# hw11.R

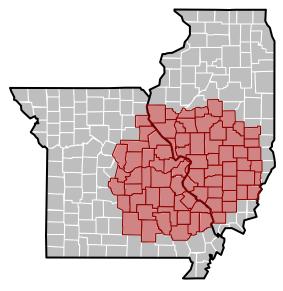
#### joebrew

Sat Nov 15 10:17:29 2014

```
# Homework 11: Disease Cluster Analysis -SaTscan
# PHC 6937: Spatial Epidemiology
# Stl_home is a shapefile which includes St Louis region county
# homicide counts (variable: HC8893) and population (variable: PO8893).
# In this assignment, you are requested to perform a purely spatial
# scan test based on Poisson model to detect any disease clusters
# in the study region.
# Steps: Import the data into SatScan and set up the session to run the analysis.
# Present the Output of the cluster information for evaluation
# (including the primary cluster and secondary cluster if necessary)
######
# Attach packages
######
library(gdata)
## gdata: read.xls support for 'XLS' (Excel 97-2004) files ENABLED.
##
## gdata: read.xls support for 'XLSX' (Excel 2007+) files ENABLED.
## Attaching package: 'gdata'
##
## The following object is masked from 'package:stats':
##
       nobs
##
##
## The following object is masked from 'package:utils':
##
##
       object.size
library(dplyr)
##
## Attaching package: 'dplyr'
## The following object is masked from 'package:stats':
##
##
       filter
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
```

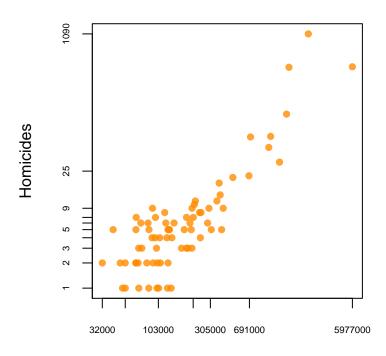
```
library(maptools)
## Loading required package: sp
## Checking rgeos availability: TRUE
library(rgdal)
## rgdal: version: 0.9-1, (SVN revision 518)
## Geospatial Data Abstraction Library extensions to R successfully loaded
## Loaded GDAL runtime: GDAL 1.10.1, released 2013/08/26
## Path to GDAL shared files: /usr/share/gdal/1.10
## Loaded PROJ.4 runtime: Rel. 4.8.0, 6 March 2012, [PJ_VERSION: 480]
## Path to PROJ.4 shared files: (autodetected)
library(spatstat)
##
## spatstat 1.38-1
                         (nickname: 'Le Hardi')
## For an introduction to spatstat, type 'beginner'
library(maps)
library(SpatialEpi)
######
# Set working directory to the haiti directories on local machine
if(Sys.info()["sysname"] == "Windows"){
 wd <- 'C:/Users/BrewJR/Documents/uf/phc6194/hw11'</pre>
  wd <- '/home/joebrew/Documents/uf/phc6194/hw11'</pre>
setwd(wd)
######
# Import the St Louis crime and population shapefile
######
stl <- readOGR("stl hom", "stl hom")</pre>
## OGR data source with driver: ESRI Shapefile
## Source: "stl_hom", layer: "stl_hom"
## with 78 features and 21 fields
## Feature type: wkbPolygon with 2 dimensions
######
# Visually check read-in and projection by mapping
# onto state of Missouri and Illinois
######
map("county", c("missouri", "illinois"),
   fill = TRUE,
  col = "grey",
```

```
border = "white")
map("state", c("missouri", "illinois"),
    col = "black",
    add = TRUE,
    lwd = 2)
plot(stl,
    add = TRUE,
    col = adjustcolor("red", alpha.f = 0.3),
    border = "darkred")
```



```
######
# Rename some variables for comprehensibility
######
stl$hom <- stl$HC8893
stl$pop <- stl$P08893
######
# Sanity check: population and homicide (absolute) should correlate
######
plot(log(stl$pop), log(stl$hom),
     xlab = "Population",
    ylab = "Homicides",
    xaxt = "n",
     yaxt = "n",
     main = "Population and homicides (log-log)",
     pch = 16,
     col = adjustcolor("darkorange", alpha.f = 0.8))
axis(side = 1,
     at = log(quantile(stl*pop, probs = c(0, 0.1, 0.4, 0.7, 0.8, 0.9, 1))),
     labels = round(quantile(stlpop, probs = c(0, 0.1, 0.4, 0.7, 0.8, 0.9, 1)), digits = -3),
     cex.axis = 0.6)
axis(side = 2,
     at = log(quantile(stl$hom, probs = seq(0,1,0.1))),
     labels = round(quantile(stl$hom, probs = seq(0,1,0.1))),
    cex.axis = 0.6)
```

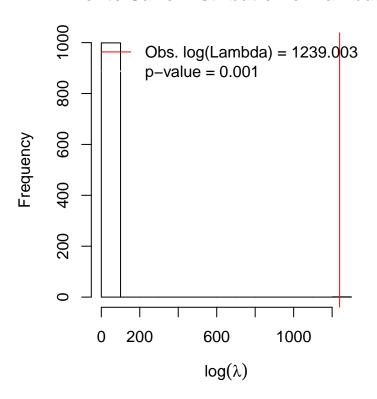
## Population and homicides (log-log)



## Population

```
######
# Use SpatialEpi's kulldorff function for clustering detection
# (spatial scan)
######
#my_geo <- matrix(coordinates(stl), ncol = 2)</pre>
my_geo <- latlong2grid(coordinates(stl))</pre>
my_cases <- stl$hom</pre>
my_population <- stl$pop</pre>
n_strata <- nrow(stl)</pre>
expected_cases <- (stl$pop / sum(stl$pop)) * sum(stl$hom) #expected(my_population, my_cases, n_strata)
## Kulldorff using Poisson likelihoods
k_poisson <- kulldorff(geo = my_geo,</pre>
                         cases = my_cases,
                        population = my_population,
                         expected.cases = expected_cases,
                        pop.upper.bound = 0.5,
                        n.simulations = 999,
                         alpha.level = 0.05,
                        plot = TRUE)
```

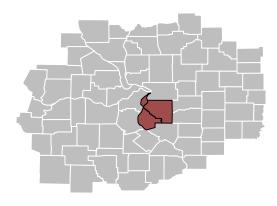
### Monte Carlo Distribution of Lambda



```
cluster <- k_poisson$most.likely.cluster$location.IDs.included

## plot
plot(stl, col = "grey", border = "white")
plot(stl[cluster,],add=TRUE,col=adjustcolor("darkred", alpha.f = 0.6))
title(main = "Most Likely Cluster after 999 Monte-Carlo simulations\n(no secondary clusters identified)</pre>
```

# Most Likely Cluster after 999 Monte–Carlo simulations (no secondary clusters identified)



```
######
# PRINT INFORMATION
######
summary(k_poisson)
                      Length Class Mode
## most.likely.cluster 8
                             -none- list
## secondary.clusters
                      O -none- NULL
## type
                        1 -none- character
## log.lkhd
                      2006 -none- numeric
## simulated.log.lkhd 999
                             -none- numeric
k_poisson$most.likely.cluster
## $location.IDs.included
## [1] 46 49 40
## $population
## [1] 4093508
## $number.of.cases
## [1] 1529
##
## $expected.cases
## [1] 409.2167
## $SMR
## [1] 3.736406
##
## $log.likelihood.ratio
## [1] 1239.003
## $monte.carlo.rank
## [1] 1
##
## $p.value
## [1] 0.001
#Cluster:
as.character(stl$NAME[cluster])
```

"St. Louis City"

## [1] "St. Clair"

"Monroe"