6. Basic Spatial Analysis

This helpsheet will explore a variety of basic spatial analysis techniques, including *clipping*, *point in polygon* and *buffering*.

Clipping

Clipping allows us to use one set of boundaries to cut another, a bit like using a cookie cutter.

```
# Load the Libaries
library(rgdal)
library(maptools)
library(rgeos)
library(stringr)
# Set working directory
setwd("M:/R work")
# Download data.zip from the web
download.file("http://data.alex-singleton.com/r-helpsheets/6/data.zip", "data.zip")
# Unzip file
unzip("data.zip")
```

```
# Read in both shape files
LSOA <- readOGR(".", "england_LSOA_2011")
outline <- readOGR(".", "England-outline")</pre>
```

First of all, we can plot the LSOA zones in Liverpool.

Plot the LSOA Map plot(LSOA)



We can also plot the England outline, but if we just run plot(outline) it will replace the LSOA plot on the display. To add the outline layer to the existing plot window, we can run the code below which will plot the

outline with a red border and we can also adjust the colour of the border to a red colour (border="red"), and the fill colour (col="#2C7FB820") a shade of blue. These represent two ways of specifying colours. The second contains eight alphanumerics, the first six relate to a HEX colour code. To view various colours that can be used in R, have a look at the website http://research.stowers-institute.org/efg/R/Color/Chart/ColorChart.pdf. The final two characters are the level of transparency (in this case 20%). Sometimes when running R in Windows, the transparency option will not work - it will just fill it with a solid colour. In this case, just remove the col = "#2C7FB820" section from the plot command to generate a red outline.

Overplot the outline map
plot(outline, add = TRUE, border = "red", col = "#2C7FB820")



As you will notice, the LSOA boundaries cross the River Mersey and stop at the river centre line. This doesn't look very nice, so we can tidy this up by getting R to clip the LSOA boundaries where they cross the England outline border.

We do this using the gIntersection command, passing it the two layer variables (outline and LSOA). We can also tell R we just want it to use the area covering the LSOAs by specifying byid = TRUE and id = my_area_id. Be aware that the gIntersection command may take up to 90 seconds to run - do not worry if your computer appears to freeze. Just wait for the command to complete.

```
# set the area we want to cut
my_area_id <- as.character(LSOA@data$ZONECODE)
# run the Intersection command, saving output to clipLSOA, this may take
# anywhere up to 90 seconds to run
clipLSOA <- gIntersection(LSOA, outline, byid = TRUE, id = my_area_id)
# replot the map as above to see what we have done
plot(clipLSOA)
plot(outline, add = TRUE, border = "red", col = "#2C7FB820")</pre>
```



We have now removed the parts of the LSOAs that overlap the coastline, and the map looks much more attractive.

Point in Polygon Analysis

Point in polygon analysis is useful when you want to create a subset of points from a larger set based on their spatial location. In this example we will load a list of locations that relate to all doctors surgeries in England, and use the polygons of ward boundaries in Leeds to create a subset of the Leeds doctors surgeries. To begin with, we need to load the libraries and get the GP and Wards data.

```
# Load the Libaries
library(maptools)
```

```
library(rgeos)
# Set working directory
setwd("M:/R work")
# Download data.zip from the web
download.file("http://data.alex-singleton.com/r-helpsheets/6/data.zip", "data.zip")
# Unzip file
unzip("data.zip")
```

```
# Read in shapefile
Wards <- readShapeSpatial("CAS-leeds", proj4string = CRS("+init=epsg:27700"))</pre>
```

It's worth having a quick look at the Leeds data so we know what it looks like:

Plot Wards to check it has been read in correctly
plot(Wards)



The doctors surgeries data is quite untidy - once we've read it in, we need to remove some extra columns that we don't need, and rename the ones we do.

```
# Get Data
GP <- read.csv("General Practices 2006.csv", header = TRUE, skip = 3)
# Extract the columns we want
GP <- subset(GP, select = c("Practice.Doctor.s.Name", "Easting", "Northing"))
# Rename the columns to something more helpful
colnames(GP) <- c("Surgery", "Easting", "Northing")
# Do a plot to check what the data look like
plot(GP$Easting, GP$Northing)</pre>
```



This should look like the above. The next stage is to convert the data into a SpatialPointsDataFrame.

The first line contains a subset command which removes any of the entries which have a blank value for Northings or Eastings. != means 'not equal to' and & means 'AND' so in "English" the command reads "overwrite the GP data frame with a subset of the GP data frame where the Easting field is not blank and the Northing field is not blank".

In a SpatialPointsDataFrame each entry must have a unique ID, so the second line creates an ID in the column GP_ID. The third line brings together the different elements to create the SpatialPointsDataFrame, GP-SP. GP[2] and GP[3] are the Easting and Northing columns respectively, and the data = section tells R which bits of the data frame to include. In this case we only want the surgery name (GP\$Surgery) and the ID number (GP\$GP_ID). The final term (proj4string) specifies which projection the data set is in - in this case, British National Grid (epsg:27700).

Show the results
plot(GP_SP)



This plot will look similar to the previous one, but the data are now stored in a SpatialPointsDataFrame. We now can calculate a point in polygon, i.e. to select those points which lie within the boundary of Leeds. The forth line below uses a <code>!is.na</code> command. <code>is.an</code> is a command to test whether a value is 'NA' and <code>!</code> means the inverse, so the command is testing whether the value (of GP_SP@data\$label) is not NA.

```
# point in polygon - returns a dataframe of the attributes of the polygons
# that the point is within.
o <- over(GP_SP, Wards)
# Many of these will be NA values - because most GPs are not in Leeds!
head(o)
# Add the attributes back onto the GP_SP SpatialPointsDataFrame (they are
# the same length)
GP_SP@data <- cbind(GP_SP@data, o)
# Use the NA values to remove those points not within Leeds
GP_SP_Leeds <- GP_SP[!is.na(GP_SP@data$label), ]
# Map your results
plot(GP_SP_Leeds)
```



We can also plot the points over the Leeds LSOAs:

plot(Wards) plot(GP_SP_Leeds, add = TRUE)



We can also view the data in the ${\tt GP_SP_Leeds}$ data frame.

```
# View the data slot of the results
head(GP_SP_Leeds@data)
```

```
GP.Surgery GP.GP_ID gid ons_label
                                                                 name
3591 Dr A M Marshall & Partners
                                     3591 1664
                                                  OODAFA Aireborough
3592
       Dr Adams R J & Partners
                                                  OODAFU Morley North
                                     3592 1662
3593
       Dr Adams R J & Partners
                                    3593 1662
                                                  OODAFU Morley North
3594
       Dr Adams R J & Partners
                                    3594 1662
                                                  OODAFU Morley North
3595
                     Dr Ali S A
                                    3595 3470
                                                  OODAFQ
                                                              Hunslet
3596
      Dr Allman I G & Partners
                                    3596 3430
                                                  OODAGG
                                                             Weetwood
      label
3591 08DAFA
3592 08DAFU
3593 08DAFU
3594 08DAFU
3595 08DAFQ
3596 08DAGG
```

Buffers

This section looks at buffers. It carries on from the section on points in polygon, so make sure you complete that section first.

Buffers are often used in spatial analysis for defining context of points. In this example we will calculate a buffer from the doctors surgeries of a 10 minute walking distance, based on an average of 3 mph, which is around 1608m.

The rgeos package has a function called gBuffer() that can be used to create buffers around points, lines or polygon objects. In the following example we create a new SpatialPolygons object called GP_SP_Leeds_Buffers. This then needs to be converted into a SpatialPolygonsDataFrame object by joining the @data from GP_SP_Leeds back onto GP_SP_Leeds_Buffers. Spatial Polygons objects do not have the data slot.

```
# buffers
GP_SP_Leeds_Buffers <- gBuffer(GP_SP_Leeds, width = 1608, byid = TRUE)
# Convert GP_SP_Leeds_Buffers into a SpatialPolygonsDataFrame (rather than
# SpatialPolygons) by joining the data of the GP_SP_Leeds
# SpatialPolygonsDataFrame
GP_SP_Leeds_Buffers <- SpatialPolygonsDataFrame(GP_SP_Leeds_Buffers, GP_SP_Leeds@data)</pre>
```

We can also now plot this on top of the Wards map.

```
# Wards wards
plot(Wards, axes = FALSE, col = "#6E7B8B", border = "#CAE1FF")
# GP locations
plot(GP_SP_Leeds, pch = 19, cex = 0.4, col = "#5CACEE", add = TRUE)
# catchment buffers
plot(GP_SP_Leeds_Buffers, axes = FALSE, col = NA, border = "red", add = TRUE)
```

