Introduction to the analysis of spatial data using R

Ana I. Moreno-Monroy

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Chapter 5: The power of rasters

- Raster data: the basics
- Crop, mask and plot
- Extracting raster data by polygons: an example for Poland

Raster data: the basics

- Raster data is a type of spatial data organised in grid (matrix) format of equally-spaced grid-cells (e.g. 1-square km)
- Each grid-cell of a raster contains information on a continuous variable (elevation, tempetarure, population density) for each grid-cell
- Raster can also contain categorical data (resulting from processed continuous data, e.g. land use)
- The extent of rasters can be manipulated to match a spatial object in order to extract information from grid-cells
- Raster data can be transformed into spatial data (points or spatial polygons), and spatial data can be converted into a raster (i.e. can be "rasterized")

Raster data features

- Raster data can be quite heavy if it has global extent and high-level of spatial resolution (e.g. 100-square meters)
- Shapefile folders is often available in GeoTIFF format (with .tif or .tiff extension)
- Examples of raster data available for free online are the friction surface (372MB) and accessibility to cities (520MB)
- The main package to work with rasters in R is the raster package

Importing a raster into R

• We can use the raster function of the raster package to import a raster

```
travel_time <- raster("Q:/accessibility_to_cities_2015_v1.tif")</pre>
```

• We can verify that the properties of the imported "Rasterlayer" object share some similarities with other spatial objects (CRS definition) but do not share the same structure

```
str(travel_time)
```

```
## Formal class 'RasterLayer' [package "raster"] with 12 slots
## ..@ file :Formal class '.RasterFile' [package "raster"] with 13 slots
## .....@ name : chr "Q:\\accessibility_to_cities_2015_v1.tif"
## .....@ datanotation: chr "INT4S"
## .....@ byteorder : chr "little"
## .....@ nodatavalue : num -Inf
## .....@ NAchanged : logi FALSE
```

```
##
     .. .. ..@ nbands
                            : int 1
                           : chr "BIL"
##
     .. .. ..@ bandorder
     .. .. ..@ offset
##
                           : int 0
##
     .. .. ..@ toptobottom : logi TRUE
##
     .. .. ..@ blockrows
                           : int 8
     .. .. ..@ blockcols
                          : int 43200
##
                           : chr "gdal"
##
     .. .. ..@ driver
     .. .. ..@ open
##
                            : logi FALSE
                 :Formal class '.SingleLayerData' [package "raster"] with 13 slots
##
     ..@ data
##
     .. .. .. @ values
                         : logi(0)
##
     .. .. ..@ offset
                          : num 0
     .. .. ..@ gain
                          : num 1
##
     .. .. .. @ inmemory : logi FALSE
##
     .. .. .. @ fromdisk : logi TRUE
##
##
     .. .. ..@ isfactor : logi FALSE
##
     .. .. ..@ attributes: list()
     .. .. .. @ haveminmax: logi TRUE
##
##
     .. .. ..@ min
                       : num -2.15e+09
##
                         : num 2.15e+09
     .. .. ..@ max
##
     .. .. ..@ band
                         : int 1
                         : chr ""
##
     .. .. ..@ unit
                        : chr "accessibility_to_cities_2015_v1"
##
     .. .. .. @ names
     ..@ legend :Formal class '.RasterLegend' [package "raster"] with 5 slots
##
                         : chr(0)
##
     .. .. ..@ type
##
     .. .. ..@ values
                          : logi(0)
##
     .. .. ..@ color
                          : logi(0)
##
     .. .. ..@ names
                          : logi(0)
##
     .. .. .. @ colortable: logi(0)
##
     ..@ title : chr(0)
     ..@ extent :Formal class 'Extent' [package "raster"] with 4 slots
##
##
     .. .. ..@ xmin: num -180
##
     .. .. ..@ xmax: num 180
##
     .. .. ..@ ymin: num -60
##
     .. .. ..@ ymax: num 85
##
     .. @ rotated : logi FALSE
##
     ..@ rotation:Formal class '.Rotation' [package "raster"] with 2 slots
##
     .. .. ..@ geotrans: num(0)
     .. .. ..@ transfun:function ()
##
##
     ..@ ncols
                 : int 43200
##
     ..@ nrows
                 : int 17400
                 :Formal class 'CRS' [package "sp"] with 1 slot
##
##
     .....@ projargs: chr "+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0"
     ..0 history : list()
##
##
     ..@ z
                 : list()
```

Basic raster operations

- The function **getValues** can be used to extract values stored in grid-cells. The function **cellStats** will compute a specified summary statistic (e.g. min, max, mean) across the raster cells
- The value of raster cells can be modified using algebraic operators (+, -, *, /), logical operators (e.g., >, <=), or functions (e.g. abs, min)
- The **raster** package provides its own plot functionality. It can be implemented by simply calling the **plot** function on the raster object

- Large raster files will take time to process. As an example, running **cellStats**(travel_time, 'mean') takes an i7, 2.60GHz laptop 153.3 seconds
- Before plotting or performing other operations on rasters, it is often handy to crop (global/large) rasters to smaller extends to increase speed

Adapting rasters to a given extent

- The function **crop** provides a fast way to obtain a rectangle where a spatial object fits entirely
- As an example, we can crop the accessibility to cities raster to the extent of the NUTS3 regions in Poland
- First we import the polygons, verify the two objects share the same CRS and reproject if necessary:

```
poland_borders <- readOGR(dsn = "Q:/TL2", layer = "Poland_NUTS2")

## OGR data source with driver: ESRI Shapefile

## Source: "Q:\TL2", layer: "Poland_NUTS2"

## with 16 features

## It has 4 fields

proj4string(poland_borders) == proj4string(travel_time)

## [1] FALSE

poland_borders <- spTransform(poland_borders, proj4string(travel_time))</pre>
```

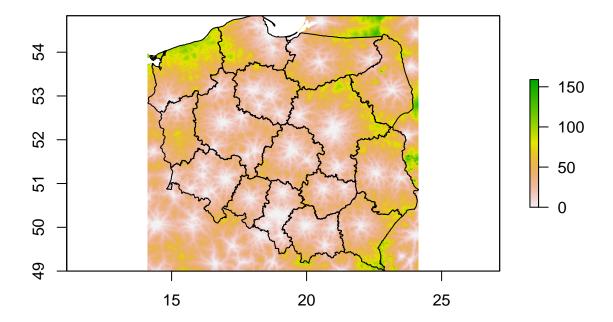
Cropping and plotting

- We then **crop** the raster to the extent of the polygons and **plot** the resulting raster and overlay the polygons to verify the result
- The scale indicates that travel times to the most proximate city within the extent of Poland vary from 0 to 150 minutes. We can get the mean value by using the **cellStats** function

```
travel_time_poland <- crop(travel_time, poland_borders)
cellStats(travel_time_poland, 'mean')

## [1] 35.7546

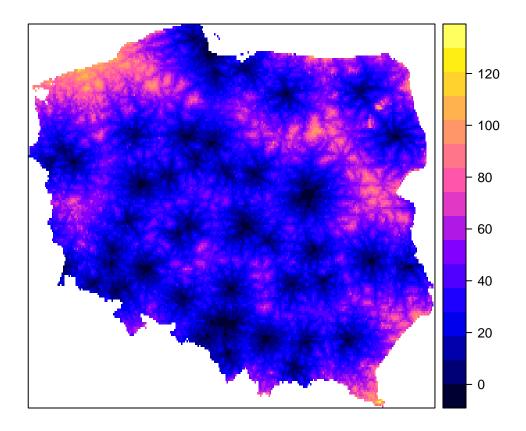
plot(travel_time_poland)
plot(poland_borders, add=TRUE)</pre>
```



Masking

- \bullet If we want the crop the raster object more closely to a spatial object, we can use the function \mathbf{mask} to set all values in the extent outside polygon borders to NA
- The **mask** function runs slower than **crop**. Running it on rasters of smaller extent (like those that had been previously cropped) can help
- As an alternative to **plot**, we can use the **spplot** function of the raster package to visualise the result

travel_time_poland_mask <- mask(travel_time_poland, poland_borders)
spplot(travel_time_poland_mask)</pre>



Rasters in interactive maps

• To get an interactive map with several options for background tiles we can use **mapview** on the masked raster (though mapview adjusts to a smaller resolution)

mapview::mapview(travel_time_poland_mask)



The html can be accessed here

Extracting raster data by polygons

- The function **extract** of the **raster** package allows perfoming operations on grid-cells that fall on a given extent, such as sum, mean, and median or any user-speficied function with the option "fun"
- The same operation is performed across each observation of the spatial object when combined with the option "byid=TRUE"
- For instance, for the case of Poland, we can extract the median travel time by NUTS3 region (leaving out NAs) by simply specifying:

```
median_travel_time <- extract(travel_time_poland, poland_borders, fun=median, na.rm=TRUE)</pre>
```

 \bullet Which returns a [1:R, 1] matrix, where R is the number of polygons (NUTS3 regions) in our spatial polygon data frame

```
str(median_travel_time)
```

```
## num [1:16, 1] 28 33 21 40 38 24 37 13 38 27 ...
```

Merging and plotting extracted data

• As the order of observations is preserved, we can add this variable directly to the SpatialPolygon-DataFrame and visualise the result using the **spplot** function from the **sp** package

```
poland_borders$median_travel_time <- median_travel_time[,1]
summary(poland_borders$median_travel_time)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 13.00 26.25 30.50 32.19 38.50 60.00

my.palette <- RColorBrewer::brewer.pal(n = 7, name = "OrRd")
spplot(poland_borders, "median_travel_time", col.regions = my.palette, main = "Minutes to the closest c</pre>
```

Minutes to the closest city (median)

