Introduction to handling raster time series in R

1 Introduction

This tutorial will showcase how to work with raster data efficiently. The analysis will be shown in Julia, Python and R to showcase the similarities and differences in handling raster data in these ecosystems.

In this tutorial we are going to use the COSMO REA reanalyis near surface air temperature data. The data is an reanalysis dataset on a 6km by 6km grid. We are going to use the monthly average values, but the data is also avialable with an hourly or daily temporal resolution. The data was produced in the GRIB format but was converted to NetCDF files in the NFDI4Earth Pilot.

2 Time series analysis

2.1 Loading of necessary packages

First we load the relevant packages in the different languages for working with raster and vector data and also the packages for plotting.

```
library("rnaturalearth")
library("rnaturalearthdata")
library("sf")
library("raster")
library("stars")
library("lubridate")
```

Now we download the airtemperature data for 1995 to the data/ folder. In the first part of the tutorial we are going to only use the data from one single year, and later on we are going to combine the datasets from different years together. The data will only be downloaded if it is not yet available on the local computer.

```
setwd("data/")
destfile <- getwd()

# URL of DKRZ folder containing COSMO-REA Dataset files
baseurl <- "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-1
urlList <- "0"</pre>
```

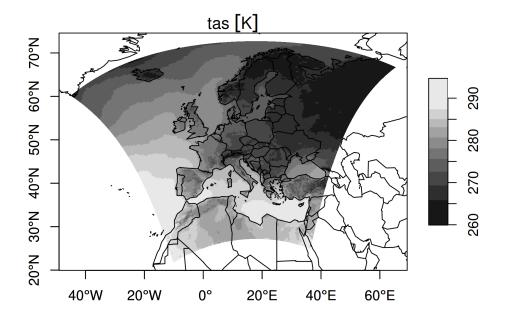
```
12 <- format(seq(as.Date("1995-12-01"), as.Date("2018-12-01"), by="year"), format = "%Y%m")
for (i in 1:length(l1)) {
  x <- paste0(baseurl,l1[i],"-",l2[i],".nc")</pre>
# Use the following line to download the data to the destfile folder
#download.file(x, paste0(destfile,"\\", basename(x)), mode = "wb") #download files
 urlList[i] <- x
  i = i + 1
}
urlList
 [1] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [2] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [3] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [4] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [5] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [6] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [7] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [8] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
 [9] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[10] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[11] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[12] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[13] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[14] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[15] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[16] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[17] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[18] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[19] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[20] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[21] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[22] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[23] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
[24] "http://esgf1.dkrz.de/thredds/fileServer/cosmo-rea/reanalysis/EUR-6km/DWD/ECMWF-ERAINT,
```

11 <- format(seq(as.Date("1995-01-01"), as.Date("2018-01-01"), by="year"), format = "%Y%m")

2.2 Opening the data and first map

Now we are going to open the raster data as a datacube and plot a first overview map.

```
file = paste0(destfile, "/", basename(urlList[1]))
m = read_mdim(file, "tas", curvilinear = c("longitude", "latitude"))
plot(m[,,,1], downsample = 5, axes = TRUE, reset = FALSE) # downsample to save time
maps::map(add = TRUE)
```

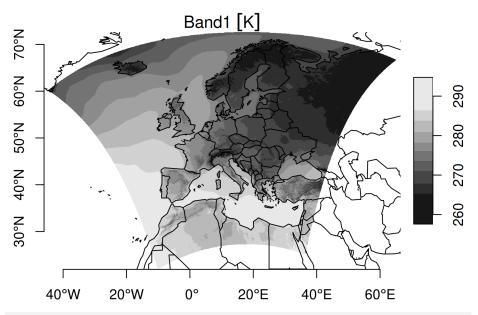


2.3 Coordinate reference system of the data

The data is in a rotated latitude longitude grid. This rotation helps to reduce the spatial distortions of the data because of the projection. For an introduction into the concepts of Here we construct a projection string from the metadata of the dataset so that we can use this projection information latter on for converting between different coordinate reference systems.

In R we are warping the raster data from the rotated longitude latitude grid to an unrotated longitude latitude grid.

```
if (isFALSE(file.exists("out2.nc"))) {
    gdal_utils("warp", file, "out2.nc")
}
m2 = read_mdim("out2.nc")  # now has time over bands, caused
plot(m2, axes = TRUE, reset = FALSE)  # fast: regular grid
maps::map(add = TRUE)
```



```
# Add time dimension and crs to re-gridded dataset
t = st_get_dimension_values(m, "time")
merge(m2, name = "time") |>
   st_set_dimensions("time", values = t) |>
   setNames("tas") -> m3
m3 <- st_set_crs(m3, st_crs(4326))
m3</pre>
```

stars object with 3 dimensions and 1 attribute attribute(s), summary of first 1e+05 cells:

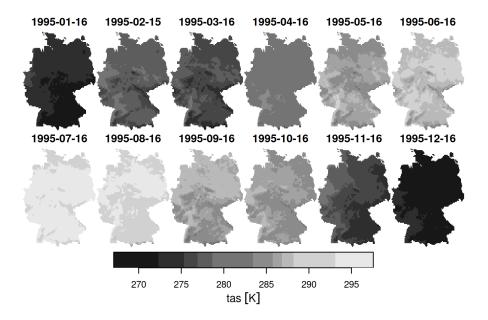
Min. 1st Qu. Median Mean 3rd Qu. Max. NA's tas [K] 280.5498 285.1855 286.6836 286.7361 287.6787 294.75 89615 dimension(s):

```
from
            to offset
                         delta refsys
                                                          values x/y
lon
        1 1469 -44.75 0.07483 WGS 84
                                                            NULL [x]
lat
        1
           678
                 21.9 0.07483 WGS 84
                                                            NULL [y]
time
            12
                                 Date 1995-01-16,...,1995-12-16
        1
```

2.4 Restricting to an area of interest

Now we load the polygon data for the border of Germany to restrict our data to the bounding box of Germany.

```
# Subset for Germany
germany <- ne_countries(type = "countries", country = "Germany", scale = "medium", returncle
rger <- m3[germany]
plot(rger)</pre>
```



2.5 Split the time series into two seasons

Now we split the time series data into two datasets by season.

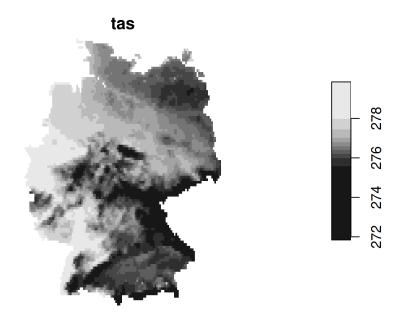
Hereby we define the summer as the time between the spring and autumn equinox. Since we are using monthly data in this example, we define summer as April to September. We can define the winter as every month that is not included in the summer dataset.

```
# Split dataset into two by season
# Definition: Summer (April - September) & Winter (October - March)
a = seq(from = 4, to = 288, by = 12)
b = seq(from = 9, to = 288, by = 12)
winter <- rger[,,,4:9]

c = seq(from = 1, to = 288, by = 12)
d = seq(from = 3, to = 288, by = 12)
e = seq(from = 10, to = 288, by = 12)
f = seq(from = 12, to = 288, by = 12)
summer <- c(rger[,,,1:3], rger[,,,10:12])</pre>
```

Now we compute the standard deviation and the mean of the time series for the "summer" and the "winter" dataset.

```
summer_mean_tas <- aggregate(summer, by = "year", FUN = mean)
plot(summer_mean_tas)</pre>
```



2.6 Summary

In this tutorial we learned how to work with raster data in R. We explored the COSMO-REA dataset and computed temporal statistics for different seasons. For the computations we selected a smaller area of interest.