Mapping in R

Geographic data

R has a series of special object types for spatial data. For many mapping / GIS tasks, you will need your data to be in one of these objects.

Spatial objects:

- SpatialPolygons
- SpatialPoints
- SpatialLines

Spatial objects + dataframes:

- SpatialPolygonsDataFrame
- SpatialPointsDataFrame
- SpatialLinesDataFrame

The tigris package lets you pull spatial data directly from the US Census. This data comes in directly as a spatial object.

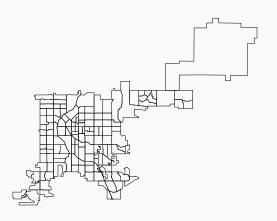
```
library(tigris)
denver_tracts <- tracts(state = "CO", county = "031", cb = TRUE)
class(denver_tracts)</pre>
```

```
## [1] "SpatialPolygonsDataFrame"
## attr(,"package")
## [1] "sp"
```

For more on this package, see the related article in *The R Journal*: https://journal.r-project.org/archive/accepted/walker.pdf.

You can plot a spatial object in R just by calling plot:

plot(denver_tracts)



These spatial objects come with a number of special *methods*, or functions that work for the specific object type. You can list these methods using name:

For example, bbox will print out the *bounding box* of the spatial object (range of latitudes and longitudes included).

```
bbox(denver_tracts)
```

```
## min max
## x -105.10993 -104.60030
## y 39.61443 39.91425
```

The is.projected and proj4string functions give you some information about the current Coordinate Reference System of the data.

```
is.projected(denver_tracts)

## [1] FALSE

proj4string(denver_tracts)
```

```
## [1] "+proj=longlat +datum=NAD83 +no_defs +ellps=GRS80 +towgs8
```

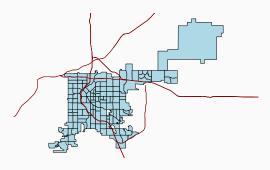
You can access a "slot" in a spatial object with a dataframe to pull out the data. This is similar to indexing a list. Just use @ instead of \$. For example, here's the dataframe for the denver_tracts spatial object:

```
head(denver_tracts@data[ , 1:4])
```

##		STATEFP	${\tt COUNTYFP}$	TRACTCE	AFFGEOID
##	25	08	031	000201	1400000US08031000201
##	26	08	031	000302	1400000US08031000302
##	27	08	031	001101	1400000US08031001101
##	28	08	031	002802	1400000US08031002802
##	29	08	031	003300	1400000US08031003300
##	30	80	031	004006	1400000US08031004006

You can add different layers of spatial objects onto the same plot. To do that, just use add = TRUE for added layers. For example, to add primary roads to the Denver census tract map, you could run:

```
denver_roads <- primary_roads()
plot(denver_tracts, col = "lightblue")
plot(denver_roads, add = TRUE, col = "darkred")</pre>
```



If you read in a shapefile, it will automatically be one of these shape objects. However, you can also convert other data into shape objects.

- Functions from sp package convert data into spatial objects
- fortify converts from a spatial object to a dataframe (useful for ggplot plotting)

You can use the fortify function from ggplot2 to convert the spatial object into a dataframe, so you can plot it using polygons in ggplot2.

```
fortify(denver_tracts) %>%
  dplyr::select(1:4) %>% dplyr::slice(1:5)
```

```
## # A tibble: 5 x 4

## long lat order hole

## <dbl> <dbl> <int> <lg!>
## 1 -105.0251 39.79400 1 FALSE

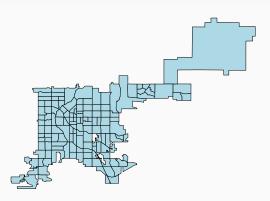
## 2 -105.0213 39.79398 2 FALSE

## 3 -105.0208 39.79109 3 FALSE

## 4 -105.0158 39.79107 4 FALSE

## 5 -105.0064 39.79105 5 FALSE
```

```
denver_tracts %>%
  fortify() %>%
  ggplot(aes(x = long, y = lat, group = group)) +
  geom_polygon(fill = "lightblue", color = "black") +
  theme_void()
```



Spatial objects can have different Coordinate Reference Systems (CRSs). CRSs can be *geographic* (e.g., WGS84, for longitude-latitude data) or *projected* (e.g., UTM, NADS83).

There is a website that lists projection strings and can be useful in setting projection information or re-projecting data:

http://www.spatialreference.org

Here is an excellent resource on projections and maps in R from Melanie Frazier: https://www.nceas.ucsb.edu/~frazier/RSpatialGuides/OverviewCoordinateReferenceSystems.pdf

To tell R the Coordinate Reference System of some data, set this attribute with proj4string (similar to setting column names with colnames):

```
## Generic code
proj4string(my_spatial_object) <- "+proj=longlat +datum=NAD83"</pre>
```

This does not create a projection. Instead, this is just how you tell R what projection the data already is in.

The CRS function creates CRS class objects that can be used to specify projections. You input a character string of projection arguments into this function (for example, CRS("+proj=longlat +datum=NAD27")). You can also use, however, use a shorter EPSG code for a projection (for example, CRS("+init=epsg:28992")).

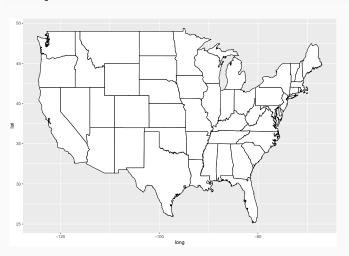
```
library(sp)
CRS("+proj=longlat +datum=NAD27")
## CRS arguments:
    +proj=longlat +datum=NAD27 +ellps=clrk66
## +nadgrids=@conus,@alaska,@ntv2_0.gsb,@ntv1_can.dat
CRS("+init=epsg:28992")
## CRS arguments:
    +init=epsg:28992 +proj=sterea +lat_0=52.15616055555555
## +lon_0=5.38763888888889 +k=0.9999079 +x_0=155000 +y 0=463000
```

To **change** the projection of a spatial object, you can use the spTransform function from the sp package.

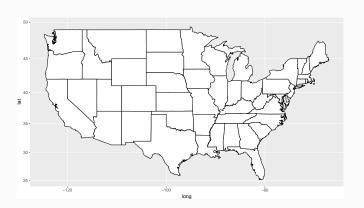
The coord_map function in ggplot2 can help you with map projections, as well. Here's an example from the help file with a US map.

The default is Cartesian coordinates:

usamap



Mercator projection:



Gilbert projection



Conic projection:



Shapefiles

- File format (ESRI, but usuable by other software)
- Not a single file, but rather a directory of related files (e.g., .shp, .shx, .dbf, .prj)
- Typically includes both geographic information (e.g., locations of county boundaries) and attribute information (e.g., median income of each county)
- To read shapefiles into R, use the readOGR function from the rgdal package
- You can also write out shapefiles you've created or modified in R, using writeOGR.

R as GIS

Capabilities

You can use R for a number of GIS-style tasks:

- Clipping
- Creating buffers
- Measuring area of a polygon
- Counting points in polygon

Capabilities

There are some advantages to using R for this:

- R is free
- You can write all code in a script, so research is more reproducible
- You save time and effort by staying in one software system, not going between different software

There are some advantages to GIS, too, though:

- More user-friendly at the start (point-and-click)
- R spatial functionality is still spread over lots of packages, with different syntax and conventions.

Spatial Points

For an example, I've cleaned up some FARs data at the driver level for 2001–2010:

```
load("../data/fars_colorado.RData")
driver_data %>%
  dplyr::select(1:5) %>% dplyr::slice(1:5)
```

```
## Warning in as.POSIX1t.POSIXct(x, tz): unknown timezone 'defau
## Denver'
```

```
## # A tibble: 5 x 5
                                      date latitude
##
    state st_case county
##
    <dbl>
            <dbl> <dbl>
                                    <dttm> <dbl>
## 1
        8
            80001
                     51 2001-01-01 10:00:00 39.10972
        8
            80002
                     31 2001-01-04 19:00:00 39.68215
## 2
## 3
        8
            80003
                     31 2001-01-03 07:00:00 39.63500
        8
            80004
                     31 2001-01-05 20:00:00 39.71304
## 4
        8
            80005
                     29 2001-01-05 10:00:00 39.09733
## 5
```

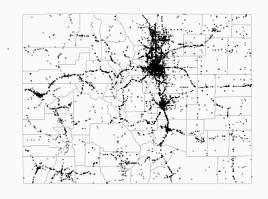
Spatial Points

Here is how you would plot fatal accidents (by driver) in Colorado without using spatial objects:

```
ggplot2::map_data("county", region = "Colorado") %>%
  ggplot2::ggplot(ggplot2::aes(x = long, y = lat,
                               group = subregion)) +
    ggplot2::geom_polygon(color = "gray", fill = NA) +
    ggplot2::theme void() +
    ggplot2::geom point(data = driver data,
                        ggplot2::aes(x = longitud,
                                     y = latitude,
                                     group = NULL),
                        alpha = 0.5, size = 0.7)
```

Spatial Points

Fatal accidents (by driver) in Colorado, 2001–2010:



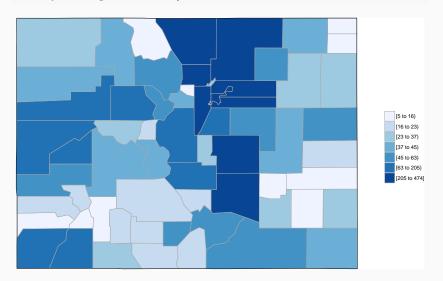
You can also make a choropleth of county accident counts without using spatial data, by using the choroplethr package.

To do this, you'll first need to use dplyr functions to limit to unique accidents (rather than drivers) and add up the number of accidents in each county. In this case, it's possible to add up accidents by county because county is included as a column in our data.

```
county_accidents <- driver_data %>%
  dplyr::select(state, st case, county, latitude, longitud) %>%
  dplyr::distinct() %>%
  dplyr::mutate(county = str_pad(county, width = 3,
                          side = "left", pad = "0")) %>%
  tidyr::unite(region, state, county, sep = "") %>%
  dplyr::group_by(region) %>%
  dplyr::summarize(value = n()) %>%
  dplyr::mutate(region = as.numeric(region))
county accidents %>% slice(1:4)
```

```
## # A tibble: 4 x 2
## region value
## <dbl> <int>
## 1 8001 372
## 2 8003 47
## 3 8005 305
```

county_choropleth(county_accidents, state_zoom = "colorado")



This "out-of-the-box" solution let us looks at accident counts by county, but what if we want to look at a geographical unit for which we don't have an identifying column?

For example, we might want to look at accident counts by census tract in Denver. To do this, we'll need to link each accident (point) to a census tract (polygon), and then we can count up the number of points linked to each polygon.

First, I've created a dataframe with only accidents in Denver (based on the county column in the accident data):

```
denver_fars <- driver_data %>%
  filter(county == 31)
```

To do this, both the census tracts and the accident data need to be in spatial objects.

```
library(sp)
denver_fars_sp <- denver_fars %>%
    dplyr::rename(longitude = longitud)
coordinates(denver_fars_sp) <- c("longitude", "latitude")
proj4string(denver_fars_sp) <- CRS("+init=epsg:4326")</pre>
```

Note that the dataframe is changed into a spatial object by changing its coordinates attribute, and that the CRS was set uas the proj4string attribute.

##

```
summary(denver_fars_sp)
## Object of class SpatialPointsDataFrame
## Coordinates:
##
                  min
                     max
## longitude -105.10973 -104.0122
## latitude 39.61715 39.8381
## Is projected: FALSE
## proj4string:
## [+init=epsg:4326 +proj=longlat +datum=WGS84 +no_defs +ellps=W
## + towgs84 = 0,0,0
## Number of points: 695
## Data attributes:
##
       state st case county
                                             date
   Min. :8 Min. :80001 Min. :31 Min. :2001-01-03
##
   1st Qu.:8 1st Qu.:80121 1st Qu.:31 1st Qu.:2003-01-06
##
```

Median:8 Median:80268 Median:31

Median : 2005-01-35

To be able to pair up polygons and points, their spatial objects need to have the same CRS. To help later with calculating the area of each polygon, I'll use a projected CRS that is reasonable for Colorado.

```
proj4string(denver_tracts)

## [1] "+proj=longlat +datum=NAD83 +no_defs +ellps=GRS80 +towgs8

CRS(proj4string(denver_tracts))

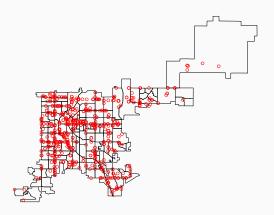
## CRS arguments:
## +proj=longlat +datum=NAD83 +no_defs +ellps=GRS80 +towgs84=0,
```

To reproject spatial data, you can use the spTransform function:

The spTransform function transforms the coordinates in a spatial object into a new coordinate reference system.

Here is a map of the tracts with the accidents overlaid:

```
plot(denver_tracts)
plot(denver_fars_proj, add = TRUE, col = "red", pch = 1)
```



Now, the data's in a format where we can link spatial points to spatial polygons.

The poly.counts function in the GISTools package will measure the number of points that fall within each polygon.

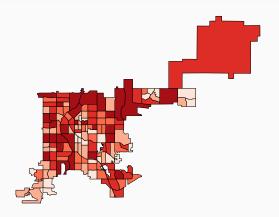
It results in a vector with one element for each polygon (census tract in our example), where the element name identifies the polygon and the cell value gives the count within that polygon.

```
tract_counts <- poly.counts(denver_fars_proj, denver_tracts)
head(tract_counts)</pre>
```

```
## 25 26 27 28 29 30
## 7 2 2 0 0 4
```

You can use a choropleth to show these accident counts. The quickest way to do this is probably to use the choropleth function in the GISTools package.

choropleth(denver_tracts, tract_counts)



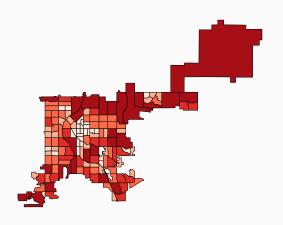
Determining area

There is another function in the package that calculates the area of each polygon.

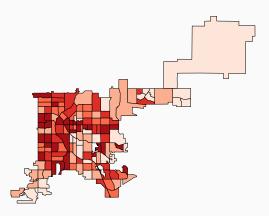
```
library(GISTools)
head(poly.areas(denver_tracts_proj))
```

```
## 25 26 27 28 29 30
## 2100172.2 1442824.1 897886.3 881530.5 1282812.2 1948187.1
```

choropleth(denver_tracts, poly.areas(denver_tracts_proj))



You can combine these ideas to create a choropleth of the rate of fatal accidents per population in Denver census tracts.



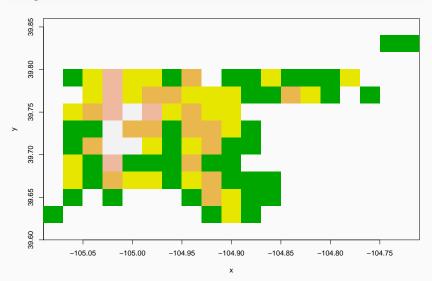
When mapping in R, you may also need to map raster data.

You can think of this as pixels— the graphing region is divided into even squares, and color is constant within each square.

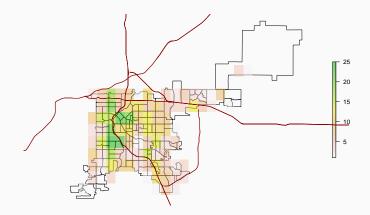
There are functions that allow you to "rasterize" data. That is, you take spatial points data, divide the region into squares, and count the number of points (or other summary) within each square.

```
bbox(denver_fars_sp)
##
                     min
                               max
## longitude -105.10973 -104.0122
## latitude 39.61715 39.8381
library(raster)
denver raster \leftarrow raster(xmn = -105.09, ymn = 39.60,
                         xmx = -104.71, ymx = 39.86,
                         res = 0.02)
den_acc_raster <- rasterize(geometry(denver_fars_sp),</pre>
                             denver_raster,
                             fun = "count")
```

image(den_acc_raster, col = terrain.colors(5))



```
plot(denver_tracts)
plot(den_acc_raster, add = TRUE, alpha = 0.5)
plot(denver_roads, add = TRUE, col = "darkred")
```



Other capabilities

You can also use R for other spatial tasks:

- Kernel density estimation
- Identifying clusters
- Kriging
- Measuring spatial autocorrelation

Find out more

- Applied Spatial Data Analysis with R by Roger Bivand (available online through CSU library)
- An Introduction to R for Spatial Analysis and Mapping by Chris Brunsdon and Lex Comber
- CRAN Spatial Data Task View
- R Spatial Cheatsheet
- Great blog post (among many) by Zev Ross