## **Topic 2: Using geographic maps in R**

This topic shows how to use geographic maps in R with sp and sf objects. In particular, we consider

- Basic geographic maps in R.
- Library maps.
- GADM database of Global Administrative Areas.
- Plotting sp objects using the world map.
- Plotting sf objects using the world map.

## Geographic maps in R

The maps library for R is a basic tool for creating maps of countries and regions of the world. For example, you can create a map of the USA and its states as:

```
> library(maps)
> map("state", interior = FALSE)
> map("state", boundary = FALSE, col = "gray", add = TRUE)
```



The coordinate system of the plot is latitude and longitude, so it's easy to overlay other spatial data on this map.

Unfortunately, the data for the MAPS library isn't sufficient for some applications. Also in many R packages maps are fairly low-resolution, and political boundaries can be incomplete or out-of-date.

To produce high-quality maps, it is better to use free online resources where you can find up-to-date high-resolution map data for use with R.

### **GADM** database of Global Administrative Areas

GADM is a spatial database of the location of the world's administrative areas (or administrative boundaries) for use in GIS and similar software.

Administrative areas in this database are countries and lower-level subdivisions, such as provinces, departments, counties, etc. GADM describes where

these administrative areas are (the "spatial features"), and it provides some attributes for each area.

GADM goal is to map all administrative areas of all countries, at all levels, at any time period. They use a high spatial resolution and an extensive set of attributes for each spatial feature. The current version of GADM maps has 400,276 administrative areas.

GADM maps are available as native R objects that can be plotted directly with the  ${\tt SPPLOT}$  function (from the  ${\tt SP}$  package).

For example, here's how to load the data for Australia, and then plot each state with a random colour:

```
> library(sp)
> con<-gzcon(url(
+ "http://biogeo.ucdavis.edu/data/gadm2.8/rds/AUS_adm1.rds"))
> data <- readRDS(con)</pre>
```

> library(RColorBrewer)
> col <- rainbow(length(levels(data\$NAME\_1)))
> data\$NAME\_1 = as.factor(data\$NAME\_1)
> spplot(data, "NAME\_1", col.regions=col,
+ main="Australia Regions", colorkey = FALSE, lwd=.4)
> col <- rainbow(length(levels(data\$NAME\_1)))
> spplot(data, "NAME\_1", col.regions=col,
+ main="Australia Regions", colorkey = FALSE, lwd=.4)

#### Australia Regions



To use the Australia GADM maps in sf, first, download the data form the webpage

https://geodata.ucdavis.edu/gadm/gadm4.1/gpkg/gadm41\_AUS.gpkg or the LMS folder Data.

Then use the following commands to plot states (their information is in the layer ADM\_ADM\_1 and names are in NAME\_1) in different colours. The parameters key.pos and key.width adjust the legend.

```
> data <- st_read("gadm41_AUS.gpkg", layer="ADM_ADM_1")
Reading layer 'ADM_ADM_1' from data source</pre>
```

using driver 'GPKG'

Simple feature collection with 11 features and 11 fields

Geometry type: MULTIPOLYGON

Dimension: XY

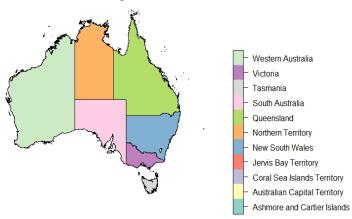
Bounding box: xmin: 112.9211 ymin: -55.11694

xmax: 159.1092 ymax: -9.142176

Geodetic CRS: WGS 84

```
> str(data)
Classes sf and 'data.frame': 11 obs. of 12 variables:
$ GID_1 : chr "AUS.1_1" "AUS.2_1" "AUS.3_1" ...
$ GID_0 : chr "AUS" "AUS" "AUS" "AUS" ...
$ COUNTRY : chr "Australia" "Australia" "Australia" ...
$ NAME_1 : chr "Ashmore and Cartier Islands" ...
$ VARNAME_1: chr "NA" "NA" "NA" "NA" ...
$ NL_NAME_1: chr "NA" "NA" "NA" "NA" ...
$ TYPE_1 : chr "Territory" "Territory" "Territory" ...
$ ENGTYPE_1: chr "Territory" "Territory" "Territory" ...
$ CC_1 : chr
                 "12" "8" "11" "10" ...
$ HASC_1 : chr "AU.AS" "AU.AC" "AU.CR" "AU.JB" ...
$ ISO_1 : chr "NA" "AU-ACT" "NA" "NA" ...
$ geom :sfc_MULTIPOLYGON of length 11;
> plot(data["NAME_1"], key.pos = 4, key.width = lcm(7),
+ main = "Australia Regions")
```





# Plotting sp/sf objects using world map

**Example 1.** Let us plot cities with the largest population on the world map. We will use the dataset worldcities.csv from the World Cities Database (https://simplemaps.com/data/world-cities) with information about large cities across the world. It is also available in the LMS folder Data.

```
> cities <- read.csv("worldcities.csv", header = TRUE)</pre>
> str(cities)
'data.frame': 12893 obs. of 11 variables:
$ city : chr "Malishev" "Prizren" "Zubin Potok" ...
$ city_ascii: chr "Malisheve" "Prizren" "Zubin Potok" ...
$ lat
      : num 42.5 42.2 42.9 42.6 42.3 ...
      : num 20.7 20.7 20.7 21.6 21.4 ...
$ lng
$ country : chr "Kosovo" "Kosovo" "Kosovo" "Kosovo" ...
$ iso2 : chr
                  "XK" "XK" "XK" "XK" ...
$ admin_name: chr
                  "Malishev" "Prizren" "Zubin Potok"
$ capital : chr
                 "admin" "admin" "admin" "admin" ...
$ population: num
                  NA NA NA NA NA NA NA NA NA ...
$ id
           : int
                  1901597212 1901360309 1901608808
```

We consider only the cities with a population of more than 10000000:

```
> cities <-cities[complete.cases(cities), ]
> sum(cities$population > 10000000)
[1] 19
> mcities <- cities[cities$population > 10000000, ]
> mcities <- mcities[, c("lng","lat", "population")]</pre>
```

```
> mcities <- mcities [, c("lng", "lat", "population")]
 > mcities
                    lat population
            lng
       -99.1310 19.4424
                          19028000
1071
 1683
       120.9822 14.6042 11100000
 1786
      66.9900 24.8700
                         12130000
 2216
     37.6155 55.7522
                         10452000
 3390 29.0100 41.1050
                         10061000
 4218
       -58.3975 -34.6025
                          12795000
 4530
      90.4086 23.7231
                          12797394
 5045
       -43.2250 -22.9250
                         11748000
 5097
       -46.6250 -23.5587
                          18845000
 5736
       121.4365 31.2165
                         14987000
5960 116.3883 39.9289
                          11106000
 6538
      31.2500 30.0500
                          11893000
 7441
     72.8570 19.0170
                          18978000
7468 88.3247 22.4950
                          14787000
7512 77.2300 28.6700
                          15926000
7812 139.7514 35.6850
                          35676000
7849 135.4601 34.7500
                          11294000
 9739
       -73.9249 40.6943
                          19164071
10410 -118.4068 34.1140
                          12740381
```

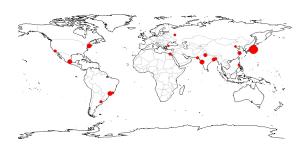
We rescale the population in 10000000 and create SpatialPointsDataFrame using cities' locations and rescaled population:

```
> mcities$population <- mcities$population/10000000
> mcities_coor <- cbind(mcities$lng, mcities$lat)
> row.names(mcities_coor) <- 1:nrow(mcities_coor)
> row.names(mcities) <- 1:nrow(mcities)
> str(mcities_coor)
> llCRS <- CRS("+proj=longlat +ellps=WGS84")
> mcities_sp <- SpatialPoints(mcities_coor, proj4string = llCRS)
> summary(mcities_sp)
```

```
> mcities_spdf <- SpatialPointsDataFrame(mcities_coor,
+ mcities, proj4string = 11CRS, match.ID = TRUE)
> summary(mcities_spdf)
Object of class SpatialPointsDataFrame
Coordinates:
min
        max
coords.x1 -118.4068 139.7514
coords.x2 -34.6025 55.7522
Is projected: FALSE
proj4string : [+proj=longlat +ellps=WGS84]
Number of points: 19
Data attributes:
lng
                 lat
                               population
Min. :-118.41 Min. :-34.60 Min. :1.006
```

Finally, we add the cities to the world map using their populations as the graphical parameter cex, that controls the sizes of the circled on the map.

```
> library(maps)
> map("world", interior = FALSE)
> map("world", boundary = FALSE, col = "gray", add = TRUE)
> plot(mcities_spdf, pch = 16, col = "red",
+ cex = (mcities_spdf$population), add = TRUE)
```



### Example 2.

In this example we visualise the cities by using the sf package.

After transforming the data to the sf format it is straightforward to plot them, by using the MAPVIEW command:

```
> mcities_sf <- st_as_sf(mcities_spdf)</pre>
> str(mcities_sf)
Classes sf and 'data.frame': 19 obs. of 4 variables:
$ lng : num -99.1 121 67 37.6 29 ...
$ lat : num 19.4 14.6 24.9 55.8 41.1 ...
$ population: num 1.9 1.11 1.21 1.05 1.01 ...
$ geometry :sfc_POINT of length 19; first list element:
'XY' num -99.1 19.4
- attr(*, "sf_column")= chr "geometry"
- attr(*, "agr")= Factor w/ 3 levels "constant", "aggregate",..:
NA NA NA
..- attr(*, "names")= chr [1:3] "lng" "lat" "population"
> mapview(mcities_sf, col.regions = "red", cex = "population",
+ fgb = FALSE)
```



Now, let us draw straight lines connecting the first 6 cities. First, we create a linestring object from the cities' coordinates. Then, we add the coordinate reference system to it and plot the result:

```
> lnstr_sfg2 <- st_linestring(as.matrix(mcities_sf[1:6,
+ c("lng", "lat"), drop = TRUE]))
> lnstr_sfc <- st_sfc(lnstr_sfg2, crs = 4326)
> mapview(lnstr_sfc, alpha.regions = 0, color = "red",
+ lwd = 2, fgb = FALSE)
```



## **Key R commands**

st_multipoint()	creates the multipoint simple feature
st_sfc()	creates simple feature geometry list columns
st_sf()	creates sf objects
st_polygon()	creates the polygon simple feature
st_as_sf()	converts external objects to an sf object
st_crs()	sets or retrieves coordinate reference system
as()	converts sf to a Spatial* object
mapview()	produces an interactive map with a spatial object
map()	draws geographic maps
st_read()	reads simple features or layers from a file
rainbow()	create a vector of n colours
row.names()	gets and sets row names
st_linestring()	creates the linestring simple feature