mapping_R Markup

Introduction

This tutorial walks through three examples demonstrating how geospatial data can be manipulated and visualized in R. We assume you have downloaded R and know how to run R commands. For those unfamiliar with R or GIS mapping, below are a few introductory resources for exploration:

- R
- Tutorials for learning R on r-bloggers.com
- Quick R by Robert I. Kabacoff, Ph.D.
- R for Data Science by Garrett Grolemund and Hadly Wickham
- GIS mapping
 - A Gentle Introduction to GIS by QGIS; map projections in particular
 - Introduction to GIS by the International Centre for Integrated Mountain Development; information about map projections on p.72

As you go through the tutorial, any time you'd like to see more information about a function, place a ? before its name:

```
`?`(county_choropleth_acs)
```

R packages for geospatial data

Each example within this tutorial introduces a variety of R packages to visualize maps in R. Below we provide short descriptions of the R 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 choroplethGCT-PEPANNRES 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.
- 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 simple package for converting from latitude and logitude into projected coordinates.
- EXAMPLE 3: Migration Distances 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)
 - mapproj simple package for converting from latitude and logitude into projected coordinates.

EXAMPLE 1: Mapping Census Data

This example introduces how to access and plot American Community Survey (ACS) yearly census data by the U.S. Census Bureau through choropleth maps.

${\bf Packages\ used:\ ACS,\ choroplethr,\ choroplethrMaps}$

After installing packages (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")
```

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

```
`?`(county_choropleth_acs)
```

You can explore the ACS data on the Census Bureau's website: http://factfinder.census.gov/faces/nav/jsf/pages/searchresults.xhtml Be sure to set the dataset to one of the ACS sources under Topics -> dataset. Use the ID column to find your tableID variable. Here's another example to get you started playing around with these functions:

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

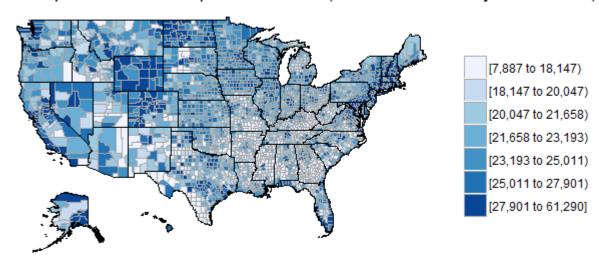


Figure 1: US county chloropleth using ACS per capita income data

Median Age by Sex: Median age -- Total:



EXAMPLE 2: Working with Shapefiles, Projections, and Visualization

This example loads and plots both polygon and point shapefiles, uses the rgdal library to change projections, uses the over function to join datasets, and provides a few visualization examples.

Packages used: maptools, rgdal, RColorBrewer, ggplot2

After installing packages (see introduction), load needed packages:

```
library(maptools)
library(rgdal)
library(RColorBrewer)
library(ggplot2)
```

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

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>
```



Change symbology of points

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 = 0.8, pch = 24, col = 554, bg = "grey")
```



Join polygon data to points

Use over() function to overlay cultural_proj(points) and florida(polygon) layers.

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

Change symbology of points to gradient of colors

```
library(RColorBrewer)
```

```
brks <- c(0.5, 1, 1.5, 2) * 1e+06
cols <- brewer.pal(5, "Greens")

mapcols <- cols[findInterval(cultural_proj$pop, vec = brks)]
plot(cultural_proj, col = mapcols, pch = 20)</pre>
```

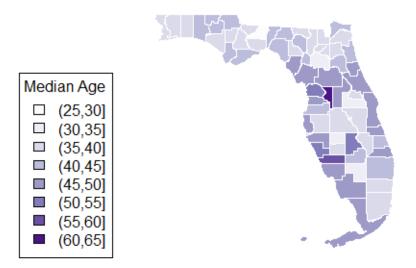


Base R instructions for choropleth

```
brks <- c(25, 30, 35, 40, 45, 50, 55, 60, 65)
cols <- brewer.pal(8, "Purples")

mapcols <- cols[findInterval(florida$DP0020001, vec = brks)]
plot(florida, col = mapcols, border = "white")</pre>
```

```
legend("bottomleft", legend = levels(cut(florida$DP0020001, brks)), fill = cols,
    title = "Median Age")
```



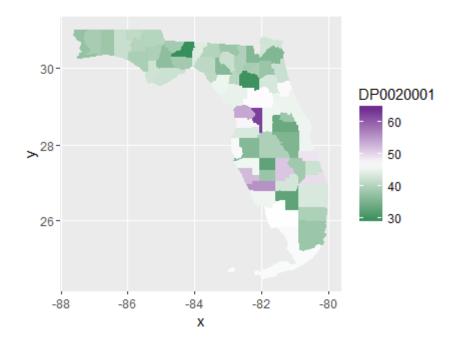
Using ggplot2 package

ggplot2 is a powerful package for visualizing data and contains a variety of functions that make it versatile. See http://docs.ggplot2.org/current/index.html for a helpful list describing ggplot2's various functions.

```
library(ggplot2)
```

```
fl_shapes <- fortify(florida, region = "GEOID10")

ggplot() + geom_map(data = as.data.frame(florida), aes(map_id = GEOID10, fill = DP0020001),
    map = fl_shapes) + expand_limits(x = fl_shapes$long, y = fl_shapes$lat) +
    scale_fill_gradient2(low = "seagreen", mid = "white", high = "darkorchid4",
        midpoint = 47, limits = c(29, 65)) + coord_map(projection = "mercator")</pre>
```



EXAMPLE 3: Migration Distances Map

This example loads both a shapefile and .csv file, transforms the data set with a melt function, and writes a function that maps the distance data contained in the .csv file.

Packages used: maps, geosphere, reshape, maptools

After installing packages (see introduction), load needed packages:

```
library(maps)
library(geosphere)
library(reshape)
library(maptools)
```

Import shapefile and .csv file

Import shapefile of map of continental United States. Choose state_shapes/tl_2014_us_state.shp file when prompted.

```
state <- readShapeSpatial(file.choose())</pre>
```

Import data file of migration distances between U.S.A. states. Choose state_migrations_2014.csv file when prompted.

```
migration <- read.csv(file.choose())</pre>
```

Extract state names and geographic coordinate (latitude and longitude) information from state shapefile; save it into a data frame called centrs.

```
centrs <- data.frame(as.character(state@data$NAME), coordinates(state))
colnames(centrs) <- c("name", "long", "lat")</pre>
```

Reshape data with melt function

Redefine the migration data set to only include columns 1 & 6-through-56 of data. Then use melt function from reshape package to transform data set into rows representing unique instances of data, based on a selected variable id (in our case, the from_state variable). For more on the melt function, see https://www.r-bloggers.com/melt/.

```
migration <- migration[c(1, 6:56)]
long_mig <- melt(migration, id.vars = "from_state")</pre>
```

Create a mapping function

Create a function named data_from_state that maps migration distances from any state selected. NOTE: To use this function, a map must be drawn (i.e. a new plot must be called) first.

```
draw_from_state <- function(centrs, migrations, state_name, color = rgb(0, 0,</pre>
    0, alpha = 0.5)
    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),
            from_long <- centrs[centrs$name == as.character(migrations[i, ]$from_state),</pre>
            from_lat <- centrs[centrs$name == as.character(migrations[i, ]$from_state),</pre>
            to_long <- centrs[centrs$name == as.character(migrations[i, ]$variable),</pre>
            to_lat <- centrs[centrs$name == as.character(migrations[i, ]$variable),</pre>
                 ]$lat
            number <- migrations[i, ]$value</pre>
            lines(gcIntermediate(c(from_long, from_lat), c(to_long, to_lat),
                 n = 50, addStartEnd = TRUE), lwd = sqrt(number)/50, col = color)
        }
    }
}
```

Using our mapping function: example 1

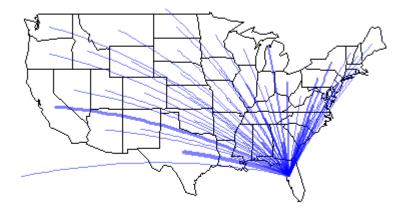
First draw imported state shapefile.

```
map("state")
```



Now use written function to map migration distances from Florida.

```
draw_from_state(centrs, long_mig, "Florida", rgb(0, 0, 1, 0.5))
```



Using our mapping function: example 2

First draw a world map (limited to North and Central America by creating x- and y-coordinate limits), and subsquently use written function to map migration distances from Wyoming onto map.

