Spatial data manipulation (part 2) Lecture 7

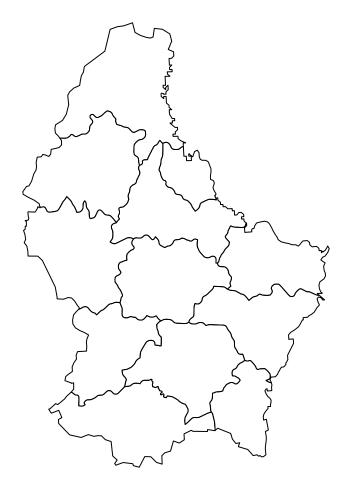
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1. Vector data manipulation

First, we will read a "SpatialPolygons"

plot(p)

```
library(raster)
f <- system.file("external/lux.shp", package="raster")</pre>
p <- shapefile(f)</pre>
p
## class
         : SpatialPolygonsDataFrame
## features
             : 5.74414, 6.528252, 49.44781, 50.18162 (xmin, xmax, ymin, ymax)
## extent
## crs
              : +proj=longlat +datum=WGS84 +no_defs
## variables : 5
## names
             : ID_1,
                          NAME_1, ID_2,
                                          NAME_2, AREA
## min values :
                        Diekirch,
                                    1, Capellen,
                   1,
                                                    76
## max values :
                   3, Luxembourg,
                                    12,
                                           Wiltz,
                                                   312
par(mai=c(0,0,0,0))
```



Basic operations are similar to the operations that you make with a data.frame.

1.1. Geomtery and attributes

We can extract the attributes (the variables in a data.frame for example) doing:

```
d <- data.frame(p)
head(d)</pre>
```

##		ID_1	NAME_1	ID_2	NAME_2	AREA
##	1	1	Diekirch	1	Clervaux	312
##	2	1	Diekirch	2	Diekirch	218
##	3	1	Diekirch	3	Redange	259
##	4	1	Diekirch	4	Vianden	76
##	5	1	Diekirch	5	Wiltz	263
##	6	2	Grevenmacher	6	Echternach	188

If we want to extract the geometry of the p object:

```
g <- geom(p)
head(g)
```

```
object part cump hole
## [1,]
            1
                 1
                          0 6.026519 50.17767
## [2,]
            1
                 1
                      1
                          0 6.031361 50.16563
## [3,]
            1
                1
                     1
                          0 6.035646 50.16410
## [4,]
            1
               1
                     1
                         0 6.042747 50.16157
## [5,]
            1 1
                     1 0 6.043894 50.16116
## [6,]
            1
                          0 6.048243 50.16008
```

1.2. Variables

We can access to the variables writing:

p\$NAME_2

```
## [1] "Clervaux" "Diekirch" "Redange" "Vianden"
## [5] "Wiltz" "Echternach" "Remich" "Grevenmacher"
## [9] "Capellen" "Esch-sur-Alzette" "Luxembourg" "Mersch"
```

With the approach below you get a new SpatialPolygonsDataFrame with only one variable.

p[, 'NAME_2']

```
## class : SpatialPolygonsDataFrame
## features : 12
## extent : 5.74414, 6.528252, 49.44781, 50.18162 (xmin, xmax, ymin, ymax)
## crs : +proj=longlat +datum=WGS84 +no_defs
## variables : 1
## names : NAME_2
## min values : Capellen
## max values : Wiltz
```

1.3. Merge

Also we can join a data.frame with a spatial object using the merge() function as:

```
dat_fr <- data.frame(district=p$NAME_1,</pre>
                     canton=p$NAME_2,
                      value=round(runif(length(p), 100, 1000)))
dat_fr <- dat_fr[order(dat_fr$canton), ]</pre>
data_tot <- merge(p, dat_fr, by.x=c('NAME_1', 'NAME_2'),</pre>
                      by.y=c('district', 'canton'))
data_tot
## class
              : SpatialPolygonsDataFrame
## features
              : 12
              : 5.74414, 6.528252, 49.44781, 50.18162 (xmin, xmax, ymin, ymax)
## extent
## crs
               : +proj=longlat +datum=WGS84 +no_defs
## variables
             : 6
## names
                                NAME_2, ID_1, ID_2, AREA, value
                      NAME_1,
## min values
                                           1,
                   Diekirch, Capellen,
                                                  1,
                                                       76,
                                                             235
## max values
              : Luxembourg,
                                Wiltz,
                                           3,
                                                 12,
                                                      312,
                                                             934
```

1.4. Records (measures)

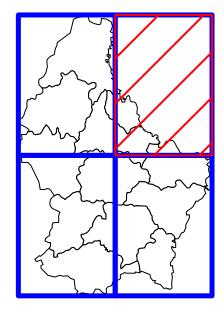
Assuming that every row is a record, then we can do:

```
i <- which(p$NAME_1 == 'Grevenmacher')</pre>
g <- p[i,]
g
              : SpatialPolygonsDataFrame
## class
## features
              : 3
## extent
               : 6.169137, 6.528252, 49.46498, 49.85403 (xmin, xmax, ymin, ymax)
               : +proj=longlat +datum=WGS84 +no_defs
## crs
## variables
              : 5
## names
               : ID_1,
                             NAME_1, ID_2,
                                               NAME_2, AREA
## min values :
                    2, Grevenmacher,
                                        6, Echternach,
                                                        129
## max values :
                    2, Grevenmacher,
                                       12,
                                               Remich,
                                                        210
```

1.5. Append and aggregate

Here we will consider an object u that contains 4 polygons, and u2 which is one of these 4 polygons. So,

```
u <- raster(p, nrow=2, ncol=2, vals=1:4)
names(u) <- 'Zone'</pre>
# coerce RasterLayer to SpatialPolygonsDataFrame
u <- as(u, 'SpatialPolygonsDataFrame')</pre>
## class
              : SpatialPolygonsDataFrame
## features : 4
## extent
             : 5.74414, 6.528252, 49.44781, 50.18162 (xmin, xmax, ymin, ymax)
## crs
              : +proj=longlat +datum=WGS84 +no_defs
## variables : 1
## names
           : Zone
## min values :
## max values :
u2 \leftarrow u[2,]
plot(p)
plot(u, add=TRUE, border='blue', lwd=5)
plot(u2, add=TRUE, border='red', lwd=2, density=3, col='red')
```



To append spatial objects of the same (vector) type you can use bind function:

```
b <- bind(p, u)
head(b)</pre>
```

```
##
     ID_1
                 NAME_1 ID_2
                                  NAME_2 AREA Zone
              Diekirch
## 1
        1
                           1
                                Clervaux
                                          312
                                                 NA
## 2
        1
              Diekirch
                           2
                                Diekirch
                                          218
                                                 NA
## 3
        1
              Diekirch
                           3
                                           259
                                 Redange
                                                 NA
                           4
                                           76
## 4
        1
               Diekirch
                                 Vianden
                                                 NA
## 5
        1
              Diekirch
                           5
                                   Wiltz
                                           263
                                                 NA
        2 Grevenmacher
## 6
                           6 Echternach
                                           188
                                                 NA
```

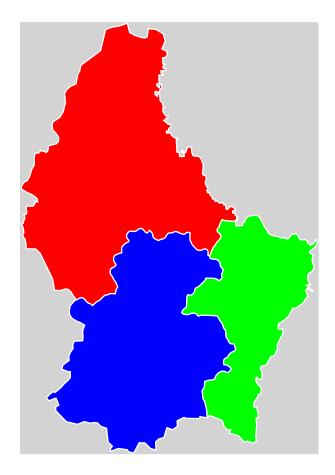
tail(b)

```
## ID_1 NAME_1 ID_2 NAME_2 AREA Zone
## 11 3 Luxembourg 10 Luxembourg 237 NA
## 12 3 Luxembourg 11 Mersch 233 NA
```

```
## 13
                  <NA>
                         NA
                                  <NA>
                                                1
        NA
                                          NA
## 14
        NA
                  <NA>
                         NA
                                  <NA>
                                          NA
                                                2
## 15
        NA
                  <NA>
                         NA
                                  <NA>
                                          NA
                                                3
## 16
        NA
                  <NA>
                         NA
                                  <NA>
                                          NA
                                                4
```

Now we will see an example of the aggregate function.

```
pa <- aggregate(p, by='NAME_1')
ua <- aggregate(u)
plot(ua, col='light gray', border='light gray', lwd=5)
plot(pa, add=TRUE, col=rainbow(3), lwd=3, border='white')</pre>
```

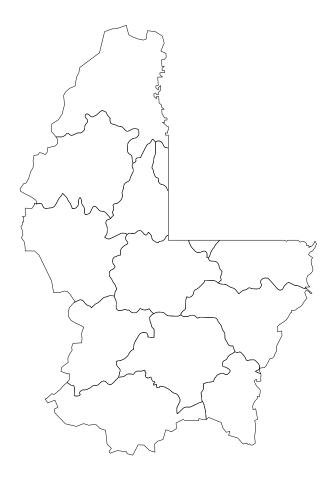


1.6. Overlay

a) Erase a part of a Spatial Polygons object

```
e <- erase(p, u2)
```

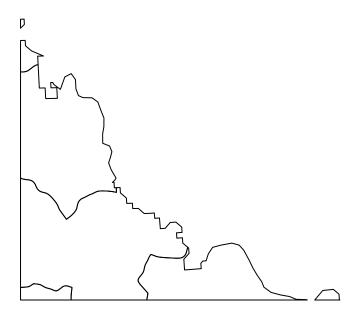
which is equivalent to write:



b) Intersect

If we want to intersect SpatialPolygons

```
inter <- intersect(p, u2)
plot(inter)</pre>
```

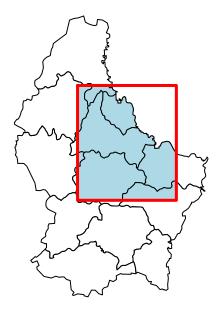


which is equivalent to write:

```
inter <- p * u2
```

Also it is possible to intersect with an extent (rectangle), for example:

```
e <- extent(6, 6.4, 49.7, 50)
pe <- crop(p, e)
plot(p)
plot(pe, col='light blue', add=TRUE)
plot(e, add=TRUE, lwd=3, col='red')</pre>
```



c) Union

we can join two SpatialPolygon objects doing the following

```
t <- union(p, u)
```

that is the same that do this:

```
t <- p + u
t
```

class : SpatialPolygonsDataFrame

features : 28

extent : 5.74414, 6.528252, 49.44781, 50.18162 (xmin, xmax, ymin, ymax)

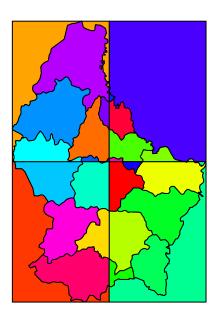
crs : +proj=longlat +datum=WGS84 +no_defs

variables : 6

names : ID_1, NAME_1, ID_2, NAME_2, AREA, Zone

```
## min values : 1, Diekirch, 1, Capellen, 76, 1 ## max values : 3, Luxembourg, 12, Wiltz, 312, 4
```

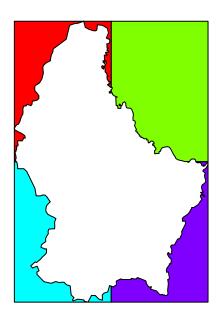
```
plot(t, col=sample(rainbow(length(t))))
```



d) Difference

We can find the symmetrical difference of two SpatialPolygons objects

```
difference <- symdif(u,p)
plot(difference, col=rainbow(length(difference)))</pre>
```



difference

```
## class : SpatialPolygonsDataFrame
```

features : 4

extent : 5.74414, 6.528252, 49.44781, 50.18162 (xmin, xmax, ymin, ymax)

crs : +proj=longlat +datum=WGS84 +no_defs

variables : 1
names : Zone
min values : 1
max values : 4

2. Raster data manipulation

2.1. Creating raster objects

Using the package "raster" and the function raster() we can build a RasterLayer easily. The default arguments in the function consider a 1 by 1 degree cells to get a raster data structure with a longitude/latitude coordinate reference system.

You can change these settings by providing additional arguments such as xmn, nrow, ncol, and/or crs, in the function.

Here an example to create a RasterLayer object.

```
library(raster)
x <- raster() # default parameters
x

## class : RasterLayer
## dimensions : 180, 360, 64800 (nrow, ncol, ncell)
## resolution : 1, 1 (x, y)
## extent : -180, 180, -90, 90 (xmin, xmax, ymin, ymax)
## crs : +proj=longlat +datum=WGS84 +no_defs</pre>
```

Set the coordinate reference system (CRS) (i.e., define the projection).

```
projection(x) <- "+proj=utm +zone=48 +datum=WGS84"
x

## class : RasterLayer
## dimensions : 180, 360, 64800 (nrow, ncol, ncell)
## resolution : 1, 1 (x, y)
## extent : -180, 180, -90, 90 (xmin, xmax, ymin, ymax)
## crs : +proj=utm +zone=48 +datum=WGS84 +units=m +no defs</pre>
```

The object x only contains the geometry of the raster, that is, the number of rows and columns, and where the raster is located in geographic space. If we want to add values associated with every cell, we can write:

```
r <- raster(ncol=10, nrow=10)
ncell(r)</pre>
```

```
## [1] 100
```

```
hasValues(r)

## [1] FALSE

values(r) <- runif(ncell(r))

hasValues(r)

## [1] TRUE

inMemory(r)

## [1] TRUE

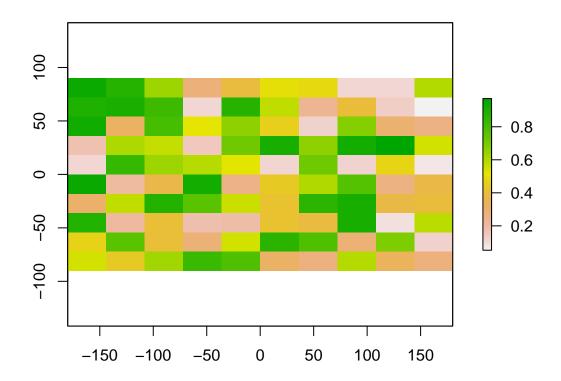
values(r)[1:10]

## [1] 0.9416227 0.8725538 0.6335872 0.2770864 0.3905521 0.5017617 0.4854508

## [8] 0.1101254 0.1090514 0.5971786

plot(r, main='Raster with 100 cells')
```

Raster with 100 cells



In cases where you change the number of columns or rows, you will lose the values associated with the RasterLayer. The same happens, in most cases, if you change the resolution directly since this can affect the number of columns or rows. If we change the extent, it does not affect the values because it only adjusts the resolution and neither the number of rows or columns.

hasValues(r) ## [1] TRUE dim(r) ## [1] 10 10 1 xmax(r)

[1] 180

Now we will change the maximum x coordinate of the extent (the bounding box) of the RasterLayer.

```
xmax(r) <- 0
hasValues(r)

## [1] TRUE

dim(r)

## [1] 10 10 1</pre>
```

But, if we change the number of columns, then the values disappear:

```
ncol(r) <- 6
hasValues(r)

## [1] TRUE

dim(r)

## [1] 10 6 1

xmax(r)

## [1] 0</pre>
```

Using the function raster() allows us to create a RasterLayer from another object, for example, a RasterLayer, RasterStack or a RasterBrick. However, it is common to create a RasterLayer object from a file.

The main characteristics of the "raster" package is that the raster datasets are stored in the disk and not in the memory of the computer (RAM). It means, we can work with large files only considering, from these large files, the structure of the data (number of rows, columns, spatial extent and the filename).

```
filename <- system.file("external/test.grd", package="raster")
filename</pre>
```

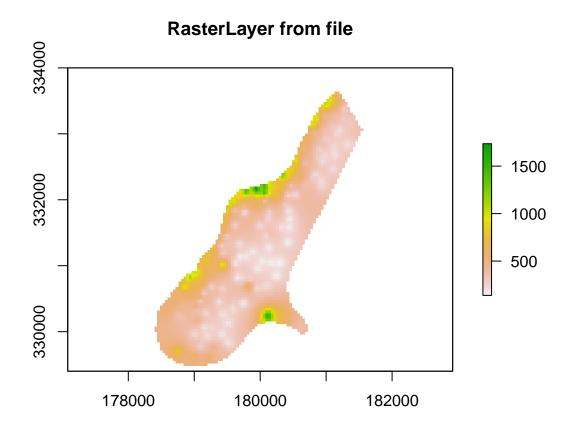
[1] "C:/Users/joaquin/AppData/Local/Programs/R/R-4.2.2/library/raster/external/test.grd"

```
r <- raster(filename)
filename(r)

## [1] "C:\\Users\\joaquin\\AppData\\Local\\Programs\\R\\R-4.2.2\\library\\raster\\external\\takentarrow
hasValues(r)

## [1] TRUE
inMemory(r)

## [1] FALSE
plot(r, main='RasterLayer from file')</pre>
```



We also can build a multi-layer objects from RasterLayer objects or from files. For example,

```
r1 <- r2 <- r3 <- raster(nrow=10, ncol=10)
# Assign random cell values
values(r1) <- runif(ncell(r1))
values(r2) <- runif(ncell(r2))
values(r3) <- runif(ncell(r3))</pre>
```

and combine those three RasterLayer into a RasterStack

```
s <- stack(r1, r2, r3)
## class
            : RasterStack
## dimensions: 10, 10, 100, 3 (nrow, ncol, ncell, nlayers)
## resolution : 36, 18 (x, y)
## extent : -180, 180, -90, 90 (xmin, xmax, ymin, ymax)
           : +proj=longlat +datum=WGS84 +no_defs
## crs
         :
## names
                   layer.1,
                                 layer.2,
                                              layer.3
## min values : 0.0019289961, 0.0009504876, 0.0022828276
## max values : 0.9994818, 0.9776395, 0.9707697
nlayers(s)
```

[1] 3

or combine those three RasterLayer in a RasterBrick

```
b1 <- brick(r1, r2, r3)
```

If you want to create a RasterBrick from an external file:

```
filename <- system.file("external/rlogo.grd", package="raster")
filename</pre>
```

[1] "C:/Users/joaquin/AppData/Local/Programs/R/R-4.2.2/library/raster/external/rlogo.grd"

```
b <- brick(filename)
b</pre>
```

```
## class : RasterBrick
## dimensions : 77, 101, 7777, 3 (nrow, ncol, ncell, nlayers)
## resolution : 1, 1 (x, y)
## extent : 0, 101, 0, 77 (xmin, xmax, ymin, ymax)
```

```
## crs : +proj=merc +lon_0=0 +k=1 +x_0=0 +y_0=0 +datum=WGS84 +units=m +no_defs
## source : rlogo.grd
## names : red, green, blue
## min values : 0, 0, 0
## max values : 255, 255, 255
nlayers(b)
```

[1] 3

and if we want to extract a RasterLayer from the previous RasterBrick created:

```
r <- raster(b, layer=2)
```

2.2. Raster algebra

Several functions can be applied to a raster object. For example; +, -, *, /, logical operators such as >, >=, <, ==, ! and functions like abs, round, ceiling, floor, trunc, sqrt, log, log10, exp, You can chek this doing the following:

```
r <- raster(ncol = 15, nrow = 15)
values(r) <- 1:ncell(r)</pre>
```

and make some operations:

```
s <- r + 10
s <- sqrt(s)
s <- s * r + 5
r[] <- runif(ncell(r))
r <- round(r)
r <- r == 1</pre>
```

If you are interested using multiple raster objects (in functions where this is relevant, such as range), these must have the same resolution and origin. Generally these objects have the same extent, but if they don't, the returned object covers the spatial intersection of the objects used.

```
r <- raster(ncol=4, nrow=4) # a new raster file
r[] <- 1

# Obtain values of r
v <- getValues(r)
length(v)</pre>
```

```
## [1] 16
```

```
# Join two raster data
s <- stack(r, r+1)
q \leftarrow stack(r, r+2, r+4, r+6)
x \leftarrow r + s + q
          : RasterBrick
## class
## dimensions : 4, 4, 16, 4 (nrow, ncol, ncell, nlayers)
## resolution : 90, 45 (x, y)
## extent : -180, 180, -90, 90 (xmin, xmax, ymin, ymax)
            : +proj=longlat +datum=WGS84 +no_defs
## crs
## source
            : memory
## names : layer.1, layer.2, layer.3, layer.4
## min values : 3,
                           6,
                                       7,
                                                10
                               6,
                                        7,
                                                10
## max values :
                   3,
Applying some functions to a raster data:
a <- mean(r,s, 10) # mean between r and s of 10 values
b \leftarrow sum(r,s)
st \leftarrow stack(r, s, a, b)
sst <- sum(st)</pre>
sst
          : RasterLayer
## class
## dimensions: 4, 4, 16 (nrow, ncol, ncell)
## resolution : 90, 45 (x, y)
## extent : -180, 180, -90, 90 (xmin, xmax, ymin, ymax)
            : +proj=longlat +datum=WGS84 +no_defs
## crs
## source : memory
## names : layer
            : 11.5, 11.5 (min, max)
## values
```

The function cellStats() allow us go get a single number summarizing the cell values of each layer (instead of a RasterLayer object)

```
## layer.1.1 layer.2.1 layer.2.2 layer.3
## 16 16 32 56 64
```

```
cellStats(sst, 'sum')
## [1] 184
```

2.3. Summarizing functions

When we are using raster object as first argument, the summary statistics functions such as min(), max() and mean() return a RasterLayer. To get a frequency table of a raster file or get the count number of cells, you can use the freq() function. You also can apply the function zonal() to a raster file to summarize the zones of the object (for example, areas with the same integer number).

```
r <- raster(ncol= 30, nrow= 20)
r[] <- runif(ncell(r))
cellStats(r, mean)</pre>
```

```
## [1] 0.520139
```

Applying the function zonal()

```
s <- r
s[] <- round(runif(ncell(r)) * 5)
zonal(r, s, 'mean')</pre>
```

```
## zone mean

## [1,] 0 0.5314018

## [2,] 1 0.4685671

## [3,] 2 0.5272841

## [4,] 3 0.5252983

## [5,] 4 0.5304549

## [6,] 5 0.5691401
```

Counting cells

```
freq(s)
```

```
##
         value count
## [1,]
             0
                   62
## [2,]
             1
                  126
## [3,]
             2
                  136
## [4,]
             3
                   96
## [5,]
             4
                  116
## [6,]
                   64
             5
```

```
freq(s, value=3)
## [1] 96
```

2.4. Helper functions

```
r <- raster(ncol=30, nrow=20)
ncol(r)
                      # num of cols
## [1] 30
nrow(r)
                      # num of rwos
## [1] 20
ncell(r)
                      # num of cells
## [1] 600
rowFromCell(r, 100)
                      # num of rows until the cell 100
## [1] 4
colFromCell(r, 100) # num of cols until the cell 100
## [1] 10
cellFromRowCol(r, 5, 5) # Cell number from row ans cols
## [1] 125
xyFromCell(r, 100) # Get coordinates in that cell
##
          Х
## [1,] -66 58.5
```

2.5. Accessing cell values

You can access to the values in a cell using the function getValues() for all the values or in a single row, or using the function getvaluesBlock() to get the values of a block (rectangle).

```
r <- raster(system.file("external/test.grd", package="raster"))
v <- getValues(r, 50)
v[35:39]

## [1] 743.8288 706.2302 646.0078 686.7291 758.0649

getValuesBlock(r, 50, 1, 35, 5)</pre>
```

[1] 743.8288 706.2302 646.0078 686.7291 758.0649

References

- Bivand, R. S., Pebesma, E. J., Gomez-Rubio, V., & Pebesma, E. J. (2008). Applied spatial data analysis with R (Vol. 747248717, pp. 237-268). New York: Springer.
- Spatial Data Science with R and "terra". https://rspatial.org/index.html.