

mapping_R Markup

R packages for working with geospatial data

Each example below starts by loading the dependant libraries. Make sure you have installed the packages below that you will be using:

```
install.packages(c("acs", "choroplethr", "choroplethrMaps", "maptools", "rgeos", "mapproj", "RColorBrewer", "maps"))
```

- **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:

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

- **choroplethrMaps** - contains maps of the USA used by the choroplethr package.
- **maptools** - contains functions for reading and manipulating geographic data, including ESRI shapefiles.
- **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 longitude into projected coordinates.
- **RColorBrewer** - provides color schemes that are especially useful for creating thematic maps.
- **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.
- **rgdal** - geospatial data abstraction and projection / transformation.
- **reshape** - ???
- **ggplot2** - package for creating and customizing graphics is R.

Example 1: mapping census data using ACS and choroplethr

Dependancies: acs, choroplethr, choroplethrMaps

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

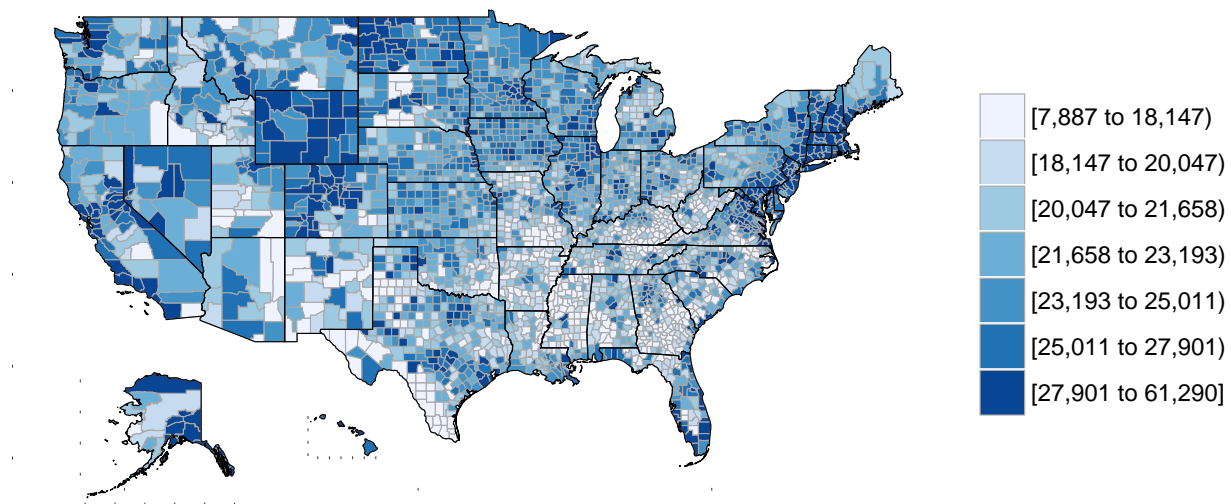
We need an api key to access the ACS data. Visit http://api.census.gov/data/key_signup.html, request a key, and paste it into the line below:

```
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")
```

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



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 and projections

Dependencies: `maptools`, `rgdal`

```
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"))
```

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)
```



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()
```

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)
## 1245	2777	# NAD83(HARN) / Florida East
## 1246	2778	# NAD83(HARN) / Florida West
## 1247	2779	# NAD83(HARN) / Florida North
## 1349	2881	# NAD83(HARN) / Florida East (ftUS)
## 1350	2882	# NAD83(HARN) / Florida West (ftUS)
## 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	3516	# NAD83(NSRS2007) / Florida West
## 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 6442          # NAD83(2011) / Florida West
## 3080 6443      # NAD83(2011) / Florida West (ftUS)
## 3747 26758          # NAD27 / Florida East
## 3748 26759          # NAD27 / Florida West
## 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
```

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))
```

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:

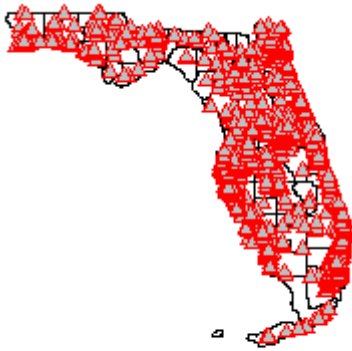
```
cultural_proj <- spTransform(cultural, CRS("+proj=longlat +datum=WGS84"))
```

```
plot(florigia)
points(cultural_proj)
```



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

```
plot(florigia)
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
```

set colors

```
library(RColorBrewer)
```

```
brks <- c(.5,1,1.5,2) * 1000000
cols <- brewer.pal(5,"Greens")
```

```
mapcols <- cols[findInterval(cultural_proj$pop, vec=brks)]
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")
```

```
mapcols <- cols[findInterval(florida$DP0020001, vec=brks)]
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")
```

```
ggplot() + geom_map(data=as.data.frame(florida),aes(map_id = GEOID10,fill=DP0020001), map=fl_shapes) +
```

networky type example

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

select - state__shapes/tl_2014_us_state.shp

```
state <- readShapeSpatial(file.choose())
```

select - /state__migrations__2014.csv

```
migration <- read.csv(file.choose())
centrs <- data.frame(as.character(state@data$NAME),coordinates(state))
colnames(centrs) <- c("name","long","lat")
```

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

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

define draw__from__state function

```
draw_from_state <- function(centrs, migrations, state_name, color=rgb(0,0,0,alpha=0.5)) {
  migrations$variable <- sub(".", " ", migrations$variable, fixed=TRUE)
  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),]$long
      from_lat <- centrs[centrs$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
      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 <- 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))
```