01-01
1	25.24902	25.26156	24.88015	24.41082	24.94491	24.89218	25.25115
	1988-01-01	1989-01-01	1990-01-01	1991-01-01	1992-01-01	1993-01-01	1994-01-01
1	25.23535	25.03982	25.42452	25.37658	25.3677	25.10143	25.3218
	1995-01-01	1996-01-01	1997-01-01	1998-01-01	1999-01-01	2000-01-01	2001-01-01
1	24.87841	25.11816	25.46691	25.25924	25.05966	25.53942	25.29299
	2002-01-01	2003-01-01	2004-01-01	2005-01-01	2006-01-01	2007-01-01	2008-01-01
1	25.82775	25.32467	25.54989	25.65566	25.43701	25.56466	25.35082
	2009-01-01	2010-01-01	2011-01-01	2012-01-01	2013-01-01	2014-01-01	2015-01-01
1	25.64773	25.44879	25.77899	25.52069	26.40573	26.00399	26.02764
	2016-01-01	2017-01-01	2018-01-01	2019-01-01	2020-01-01	2021-01-01	2022-01-01
1	25.50436	25.88986	26.02621	25.68951	25.9519	25.43524	26.02264
	2023-01-01	2024-01-01	2025-01-01	2026-01-01	2027-01-01	2028-01-01	2029-01-01
1	26.07147	25.98385	26.4558	25.86419	26.31649	25.50891	26.05761
	2030-01-01	2031-01-01	2032-01-01	2033-01-01	2034-01-01	2035-01-01	2036-01-01
1	26.26635	26.16805	25.56466	25.6921	26.2561	26.34539	25.9425

	2037-01-01	2038-01-01	2039-01-01	2040-01-01	2041-01-01	2042-01-01	2043-01-01
1	26.57571	26.29916	26.85446	26.48287	26.22824	26.75714	26.34863
	2044-01-01	2045-01-01	2046-01-01	2047-01-01	2048-01-01	2049-01-01	2050-01-01
1	26.01452	26.37426	26.30599	26.89645	26.49154	26.71673	26.64852
	2051-01-01	2052-01-01	2053-01-01	2054-01-01	2055-01-01	2056-01-01	2057-01-01
1	26.51864	26.47985	26.61602	26.9179	26.97494	26.73034	27.19863
	2058-01-01	2059-01-01	2060-01-01	2061-01-01	2062-01-01	2063-01-01	2064-01-01
1	27.05032	26.64703	27.30004	26.55456	27.39593	27.40051	27.23416
	2065-01-01	2066-01-01	2067-01-01	2068-01-01	2069-01-01	2070-01-01	2071-01-01
1	27.5408	27.24737	27.33636	27.09204	27.59817	26.69656	27.19824
	2072-01-01	2073-01-01	2074-01-01	2075-01-01	2076-01-01	2077-01-01	2078-01-01
1	26.80004	27.00533	28.2991	27.67822	27.7326	27.16381	28.68814
	2079-01-01	2080-01-01	2081-01-01	2082-01-01	2083-01-01	2084-01-01	2085-01-01
1	27.56314	27.8674	27.92907	27.94192	27.6759	27.62152	27.59582
	2086-01-01	2087-01-01	2088-01-01	2089-01-01	2090-01-01	2091-01-01	2092-01-01
1	28.27426	27.941	28.31573	27.85235	27.81686	27.98327	28.39962
	2093-01-01	2094-01-01	2095-01-01	2096-01-01	2097-01-01	2098-01-01	2099-01-01
1	28.12872	28.79971	28.32763	27.89684	28.33975	27.95684	28.80871
	2100-01-01						
1	28.1459						

```
cells <- cellFromRowCol(tas[[1]], 50, 50)
xyFromCell(tas, cells)
```

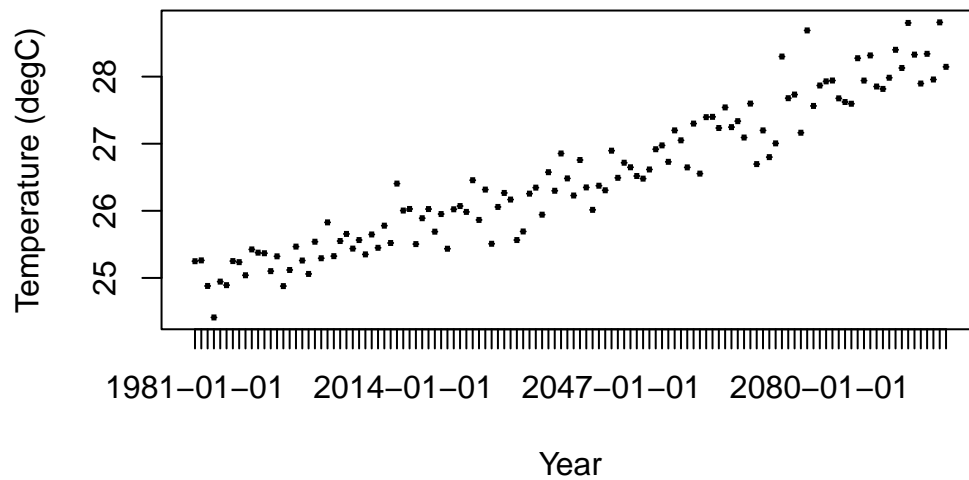
```
      x      y
[1,] 142.4 -13.9
```

```
df = melt(tas[50,50])
```

Using as id variables

Basic plot

```
plot(df, xlab = "Year", ylab = "Temperature (degC)")
```



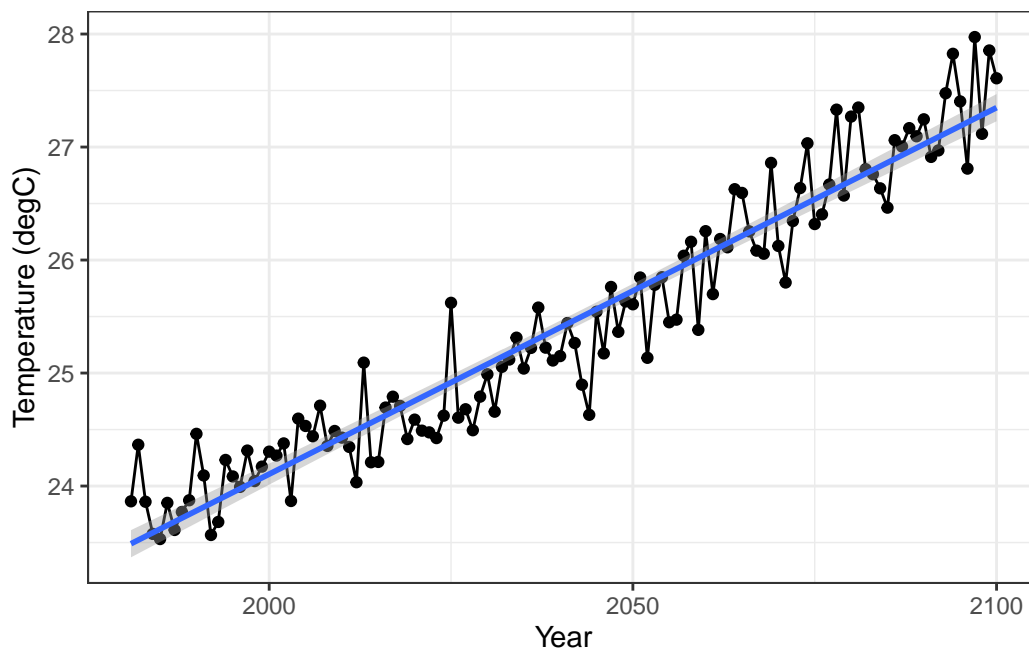
Calculating spatial average of all data

```
spat_ave = global(tas, fun=mean, na.rm=TRUE)
spat_ave$date = dates
```

ggplot

```
ggplot(data = spat_ave, aes(y=mean, x=date))+
  ylab('Temperature (degC)') + xlab('Year') +
  geom_point() +
  geom_line() +
  geom_smooth(method = "lm") +
  theme_bw()
```

`geom_smooth()` using formula = 'y ~ x'



Q. Can you add another model to this plot and compare the two?

Hint: You'll need to prepare a dataframe with data for all models in it. One of the columns will need to be the values, and the other the model name.

Part 2: Rainfall Data (multiple models)

Working with multiple Models

```
pr_files <- list.files(path = "data/annual/", pattern = "pr", full.names = TRUE)
pr_files <- list.files(path = "data/annual/", pattern = "pr.*ssp370", full.names = TRUE)
```

```
pr_data = rast(pr_files)*365 # daily mean to annual total. CCAM has a 365 day calendar.
pr_data
```

```
class      : SpatRaster
dimensions : 205, 176, 360  (nrow, ncol, nlyr)
resolution : 0.1, 0.1  (x, y)
extent     : 137.45, 155.05, -29.45, -8.95  (xmin, xmax, ymin, ymax)
coord. ref.: lon/lat WGS 84 (CRS84) (OGC:CRS84)
source(s)  : memory
varname     : pr_annual (Seasonal average of Precipitation (Annual))
names      : pr_annual_1, pr_annual_2, pr_annual_3, pr_annual_4, pr_annual_5, pr_annual_6,
min values  :      52.196,      48.72456,      85.55378,      54.58001,      150.5908,      117.2378,
```

```
max values :    6151.675,  7233.45194,  5595.35947,  6152.59279,   5482.5640,   5449.5375,  
time (days) : 1981-01-01 to 2100-01-01 (2 steps)
```

Repeating the year names multiple times to correspond with multiple models

```
years = seq(1981,2100)  
years_rep = rep(years, times =3)  
names(pr_data) = years_rep
```

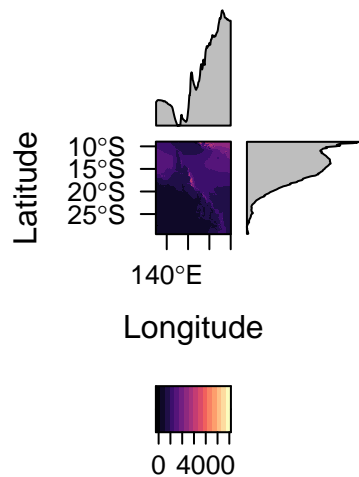
Calculating the model average

```
pr_modavg = tapp(pr_data, years, fun = mean)  
  
# Calculating climatology for baseline (1981-2100)  
pr_base = mean(pr_modavg[[1:30]]) # Converting from daily mean to annual mean  
  
# Calculating climatology for future (2071-2100)  
pr_fut = mean(pr_modavg[[91:120]]) # Converting from daily mean to annual mean
```

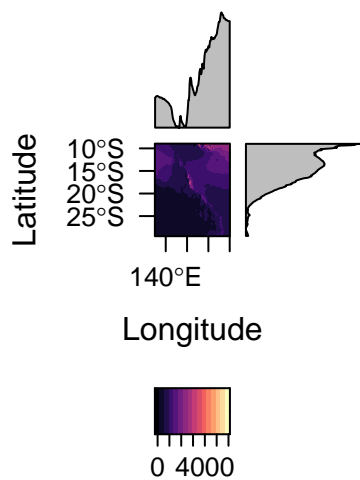
Q: Can you select the data based on the years instead?

Plot historic and future rainfall

```
levelplot(pr_base)
```

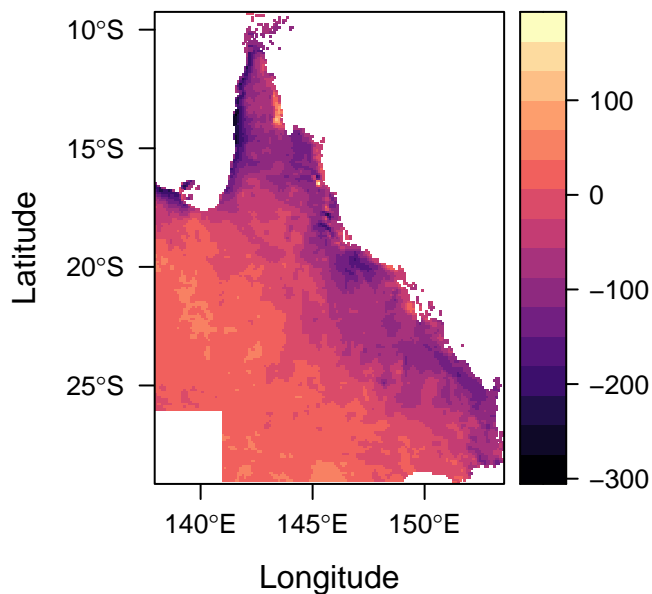


```
levelplot(pr_fut)
```



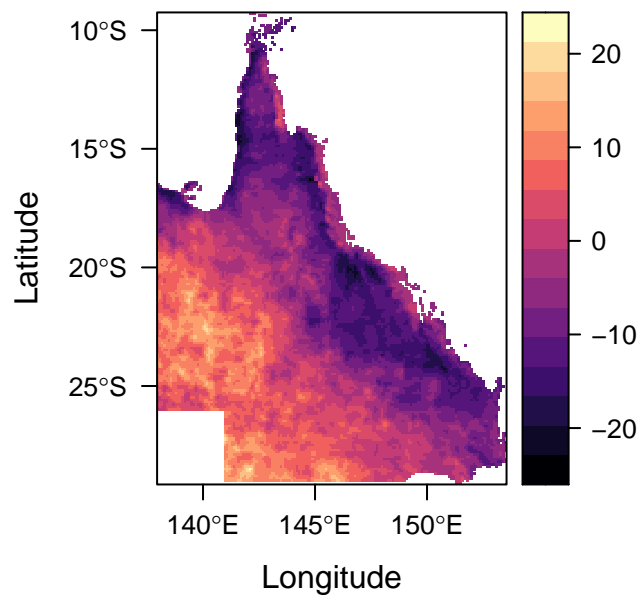
Cutting data to Queensland

```
qld_shp = vect('data/shp/QLD_State_Mask.shp')
pr_dif = pr_fut - pr_base
pr_dif_masked <- crop(pr_dif, qld_shp, mask = TRUE)
levelplot(pr_dif_masked, margin = FALSE)
```



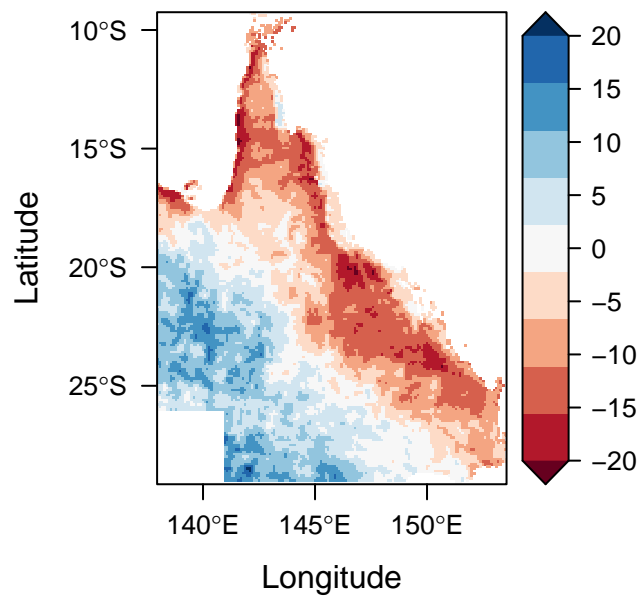
Plotting the percent change

```
pr_pdif = (pr_fut - pr_base) / pr_base * 100 #Percent difference
pr_pdif_masked <- crop(pr_pdif, qld_shp, mask = TRUE)
levelplot(pr_pdif_masked, margin = FALSE)
```



Specifying plotting bins and colours

```
my.at <- seq(-20, 20, length.out = 10)
my.at = c(-Inf, my.at, Inf)
levelplot(pr_pdif_masked, margin = FALSE, at = my.at, cuts=11, pretty=T,
          col.regions=((brewer.pal(11,"RdBu"))))
```



Q. Can you modify this plot to show more information? Can you add a title and change the colours?

Would showing multiple models on this plot help?

Q. Can we compare the results from SSP370 to another Scenario?

Part 3: Validating your data Calculating BioClim Indices

List monthly climate files

```
dir("data/monthly/")
```

```
[1] "pr_ACCESS-CM2_ssp126_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[2] "pr_ACCESS-CM2_ssp245_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[3] "pr_ACCESS-CM2_ssp370_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[4] "pr_EC-Earth3_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[5] "pr_EC-Earth3_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[6] "pr_EC-Earth3_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[7] "pr_GFDL-ESM4_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[8] "pr_GFDL-ESM4_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[9] "pr_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[10] "tasmax_ACCESS-CM2_ssp126_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[11] "tasmax_ACCESS-CM2_ssp245_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[12] "tasmax_ACCESS-CM2_ssp370_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[13] "tasmax_EC-Earth3_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[14] "tasmax_EC-Earth3_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[15] "tasmax_EC-Earth3_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[16] "tasmax_GFDL-ESM4_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[17] "tasmax_GFDL-ESM4_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[18] "tasmax_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[19] "tasmin_ACCESS-CM2_ssp126_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[20] "tasmin_ACCESS-CM2_ssp245_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[21] "tasmin_ACCESS-CM2_ssp370_r1i1p1f1_CCAM10oc_aus-10i_10km_mon_1981-2100.nc"
[22] "tasmin_EC-Earth3_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[23] "tasmin_EC-Earth3_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[24] "tasmin_EC-Earth3_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[25] "tasmin_GFDL-ESM4_ssp126_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[26] "tasmin_GFDL-ESM4_ssp245_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
[27] "tasmin_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc"
```

Load shapefile for Sunshine Coast

```
lga_shp = vect('data/shp/SunshineCoast.shp')
```

Load monthly climate data

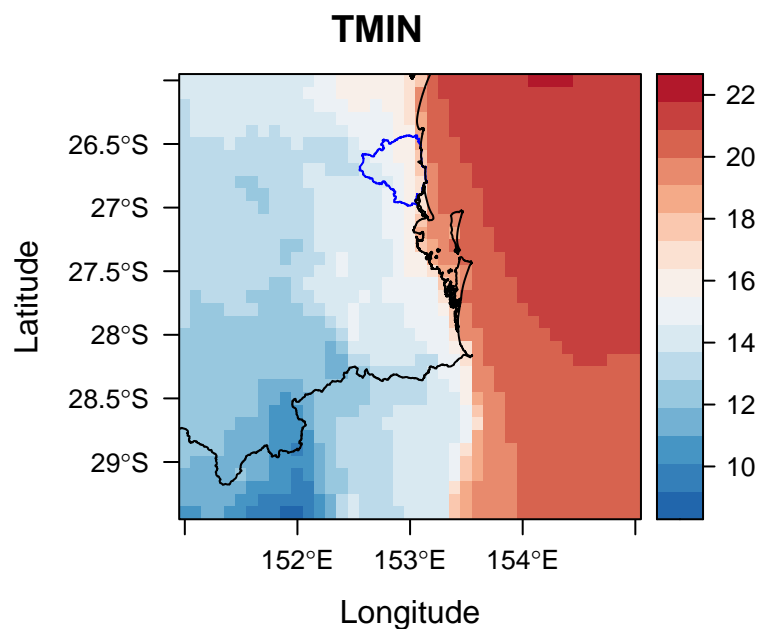

```
tmax = rast("data/monthly/tasmax_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100")
tmin = rast("data/monthly/tasmin_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100")
pr = rast("data/monthly/pr_GFDL-ESM4_ssp370_r1i1p1f1_CCAM10_aus-10i_10km_mon_1981-2100.nc")
```

Assign time labels

```
dates <- seq(as.Date("1981-01-01"), as.Date("2100-12-01"), by = "month")
names(tmax) = dates
names(tmin) = dates
names(pr) = dates
```

Plot temperature minimum

```
temp_stack <- mean(tmin[[1:360]])
lga_shp2 <- as(lga_shp, "Spatial")
qld2 <- as(qld_shp, "Spatial")
levelplot(temp_stack, margin = FALSE, par.settings = BuRdTheme, main = 'TMIN') +
  latticeExtra::layer(sp.polygons(lga_shp2, col = 'blue')) +
  latticeExtra::layer(sp.polygons(qld2))
```



Load observational data

```
dir("data/obs/")
```

```
[1] "agcd_v1_precip_r005_daily_1981_2020.nc"
[2] "agcd_v1_tmax_r005_daily_1981_2020.nc"
[3] "agcd_v1_tmin_r005_daily_1981_2020.nc"
```

```
obs_tmax = rast(list.files(path = "data/obs/", pattern = "tmax", full.names = TRUE))
obs_tmin = rast(list.files(path = "data/obs/", pattern = "tmin", full.names = TRUE))
obs_pr = rast(list.files(path = "data/obs/", pattern = "precip", full.names = TRUE))
```

Resample observational data

```
obs_tmax_regridded <- resample(obs_tmax, tmax, method = "bilinear")
obs_tmin_regridded <- resample(obs_tmin, tmin, method = "bilinear")
obs_pr_regridded <- resample(obs_pr, pr, method = "bilinear")
```

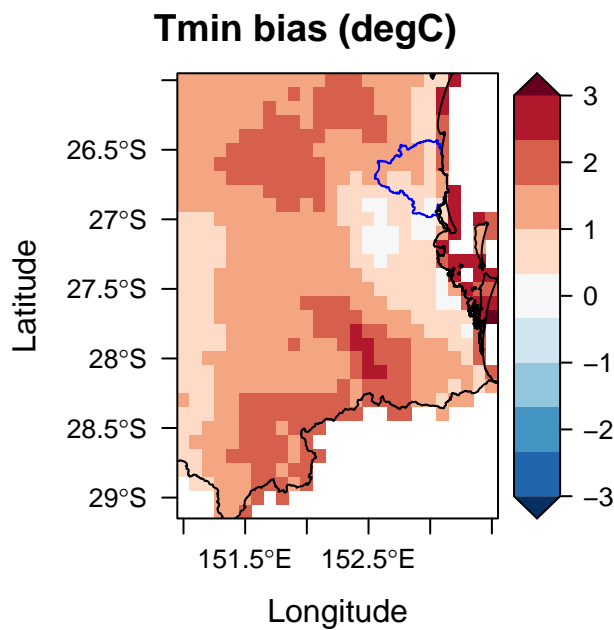
Extract historical slices

```
tmax_his = tmax[[dates >= as.Date("1981-01-01") & dates < as.Date("2021-01-01")]]
tmin_his = tmin[[dates >= as.Date("1981-01-01") & dates < as.Date("2021-01-01")]]
pr_his = pr[[dates >= as.Date("1981-01-01") & dates < as.Date("2021-01-01")]]
```

Evaluate tmin bias

```
tmin_bias = mean(tmin_his) - mean(obs_tmin_regridded)
tmin_bias_masked <- crop(tmin_bias, qld_shp, mask = TRUE)

my.at <- c(-Inf, seq(-3, 3, length.out = 10), Inf)
levelplot(tmin_bias_masked, at = my.at, margin = FALSE, cuts = 11, pretty = TRUE,
          main = 'Tmin bias (degC)',
          col.regions = rev(brewer.pal(11, "RdBu"))) +
  latticeExtra::layer(sp.polygons(lga_shp2, col = 'blue')) +
  latticeExtra::layer(sp.polygons(qld2))
```



Bias metrics

```
rmse = global((mean(tmin_his) - mean(obs_tmin_regridded))^2, fun = "mean", na.rm = TRUE)[1]
mape = global(abs((mean(tmin_his) - mean(obs_tmin_regridded)) / mean(obs_tmin_regridded)) *
print(paste("RMSE:", rmse))
```

```
[1] "RMSE: 3.48752996185524"
```

```
print(paste("MAPE:", mape))
```

```
[1] "MAPE: 11.8785189666455"
```

Mask climate data to Sunshine Coast

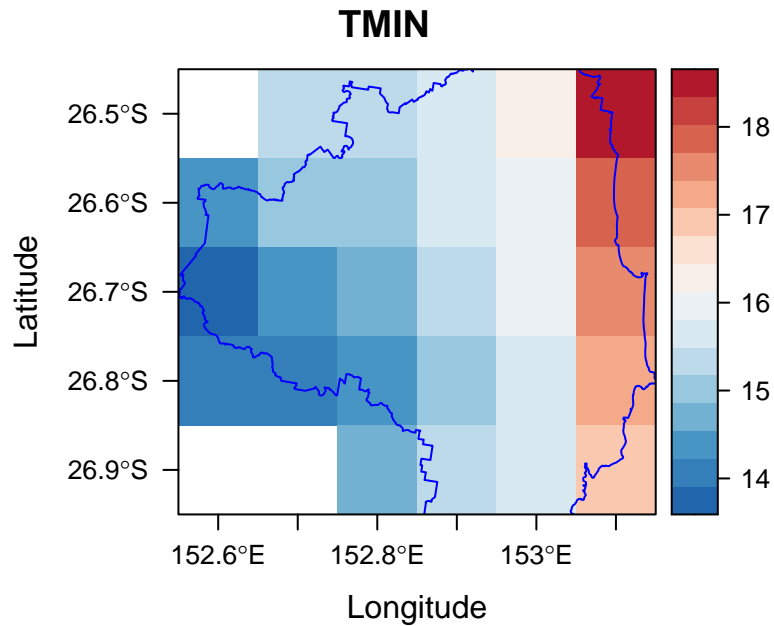
```
pr_masked <- crop(pr, lga_shp, mask = TRUE)
tmin_masked <- crop(tmin, lga_shp, mask = TRUE)
tmax_masked <- crop(tmax, lga_shp, mask = TRUE)
```

Spatial averages

```
pr_ave_coarse = global(pr_masked, fun = mean, na.rm = TRUE)
tmin_ave_coarse = global(tmin_masked, fun = mean, na.rm = TRUE)
tmax_ave_coarse = global(tmax_masked, fun = mean, na.rm = TRUE)
```

Plot masked tmin

```
tmin_masked_mean <- mean(tmin_masked[[1:360]])
levelplot(tmin_masked_mean, margin = FALSE, par.settings = BuRdTheme, main = 'TMIN') +
  latticeExtra::layer(sp.polygons(lga_shp2, col = 'blue'))
```

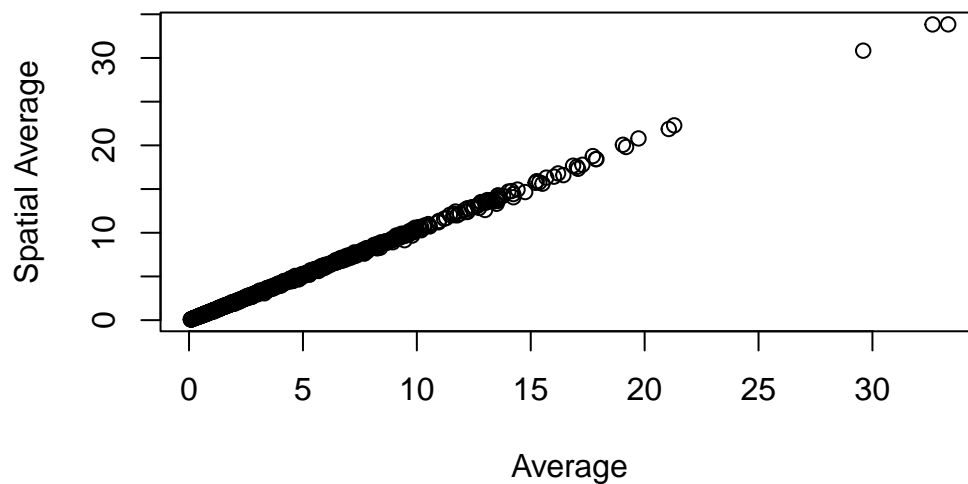


Weighted average

```
pr_ave = as.data.frame(t(terra::extract(pr, lga_shp, weights = TRUE, fun = mean, na.rm = TRUE)))
tmin_ave = as.data.frame(t(terra::extract(tmin, lga_shp, weights = TRUE, fun = mean, na.rm = TRUE)))
tmax_ave = as.data.frame(t(terra::extract(tmax, lga_shp, weights = TRUE, fun = mean, na.rm = TRUE)))
```

Compare averages

```
par(mfrow = c(1, 1))
plot(pr_ave_coarse$mean, pr_ave$V1, xlab = "Average", ylab = "Spatial Average")
```



Prepare dataframe for biovars

```
colnames(pr_ave)[1] <- "pr"
colnames(tmin_ave)[1] <- "tmin"
colnames(tmax_ave)[1] <- "tmax"
pr_ave$date <- rownames(pr_ave)
tmin_ave$date <- rownames(tmin_ave)
tmax_ave$date <- rownames(tmax_ave)
df <- merge(pr_ave, tmin_ave, by = "date", all = TRUE)
df <- merge(df, tmax_ave, by = "date", all = TRUE)
```

Subset to baseline and future

```
df$date <- as.Date(df$date)
df$year <- as.numeric(format(df$date, "%Y"))
df_base = subset(df, year >= 1981 & year <= 2010)
df_fut = subset(df, year >= 2071 & year <= 2100)
```

Calculate biovars

```
bio_base = biovars(df_base$pr, df_base$tmin, df_base$tmax)
bio_fut = biovars(df_fut$pr, df_fut$tmin, df_fut$tmax)
print(bio_base[, c("bio5", "bio6", "bio12", "bio15")])
```

bio5	bio6	bio12	bio15
31.247510	8.574035	1343.710569	88.641010

```
print(bio_fut[, c("bio5", "bio6", "bio12", "bio15")])
```

bio5	bio6	bio12	bio15
36.10810	11.37925	1266.43503	74.49997

Full raster bioclim calculation

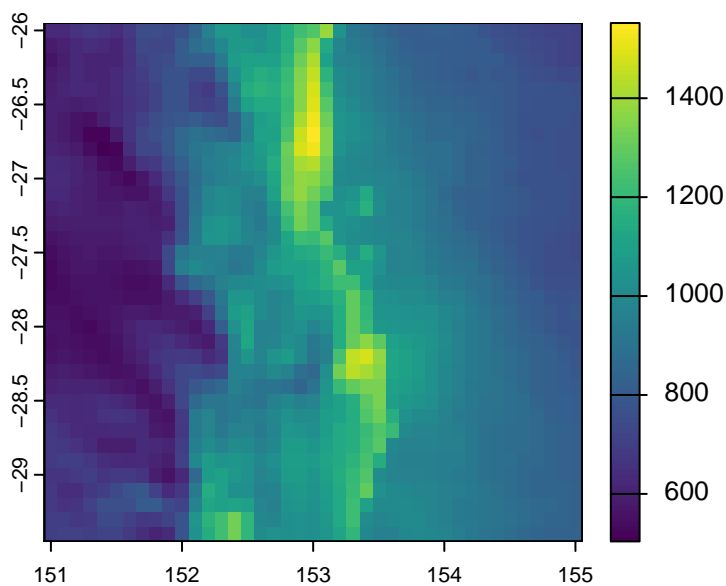
```
tmax_base = tmax[[dates >= as.Date("1981-01-01") & dates < as.Date("2011-01-01")]]
tmin_base = tmin[[dates >= as.Date("1981-01-01") & dates < as.Date("2011-01-01")]]
pr_base = pr[[dates >= as.Date("1981-01-01") & dates < as.Date("2011-01-01")]]
tmax_fut = tmax[[dates >= as.Date("2071-01-01") & dates < as.Date("2101-01-01")]]
tmin_fut = tmin[[dates >= as.Date("2071-01-01") & dates < as.Date("2101-01-01")]]
pr_fut = pr[[dates >= as.Date("2071-01-01") & dates < as.Date("2101-01-01")]]

bioclim_input_base = c(pr_base, tmin_base, tmax_base)
bioclim_input_fut = c(pr_fut, tmin_fut, tmax_fut)

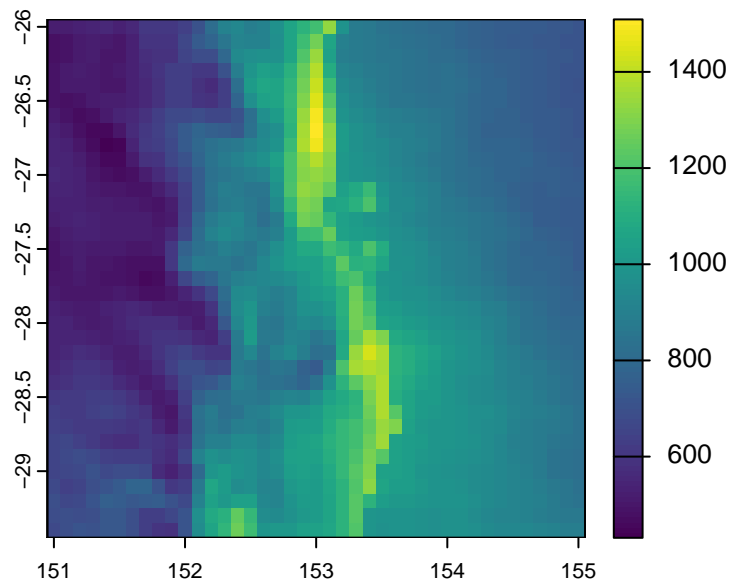
fun_bio_calc <- function(x) {
  n <- length(x) / 3
  pr <- x[1:n]
  tmin <- x[(n + 1):(2 * n)]
  tmax <- x[(2 * n + 1):(3 * n)]
  bio <- biovars(pr, tmin, tmax)
  return(bio)
}

bio_base <- app(bioclim_input_base, fun = fun_bio_calc)
bio_fut <- app(bioclim_input_fut, fun = fun_bio_calc)

plot(bio_base[[12]])
```



```
plot(bio_fut[[12]])
```



Discussion Questions

Q: Can you plot the bioclimatic indices in the past and present and compare the changes?

Q: What does the bioclimatic indicators look like for a lower emissions scenario?