

Spatial Analysis: Columbus Crime

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Spatial Analysis

```
rm(list=ls())  
library(spdep)
```

```
## Loading required package: sp
```

```
## Loading required package: spData
```

```
## To access larger datasets in this package, install the spDataLarge  
## package with: `install.packages('spDataLarge',  
## repos='https://nowosad.github.io/drat/', type='source')`
```

```
## Loading required package: sf
```

```
## Linking to GEOS 3.9.1, GDAL 3.4.3, PROJ 7.2.1; sf_use_s2() is TRUE
```

```
library(spatialreg)
```

```
## Loading required package: Matrix
```

```
##  
## Attaching package: 'spatialreg'
```

```
## The following objects are masked from 'package:spdep':  
##  
##   get.ClusterOption, get.coresOption, get.mcOption,  
##   get.VerboseOption, get.ZeroPolicyOption, set.ClusterOption,  
##   set.coresOption, set.mcOption, set.VerboseOption,  
##   set.ZeroPolicyOption
```

```
require(maptools)
```

```
## Loading required package: maptools
```

```
## Checking rgeos availability: TRUE
## Please note that 'maptools' will be retired by the end of 2023,
## plan transition at your earliest convenience;
## some functionality will be moved to 'sp'.
```

```
require(rgdal)
```

```
## Loading required package: rgdal
```

```
## Please note that rgdal will be retired by the end of 2023,
## plan transition to sf/stars/terra functions using GDAL and PROJ
## at your earliest convenience.
##
## rgdal: version: 1.5-32, (SVN revision 1176)
## Geospatial Data Abstraction Library extensions to R successfully loaded
## Loaded GDAL runtime: GDAL 3.4.3, released 2022/04/22
## Path to GDAL shared files: C:/Users/maryp/AppData/Local/R/win-library/4.2/rgdal/gdal
## GDAL binary built with GEOS: TRUE
## Loaded PROJ runtime: Rel. 7.2.1, January 1st, 2021, [PJ_VERSION: 721]
## Path to PROJ shared files: C:/Users/maryp/AppData/Local/R/win-library/4.2/rgdal/proj
## PROJ CDN enabled: FALSE
## Linking to sp version:1.5-0
## To mute warnings of possible GDAL/OSR exportToProj4() degradation,
## use options("rgdal_show_exportToProj4_warnings"="none") before loading sp or rgdal.
```

Load data

Set work directory and plot data

```
setwd("C:\\Users\\maryp\\Downloads\\Spatial_Analysis_Columbus_Crime")

unzip("columbus.zip")
CC = readOGR(dsn = ".", layer = "columbus")
```

```
## OGR data source with driver: ESRI Shapefile
## Source: "C:\\Users\\maryp\\Downloads\\Spatial_Analysis_Columbus_Crime", layer: "columbus"
## with 49 features
## It has 20 fields
## Integer64 fields read as strings: COLUMBUS_ COLUMBUS_I POLYID
```

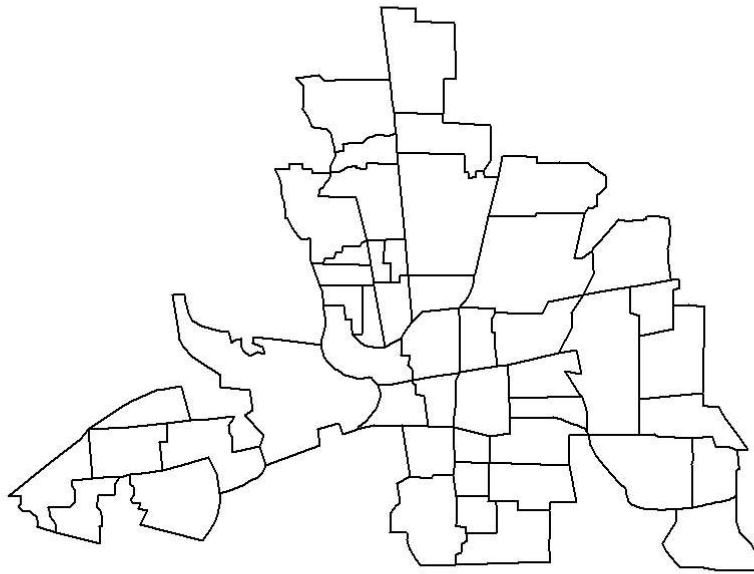
```
## check data
slotNames(CC)
```

```
## [1] "data"          "polygons"      "plotOrder"     "bbox"          "proj4string"
```

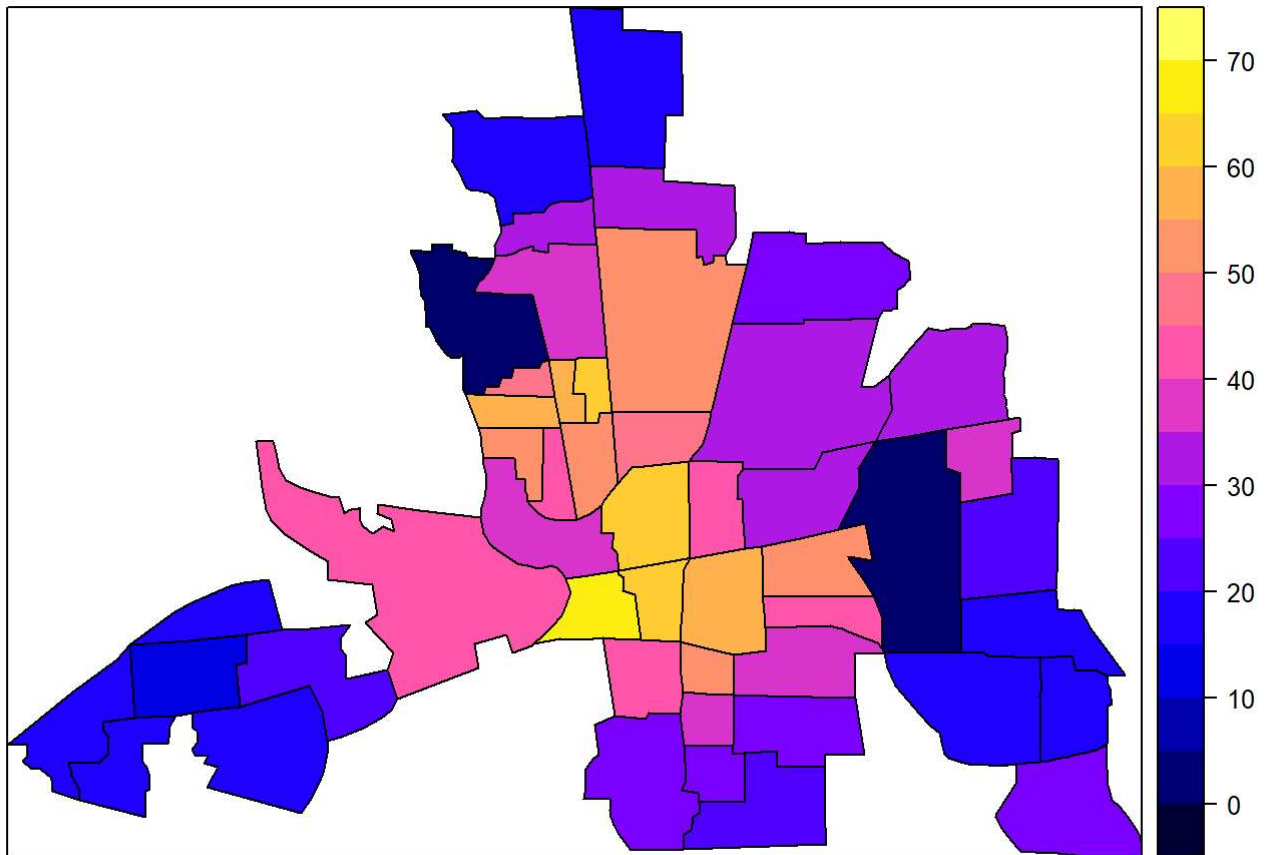
Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.

Plot spatial data

```
df = CC@data  
## SSPLIT  
plot(CC)
```



```
spplot(CC, "CRIME", pretty = TRUE)
```



Plot Columbus crime data

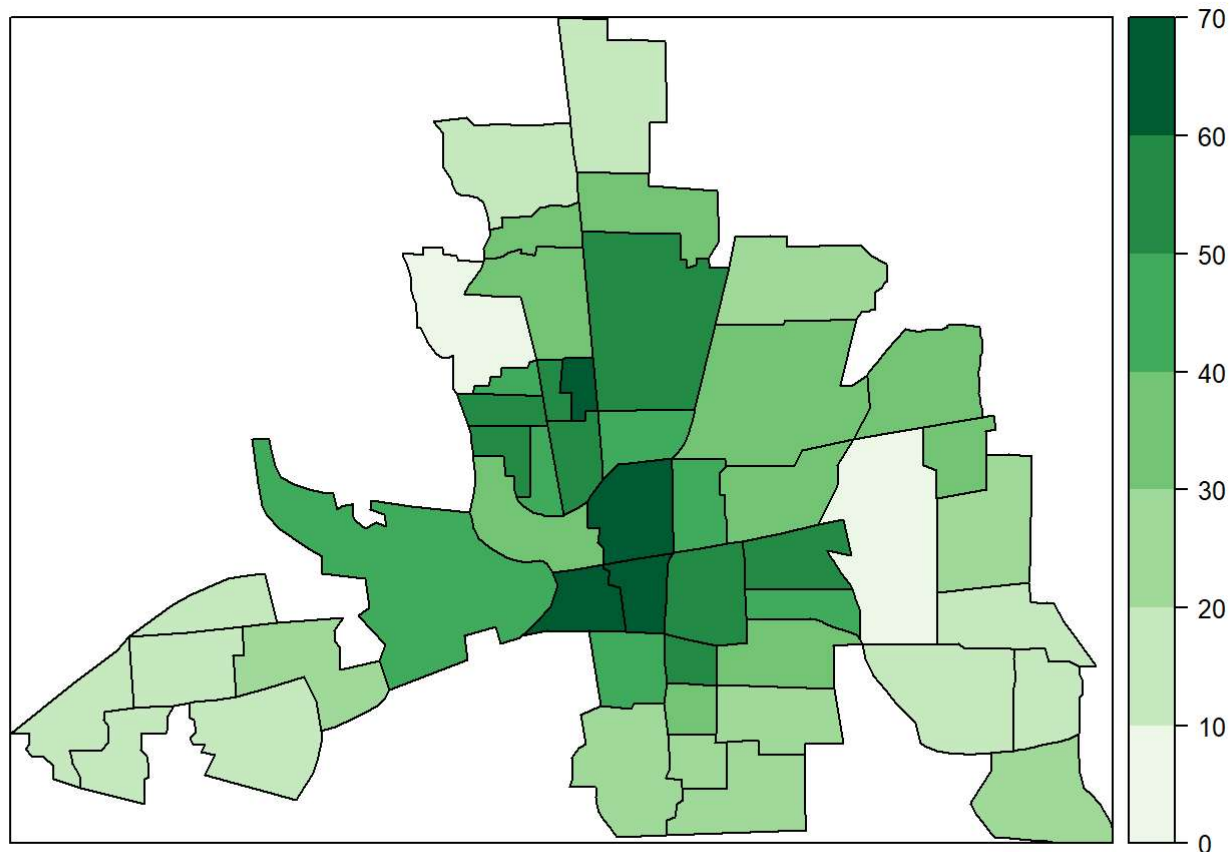
```
require(RColorBrewer)
```

```
## Loading required package: RColorBrewer
```

```
range(CC$CRIME)
```

```
## [1] 0.178269 68.892044
```

```
rng = seq(0, 70, 10)  
cls = brewer.pal(7, "Greens")  
spplot(CC, "CRIME", col.regions = cls, at = rng, sub = "Crime Data in Ohio")
```



Crime Data in Ohio

Plot neighbor coordinates for data

```
data(columbus)
help(columbus)
```

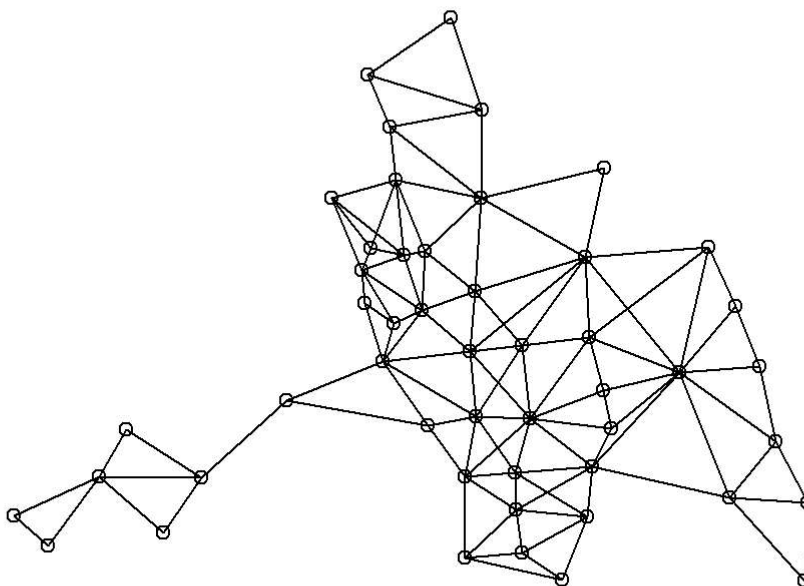
```
## starting httpd help server ... done
```

```
mydata <- columbus
attach(mydata)
Y <- cbind(CRIME)
X <- cbind(INC, HOVAL)
xy <- cbind(mydata$X, mydata$Y)
neighbors <- col.gal.nb
coords <- coords

# Neighbors summary
summary(neighbors)
```

```
## Neighbour list object:
## Number of regions: 49
## Number of nonzero links: 230
## Percentage nonzero weights: 9.579342
## Average number of links: 4.693878
## Link number distribution:
##
##  2  3  4  5  6  7  8  9 10
##  7  7 13  4  9  6  1  1  1
## 7 least connected regions:
## 1005 1008 1045 1047 1049 1048 1015 with 2 links
## 1 most connected region:
## 1017 with 10 links
```

```
plot(neighbors, coords)
```



```
##
```

Descriptive statistics

```
summary(Y)
```

```
##      CRIME
## Min.   : 0.1783
## 1st Qu.:20.0485
## Median :34.0008
## Mean   :35.1288
## 3rd Qu.:48.5855
## Max.   :68.8920
```

```
summary(X)
```

```
##      INC      HOVAL
## Min.   : 4.477   Min.   :17.90
## 1st Qu.: 9.963   1st Qu.:25.70
## Median :13.380   Median :33.50
## Mean   :14.375   Mean   :38.44
## 3rd Qu.:18.324   3rd Qu.:43.30
## Max.   :31.070   Max.   :96.40
```

OLS Regression

```
olsreg <- lm(Y ~ X)
summary(olsreg)
```

```
##
## Call:
## lm(formula = Y ~ X)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -34.418  -6.388  -1.580   9.052  28.649
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  68.6190     4.7355  14.490 < 2e-16 ***
## XINC         -1.5973     0.3341  -4.780 1.83e-05 ***
## XHOVAL       -0.2739     0.1032  -2.654  0.0109 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.43 on 46 degrees of freedom
## Multiple R-squared:  0.5524, Adjusted R-squared:  0.5329
## F-statistic: 28.39 on 2 and 46 DF, p-value: 9.341e-09
```

Spatial analysis based on contiguity

```
# Spatial weight matrix based on contiguity
```

```
listw <- nb2listw(neighbors)
summary(listw)
```

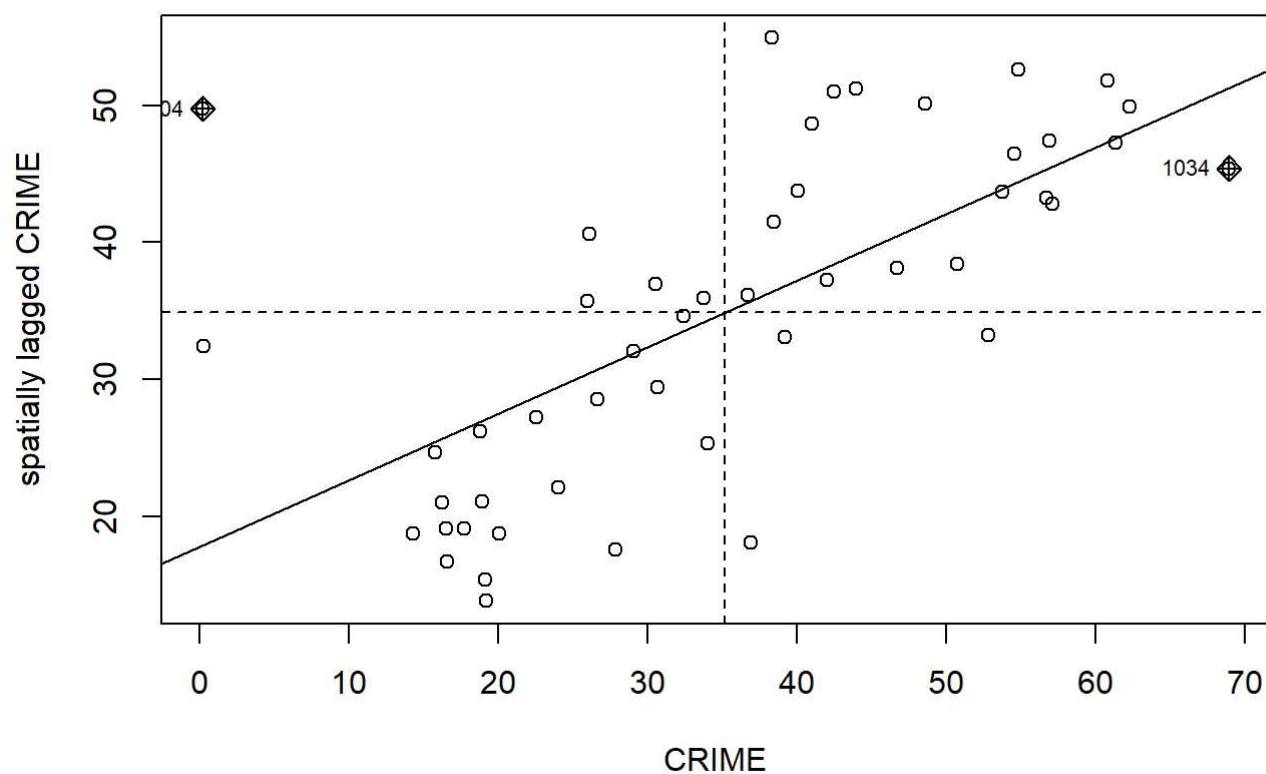
```
## Characteristics of weights list object:
## Neighbour list object:
## Number of regions: 49
## Number of nonzero links: 230
## Percentage nonzero weights: 9.579342
## Average number of links: 4.693878
## Link number distribution:
##
##  2  3  4  5  6  7  8  9 10
##  7  7 13  4  9  6  1  1  1
## 7 least connected regions:
## 1005 1008 1045 1047 1049 1048 1015 with 2 links
## 1 most connected region:
## 1017 with 10 links
##
## Weights style: W
## Weights constants summary:
##      n   nn S0      S1      S2
## W 49 2401 49 23.48489 204.6687
```

```
# Moran's I test
```

```
moran.test(CRIME, listw)
```

```
##
## Moran I test under randomisation
##
## data: CRIME
## weights: listw
##
## Moran I statistic standard deviate = 5.3427, p-value = 4.578e-08
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic      Expectation      Variance
##      0.485770914      -0.020833333      0.008991121
```

```
moran.plot(CRIME, listw)
```

```
# Lagrange multiplier test for spatial lag and spatial error dependencies
lm.LMtests(olsreg, listw, test=c("LMlag", "LMerr"))
```

```
##
##  Lagrange multiplier diagnostics for spatial dependence
##
## data:
## model: lm(formula = Y ~ X)
## weights: listw
##
## LMlag = 7.8557, df = 1, p-value = 0.005066
##
##
##  Lagrange multiplier diagnostics for spatial dependence
##
## data:
## model: lm(formula = Y ~ X)
## weights: listw
##
## LMerr = 4.6111, df = 1, p-value = 0.03177
```

```
# Spatial lag model
```

```
spatial.lag <- lagsarlm(CRIME ~ INC + HOVAL, data = mydata, listw)  
summary(spatial.lag)
```

```
##  
## Call:lagsarlm(formula = CRIME ~ INC + HOVAL, data = mydata, listw = listw)  
##  
## Residuals:  
##           Min           1Q           Median           3Q           Max  
## -37.4497093  -5.4565567   0.0016387   6.7159553  24.7107978  
##  
## Type: lag  
## Coefficients: (asymptotic standard errors)  
##           Estimate Std. Error z value Pr(>|z|)  
## (Intercept) 46.851431   7.314754  6.4051 1.503e-10  
## INC        -1.073533   0.310872 -3.4533 0.0005538  
## HOVAL      -0.269997   0.090128 -2.9957 0.0027381  
##  
## Rho: 0.40389, LR test value: 8.4179, p-value: 0.0037154  
## Asymptotic standard error: 0.12071  
##      z-value: 3.3459, p-value: 0.00082027  
## Wald statistic: 11.195, p-value: 0.00082027  
##  
## Log likelihood: -183.1683 for lag model  
## ML residual variance (sigma squared): 99.164, (sigma: 9.9581)  
## Number of observations: 49  
## Number of parameters estimated: 5  
## AIC: 376.34, (AIC for lm: 382.75)  
## LM test for residual autocorrelation  
## test value: 0.19184, p-value: 0.66139
```

```
# Spatial error model
```

```
spatial.error <- errorsarlm(CRIME ~ INC + HOVAL, data = mydata, listw)  
summary(spatial.error)
```

```
##
## Call:errorsarlm(formula = CRIME ~ INC + HOVAL, data = mydata, listw = listw)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -34.45950  -6.21730  -0.69775   7.65256  24.23631
##
## Type: error
## Coefficients: (asymptotic standard errors)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  61.053618   5.314875  11.4873 < 2.2e-16
## INC          -0.995473   0.337025  -2.9537 0.0031398
## HOVAL        -0.307979   0.092584  -3.3265 0.0008794
##
## Lambda: 0.52089, LR test value: 6.4441, p-value: 0.011132
## Asymptotic standard error: 0.14129
##      z-value: 3.6868, p-value: 0.00022713
## Wald statistic: 13.592, p-value: 0.00022713
##
## Log likelihood: -184.1552 for error model
## ML residual variance (sigma squared): 99.98, (sigma: 9.999)
## Number of observations: 49
## Number of parameters estimated: 5
## AIC: 378.31, (AIC for lm: 382.75)
```

Spatial analysis based on distance weight matrix

```
# Spatial weight matrix based on distance (with lower and upper bounds for distance, d1 and d2)

nb <- dnearneigh(xy, d1=0, d2=10)
listw <- nb2listw(nb, style="W")
summary(listw)
```

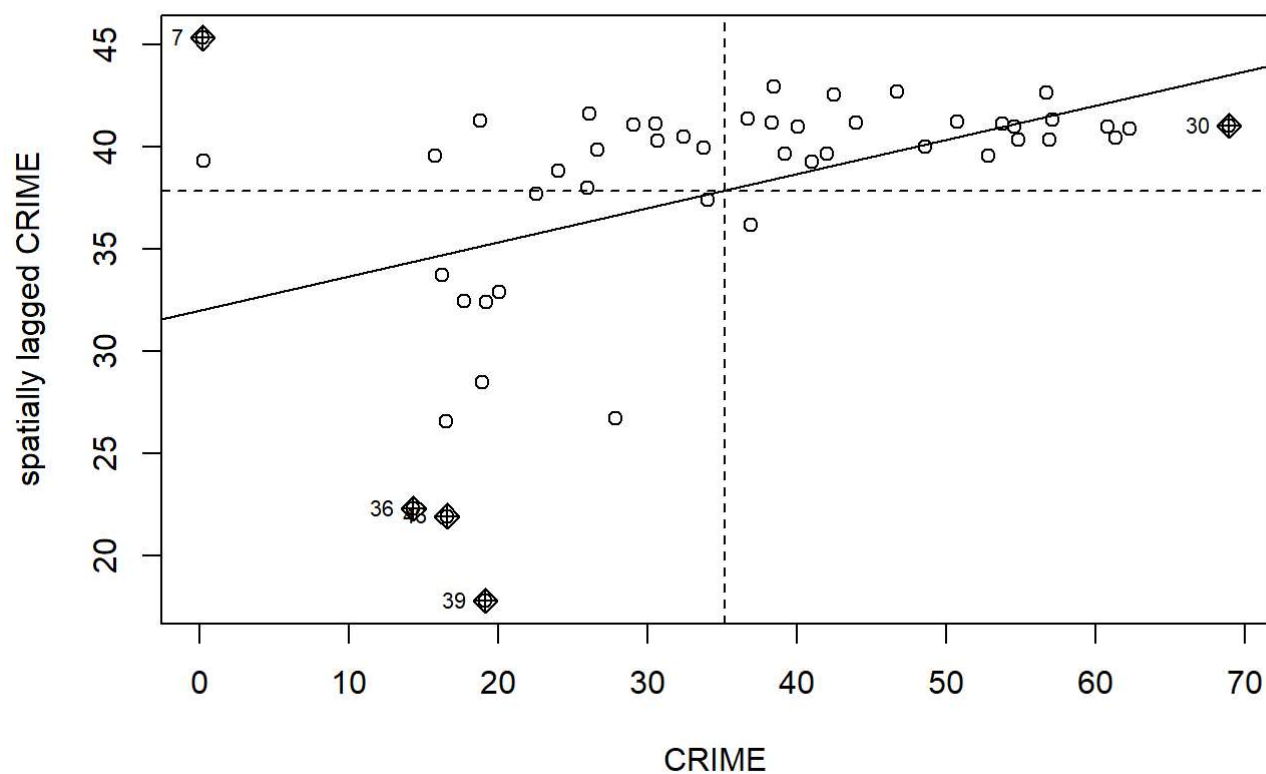
```
## Characteristics of weights list object:
## Neighbour list object:
## Number of regions: 49
## Number of nonzero links: 1234
## Percentage nonzero weights: 51.39525
## Average number of links: 25.18367
## Link number distribution:
##
## 5 6 8 10 12 13 14 15 16 18 19 20 21 24 26 28 29 30 31 32 33 34 35 36 38
## 1 2 1 1 1 1 1 1 1 1 2 2 4 3 1 2 2 2 3 3 2 6 2 2 2
## 1 least connected region:
## 39 with 5 links
## 2 most connected regions:
## 22 26 with 38 links
##
## Weights style: W
## Weights constants summary:
##      n   nn S0      S1      S2
## W 49 2401 49 4.763862 199.3227
```

```
# Moran's I test
```

```
moran.test(CRIME, listw)
```

```
##
## Moran I test under randomisation
##
## data: CRIME
## weights: listw
##
## Moran I statistic standard deviate = 5.6185, p-value = 9.629e-09
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic      Expectation      Variance
##      0.167361950      -0.020833333      0.001121945
```

```
moran.plot(CRIME, listw)
```



Lagrange multiplier test for spatial lag and spatial error dependencies

```
lm.LMtests(olsreg, listw, test=c("LMlag", "LMerr"))
```

```
##
##  Lagrange multiplier diagnostics for spatial dependence
##
## data:
## model: lm(formula = Y ~ X)
## weights: listw
##
## LMlag = 2.8736, df = 1, p-value = 0.09004
##
##
##  Lagrange multiplier diagnostics for spatial dependence
##
## data:
## model: lm(formula = Y ~ X)
## weights: listw
##
## LMerr = 0.12837, df = 1, p-value = 0.7201
```

```
# Spatial lag model
```

```
spatial.lag1 <- lagsarlm(CRIME ~ INC + HOVAL, data = mydata, listw)  
summary(spatial.lag1)
```

```
##  
## Call:lagsarlm(formula = CRIME ~ INC + HOVAL, data = mydata, listw = listw)  
##  
## Residuals:  
##      Min      1Q   Median      3Q      Max   
## -35.68762  -7.01341  -0.83503   8.40003  27.44614   
##  
## Type: lag  
## Coefficients: (asymptotic standard errors)  
##              Estimate Std. Error z value Pr(>|z|)      
## (Intercept) 51.627459  11.817848  4.3686 1.250e-05      
## INC         -1.384654   0.334827 -4.1354 3.543e-05      
## HOVAL       -0.281171   0.098167 -2.8642 0.004181      
##  
## Rho: 0.37541, LR test value: 1.9423, p-value: 0.16342  
## Asymptotic standard error: 0.26881  
##      z-value: 1.3966, p-value: 0.16254  
## Wald statistic: 1.9504, p-value: 0.16254  
##  
## Log likelihood: -186.4061 for lag model  
## ML residual variance (sigma squared): 117.06, (sigma: 10.82)  
## Number of observations: 49  
## Number of parameters estimated: 5  
## AIC: 382.81, (AIC for lm: 382.75)  
## LM test for residual autocorrelation  
## test value: 1.4922, p-value: 0.22187
```

```
# Spatial error model
```

```
spatial.error1 <- errorsarlm(CRIME ~ INC + HOVAL, data = mydata, listw)  
summary(spatial.error1)
```

```
##
## Call:errorsarlm(formula = CRIME ~ INC + HOVAL, data = mydata, listw = listw)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -34.85406  -6.71581  -0.46122   8.91963  28.71276
##
## Type: error
## Coefficients: (asymptotic standard errors)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  67.311097   4.836926  13.9161 < 2.2e-16
## INC          -1.538356   0.330112  -4.6601 3.161e-06
## HOVAL        -0.268478   0.098726  -2.7194 0.006539
##
## Lambda: 0.18231, LR test value: 0.14824, p-value: 0.70022
## Asymptotic standard error: 0.4012
##      z-value: 0.45442, p-value: 0.64953
## Wald statistic: 0.20649, p-value: 0.64953
##
## Log likelihood: -187.3031 for error model
## ML residual variance (sigma squared): 122.18, (sigma: 11.054)
## Number of observations: 49
## Number of parameters estimated: 5
## AIC: 384.61, (AIC for lm: 382.75)
```