

GIS Lab

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Getting Started

```
library( maptools )
```

```
## Warning: package 'maptools' was built under R version 3.2.3
```

```
## Loading required package: sp
## Checking rgeos availability: FALSE
##      Note: when rgeos is not available, polygon geometry      computations in maptools
depend on gpclib,
##      which has a restricted licence. It is disabled by default;
##      to enable gpclib, type gpclibPermit()
```

```
library( sp )
library( dplyr )
```

```
## Warning: package 'dplyr' was built under R version 3.2.3
```

```
##
## Attaching package: 'dplyr'
##
## The following objects are masked from 'package:stats':
##
##      filter, lag
##
## The following objects are masked from 'package:base':
##
##      intersect, setdiff, setequal, union
```

```
library( RColorBrewer )
```

```
syr <- readShapePoly( fn="01-05-2015", proj4string=CRS("+proj=longlat +datum=WGS84") )
dat <- as.data.frame( syr )
```

1. What is the average price of a single family home in each neighborhood?

```
sing.fam <- dat[dat$LandUse == "Single Family", ]
tapply(sing.fam$AssessedVa, sing.fam$Nhood, mean, na.rm=T)
```

##	Brighton	Court-Woodlawn	Downtown
##	40094.17	66871.46	110000.00
##	Eastwood	Elmwood	Far Westside
##	69104.49	49731.14	53933.33
##	Franklin Square	Hawley-Green	Lakefront
##	NA	47813.46	37808.33
##	Lincoln Hill	Meadowbrook	Near Eastside
##	61543.79	118554.66	47798.92
##	Near Westside	North Valley	Northside
##	35441.17	55983.21	47844.53
##	Outer Comstock	Park Ave.	Prospect Hill
##	73063.41	43215.57	54648.28
##	Salt Springs	Sedgwick	Skunk City
##	63101.46	127879.40	45020.43
##	South Campus	South Valley	Southside
##	109180.00	72826.85	39907.01
##	Southwest	Strathmore	Tipp Hill
##	39231.46	90516.28	63720.00
##	University Hill	University Neighborhood	Washington Square
##	85041.18	114846.58	50096.53
##	Westcott	Winkworth	
##	86743.18	109447.02	

2. What is the average value of one acre of land in each neighborhood?

```
dat$acreprice <- dat$AssessedVa/dat$Acres
tapply(dat$acreprice, dat$Nhood, mean, na.rm=T)
```

##	Brighton	Court-Woodlawn	Downtown
##	303053.6	550582.0	3252663.7
##	Eastwood	Elmwood	Far Westside
##	558296.7	345209.3	552823.7
##	Franklin Square	Hawley-Green	Lakefront
##	817599.0	653100.2	562909.0
##	Lincoln Hill	Meadowbrook	Near Eastside
##	508069.6	477755.9	427704.6
##	Near Westside	North Valley	Northside
##	279609.0	356223.0	571657.2
##	Outer Comstock	Park Ave.	Prospect Hill
##	404540.1	455879.1	897674.4
##	Salt Springs	Sedgwick	Skunk City
##	431394.5	664994.5	380891.4
##	South Campus	South Valley	Southside
##	647834.9	347411.5	299747.6
##	Southwest	Strathmore	Tipp Hill
##	253118.3	545052.0	651839.8
##	University Hill	University Neighborhood	Washington Square
##	1979056.2	749673.8	553236.6
##	Westcott	Winkworth	
##	675517.6	367041.8	

#or with dplyr package

```
dat2 <- mutate( dat, price.acre = AssessedVa / Acres )
by.hood <- group_by( dat2, Nhood )
prices <- summarise( by.hood, count = n(),
                    ave.price = mean(price.acre, na.rm= TRUE) )
as.data.frame( prices )
```

##	Nhood	count	ave.price
## 1	Brighton	2302	303053.6
## 2	Court-Woodlawn	2402	550582.0
## 3	Downtown	389	3252663.7
## 4	Eastwood	4889	558296.7
## 5	Elmwood	1444	345209.3
## 6	Far Westside	1027	552823.7
## 7	Franklin Square	89	817599.0
## 8	Hawley-Green	367	653100.2
## 9	Lakefront	312	562909.0
## 10	Lincoln Hill	1123	508069.6
## 11	Meadowbrook	1878	477755.9
## 12	Near Eastside	441	427704.6
## 13	Near Westside	1772	279609.0
## 14	North Valley	1531	356223.0
## 15	Northside	3261	571657.2
## 16	Outer Comstock	990	404540.1
## 17	Park Ave.	942	455879.1
## 18	Prospect Hill	365	897674.4
## 19	Salt Springs	1414	431394.5
## 20	Sedgwick	1138	664994.5
## 21	Skunk City	713	380891.4
## 22	South Campus	36	647834.9
## 23	South Valley	1925	347411.5
## 24	Southside	1370	299747.6
## 25	Southwest	1150	253118.3
## 26	Strathmore	1822	545052.0
## 27	Tipp Hill	1468	651839.8
## 28	University Hill	505	1979056.2
## 29	University Neighborhood	1259	749673.8
## 30	Washington Square	1180	553236.6
## 31	Westcott	1540	675517.6
## 32	Winkworth	452	367041.8
## 33	<NA>	6	145493.0

3. Drill down to the downtown area for the following:

a. Create a map that highlights parking lots.

```
these.downtown <- syr$Nhood == "Downtown"
these.downtown[ is.na(these.downtown) ] <- F
downtown <- syr[ these.downtown , ]

dat.dt <- as.data.frame( downtown )
col.vec <- ifelse( dat.dt$LandUse == "Parking", "red", NA)

plot( downtown, col=col.vec, border="lightgray", main="Parking Lots in Downtown Syracuse")
```

Parking Lots in Downtown Syracuse



b. Create a map that highlights commercial areas.

```
col.vec <- ifelse( dat.dt$LandUse == "Commercial", "red", NA)  
  
plot( downtown, col=col.vec, border="lightgray", main="Commercial Areas in Downtown Syracuse")
```

Commercial Areas in Downtown Syracuse



c. Create a map that highlights residential areas.

```
col.vec <- rep( NA, nrow(dat.dt) )  
col.vec[ dat.dt$LandUse == "Apartment" ] <- "red"  
col.vec[ dat.dt$LandUse == "Multiple Residence" ] <- "red"  
col.vec[ dat.dt$LandUse == "Single Family" ] <- "red"  
col.vec[ dat.dt$LandUse == "Three Family" ] <- "red"  
col.vec[ dat.dt$LandUse == "Two Family" ] <- "red"  
  
plot( downtown, col=col.vec, border="lightgray", main="Residential Areas in Downtown Syracuse")
```

Residential Areas in Downtown Syracuse



d. Create a map that highlights non-taxable parcels.

```
col.vec <- rep( NA, nrow(dat.dt) )  
col.vec[ dat.dt$LandUse == "Parks" ] <- "red"  
col.vec[ dat.dt$LandUse == "Schools" ] <- "red"  
col.vec[ dat.dt$LandUse == "Religious" ] <- "red"
```

```
plot( downtown, col=col.vec, border="lightgray", main="Non-Taxable Parcels in Downtown Sy  
racuse")
```

Non-Taxable Parcels in Downtown Syracuse



e. What proportion of the downtown is residential? What proportion is commercial?

```
acreage.dt <- tapply(dat.dt$Acres, dat.dt$LandUse, sum, na.rm=T)
acre.df <- data.frame( Acres=as.numeric(acreage.dt), name=names(acreage.dt), stringsAsFactors=F )
```

```
#find residential rows
which(acre.df$name == "Apartment" )
```

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

```
which(acre.df$name == "Multiple Residence" )
```

```
## [1] 6
```

```
which(acre.df$name == "Single Family" )
```

```
## [1] 12
```



```
which(acre.df$name == "Two Family" )
```

```
## [1] 14
```

```
which(acre.df$name == "Three Family" )
```

```
## [1] 13
```

```
#create proportions
res.prop.dt <- sum(acre.df[c(1, 6, 12, 13, 14),1], na.rm=T) / sum(acre.df$Acres, na.rm=T)
com.prop.dt <- acre.df[acre.df$name == "Commercial",1] / sum(acre.df$Acres, na.rm=T)

#Proportion Residential
res.prop.dt
```

```
## [1] 0.06875218
```

```
#Proportion Commercial
com.prop.dt
```

```
## [1] 0.3726347
```

4. Which neighborhood has the highest number of vacant lots (var=LandUse)?

```
vacant.lot <- dat[dat$LandUse == "Vacant Land",]
vac.nhood <- as.data.frame(table(vacant.lot$Nhood))
most.vac <- which.max(vac.nhood$Freq)
vac.nhood[most.vac, ]
```

```
##           Var1 Freq
## 13 Near Westside  425
```

Vacant buildings (var=VacantBuil)?

```
vac.build <- dat[dat$VacantBuil == "Y",]
vac.nhood <- as.data.frame(table(vac.build$Nhood))
most.vac <- which.max(vac.nhood$Freq)
vac.nhood[most.vac, ]
```

```
##           Var1 Freq
## 15 Northside   264
```

Highest proportion of both combined?

```
vac.all <- dat[dat$VacantBuil == "Y" | dat$LandUse == "Vacant Land",]
vac.nhood <- as.data.frame(table(vac.all$Nhood))
prop.all <- as.data.frame(table(dat$Nhood))
dat3 <- merge(prop.all, vac.nhood, by.x="Var1", by.y="Var1")
dat3$prop.vac <- dat3$Freq.y/dat3$Freq.x
most.prop.vac <- which.max(dat3$prop.vac)
dat3[most.prop.vac, ]
```

```
##           Var1 Freq.x Freq.y  prop.vac
## 7 Franklin Square      89     37 0.4157303
```

5. Create a map of the city that highlights the age of buildings using the following categories:

< 1900, 1900-1919, 1920-1939, 1940-1959, 1960-1979, 1980-1999, 2000-2014

For your maps, be sure to use an informative title and legend.

```
dat$Build <- 2015 - as.numeric(dat$YearBuilt)
col.vals <- brewer.pal( 7, "BuGn" )
dat$col.vec2 <- rep( NA, nrow(dat) )
dat$col.vec2[ dat$Build < 1900 ] <- "#EDF8FB"
dat$col.vec2[ dat$Build > 1899 & dat$Build < 1920 ] <- "#CCECE6"
dat$col.vec2[ dat$Build > 1919 & dat$Build < 1940 ] <- "#99D8C9"
dat$col.vec2[ dat$Build > 1939 & dat$Build < 1960 ] <- "#66C2A4"
dat$col.vec2[ dat$Build > 1959 & dat$Build < 1980 ] <- "#41AE76"
dat$col.vec2[ dat$Build > 1979 & dat$Build < 2000 ] <- "#238B45"
dat$col.vec2[ dat$Build > 1999 ] <- "#005824"

plot(syr, col=dat$col.vec2, border=NA, main="Syracuse Buildings by Age")
legend("bottomleft", legend = c("< 1900", "1900-1919", "1920-1939", "1940-1959", "1960-1979", "1980-1999", "2000-2014"),
      title = "Year Built",
      fill = col.vals,
      cex = 0.56,
      bty = "n")
```

Syracuse Buildings by Age

