mapping_R Markup

Introduction to R packages for geospatial data

This tutorial introduces a variety of R packages that may be used to manipulate and visualize geospatial data in R. Below we provide short descriptions of the packages used in each example throughout this tutorial, which can to be installed with the following code:

Packages used in:

- Example 1: mapping census data
 - ACS & choroplethr can be used together to easily make choroplethr maps using data from the American Community Survey (ACS), yearly census data collected by the U.S. Census Bureau. To access ACS data you need an API key. Visit http://api.census.gov/data/key_signup.html, request a key, and paste it into the line below:
 - choroplethrMaps contains a global map and maps of the USA used by the choroplethr package.

```
api.key.install("<ACS API key>")
```

- Example 2: working with shapefiles, projections, and visualization
 - maptools contains functions for reading and manipulating geographic data, including ESRI shapefiles.
 - **rgdal** geospatial data abstraction and projection / transformation.
 - RColorBrewer provides color schemes that are especially useful for creating thematic maps.
 - **ggplot2** package for creating and customizing graphics is R.
 - rgeos contains functions for performing geometric analysis. For example gLength() calculates the length of input geometry, while gBuffer() adds a buffer to an input feature.
 - mapproj WHERE IS THIS USED? simple package for converting from latitude and logitude into projected coordinates.
- Example 3: network-type map
 - maps another simple set of tools for creating maps, with links to several databases of spatial data.
 - geosphere supports trigonometric calculations for geographic applications. For example, computing distance to the horizon from a given location and altitude.
 - reshape reshapes data from 'wide' format (where repeated measurements are located across multiple columns) to 'long' format (where repeated measurements are spread across unique rows)

EXAMPLE 1: mapping census data

Packages used - ACS, choroplethr, choroplethrMaps

After installing (see introduction), load needed packages:

```
library(acs)
library(choroplethr)
library(choroplethrMaps)
```

We need an api key to access the ACS data. Visit http://api.census.gov/data/key_signup.html, request a key, and paste it into the line below:

```
api.key.install("<ACS API key>")
```

Great, now we have access to the census data. Table B19301 contains per capita income for the year of 2011. Lets plot it!

county_choropleth_acs(tableId="B19301")

Per capita income in the past 12 months (in 2011 inflation-adjusted dollars)

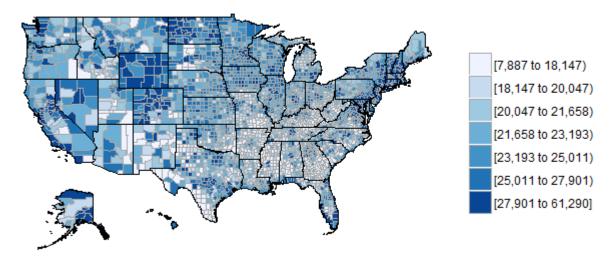


Figure 1: US county chloropleth using ACS per capita income data

To see the description of a function and its arguments in R, place a "?" before its name:

?county_choropleth_acs

EXAMPLE 2: working with shapefiles, projections, and visualization

Packages used - maptools, rgdal, RColorBrewer, ggplot2

```
library(maptools)
library(rgdal)
```

In this example we will work with the data provided along with this tutorial. Make sure you have unzipped the folder county_census before proceeding! The following prompts you to select the provided county census shapefiles at the path ... county_census/County_2010Census_DP1.shp.

counties <- readShapeSpatial(file.choose(),proj4string=CRS("+proj=longlat +datum=WGS84"))</pre>

Note that the second parameter of the maptools function readShapeSpatial was a string representing the projection of the data, called a prj4 string. So far we've used "+proj=longlat +datum=WGS84", which is simply unprojected longitude and latitude coordinates.

Inspect the first few rows of the counties data to get a feel for its structure:

head(counties@data)

Census data assigns codes to counties using the Federal Information Processing Standard (FIPS). A FIPS code starts with two digits representing the state, and is followed by three digits representing the county. For example, Florida is 12 and Clay County Florida is 12019. To select all the counties in Florida we can use a regular expression matching all codes that start with "12":

florida <- counties[substring(counties\$GEOID10,1,2)=="12",]
plot(florida)</pre>



You can look up other state and county codes using the U.S. Census Bureau site: https://www.census.gov/geo/reference/codes/cou.html

Projection and Layering with RGDAL

Next we'll work more with projections using library rgdal. We're going to read in a shape file of cultural points in Florida from the supplied data, again using function readShapeSpatial(). We know already that our cultural centers layer uses NAD83(HARN) / Florida GDL Albers. We can make the EPSG data frame of projections to find the prj4 string for this projection (use ?make_EPSG() to find out more about this table):

```
library(rgdal)
EPSG <- make_EPSG()</pre>
```

We can use regular expressions to search the note field of EPSG for any that refer to Florida:

```
EPSG[grep("florida", EPSG$note, ignore.case=TRUE), 1:2]
```

```
##
         code
                                                    note
                         # NAD83 / Florida East (ftUS)
## 705
         2236
  706
         2237
                         # NAD83 / Florida West (ftUS)
##
## 707
                        # NAD83 / Florida North (ftUS)
         2238
## 1245
         2777
                          # NAD83(HARN) / Florida East
                           # NAD83(HARN) / Florida West
## 1246
         2778
## 1247
         2779
                         # NAD83(HARN) / Florida North
## 1349
         2881
                   # NAD83(HARN) / Florida East (ftUS)
## 1350
                   # NAD83(HARN) / Florida West (ftUS)
         2882
                  # NAD83(HARN) / Florida North (ftUS)
## 1351
         2883
## 1553
         3086
                           # NAD83 / Florida GDL Albers
## 1554
         3087
                    # NAD83(HARN) / Florida GDL Albers
                      # NAD83(NSRS2007) / Florida East
## 1978
         3511
## 1979
         3512
               # NAD83(NSRS2007) / Florida East (ftUS)
## 1980
         3513
                # NAD83(NSRS2007) / Florida GDL Albers
## 1981
         3514
                     # NAD83(NSRS2007) / Florida North
## 1982
         3515 # NAD83(NSRS2007) / Florida North (ftUS)
                      # NAD83(NSRS2007) / Florida West
## 1983
         3516
## 1984
         3517
               # NAD83(NSRS2007) / Florida West (ftUS)
## 3074
         6437
                           # NAD83(2011) / Florida East
## 3075
                   # NAD83(2011) / Florida East (ftUS)
         6438
         6439
## 3076
                    # NAD83(2011) / Florida GDL Albers
                         # NAD83(2011) / Florida North
## 3077
         6440
## 3078
         6441
                  # NAD83(2011) / Florida North (ftUS)
## 3079
                           # NAD83(2011) / Florida West
         6442
## 3080
                   # NAD83(2011) / Florida West (ftUS)
         6443
## 3747 26758
                                 # NAD27 / Florida East
## 3748 26759
                                 # NAD27 / Florida West
                                # NAD27 / Florida North
## 3749 26760
## 3895 26958
                                 # NAD83 / Florida East
## 3896 26959
                                 # NAD83 / Florida West
                                # NAD83 / Florida North
## 3897 26960
```

We see the code we're looking for is 3087. Extract the prj4 string from this dataframe:

```
subset(EPSG, code==3087)
prjstring <- subset(EPSG, code==3087)$prj4</pre>
```

Inspect our prjstring variable if you want to see the format of the prj4 variable.

Now that we have the appropriate prj4 we can read in the cultural centers data. The following prompts you to select the shape file. Select the actual .shp file in the provided data from ...cultural_centers/gc_culturecenter_oct15.shp.

```
cultural <- readShapeSpatial(file.choose(),proj4string=CRS(prjstring))</pre>
```

Before we overlay the cultural points, we need to transform this layer to match that of the Florida counties layer - simple longitude and latitude in WGS84:

```
cultural_proj <- spTransform(cultural, CRS("+proj=longlat +datum=WGS84"))
plot(florida)
points(cultural_proj)</pre>
```



You can play around with the symbology for your map with some additional arguments in the points function. For example:

```
plot(florida)
points(cultural_proj, cex=.8, pch=24, col=554, bg="grey")
```



join polygon data to points

county_data <- over(cultural_proj,florida)
cultural_proj\$pop <- county_data\$DP0010001</pre>

set colors

```
library(RColorBrewer)
brks <- c(.5,1,1.5,2) * 1000000
cols <- brewer.pal(5, "Greens")</pre>
mapcols <- cols[findInterval(cultural_proj$pop, vec=brks)]</pre>
plot(cultural_proj,col=mapcols,pch=20)
base R instructions for choropleth
brks <-c(25,30,35,40,45,50,55,60,65)
cols <- brewer.pal(8,"Purples")</pre>
mapcols <- cols[findInterval(florida$DP0020001, vec=brks)]</pre>
plot(florida,col=mapcols,border="white")
legend("bottomleft", legend = levels(cut(florida$DP0020001, brks)), fill = cols, title = "Median Age")
using ggplot2
library(ggplot2)
fl_shapes <- fortify(florida,region="GEOID10")</pre>
ggplot() + geom_map(data=as.data.frame(florida),aes(map_id = GEOID10,fill=DP0020001), map=fl_shapes) +
EXAMPLE 3: network-type map
Packages used - maps, geosphere, reshape
library(maps)
library(geosphere)
library(reshape)
select - state_shapes/tl_2014_us_state.shp
state <- readShapeSpatial(file.choose())</pre>
select - /state\_migrations\_2014.csv
migration <- read.csv(file.choose())</pre>
```

```
centrs <- data.frame(as.character(state@data$NAME),coordinates(state))</pre>
colnames(centrs) <- c("name","long","lat")</pre>
## Reshape - melt
migration <- migration[c(1,6:56)]
long_mig <- melt(migration,id.vars="from_state")</pre>
map("state")
define draw_from_state function
draw_from_state <- function(centrs, migrations, state_name, color=rgb(0,0,0,alpha=0.5)) {</pre>
    migrations$variable <- sub("."," ",migrations$variable,fixed=TRUE)</pre>
    migrations <- migrations[migrations$variable==state_name & migrations$from_state != state_name,]
    for(i in 1:nrow(migrations)){
        if (nrow(centrs[centrs$name==as.character(migrations[i,]$from_state),]) > 0){
            from_long <- centrs[centrs$name==as.character(migrations[i,]$from_state),]$long</pre>
            from_lat <- centrs[centrs$name==as.character(migrations[i,]$from_state),]$lat
            to_long <- centrs[centrs$name==as.character(migrations[i,]$variable),]$long
            to_lat <- centrs[centrs$name==as.character(migrations[i,]$variable),]$lat
            number <- migrations[i,]$value</pre>
            lines(gcIntermediate(c(from_long, from_lat), c(to_long, to_lat), n=50, addStartEnd=TRUE),lw
    }
}
draw_from_state(centrs, long_mig, "Florida", rgb(0,0,1,0.5))
xlim \leftarrow c(-171.738281, -56.601563)
ylim <- c(12.039321, 71.856229)
map("world", col="#f2f2f2", fill=TRUE, bg="white", lwd=0.05, xlim=xlim, ylim=ylim)
draw_from_state(centrs, long_mig, "Wyoming", rgb(1,0,0,.5))
```