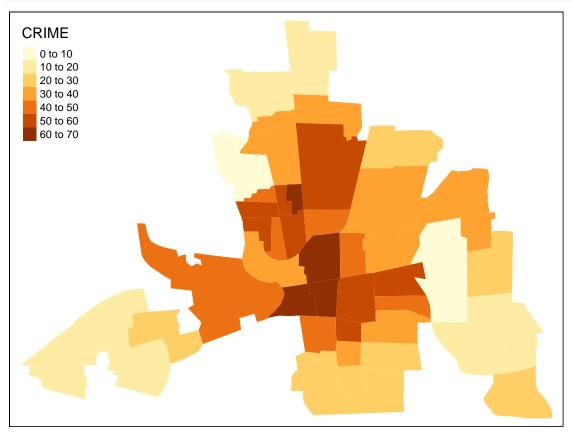
R: Interoperability

Roger Bivand

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Use R for comparison

```
library(sf)
## Linking to GEOS 3.7.2, GDAL 3.0.1, PROJ 6.2.0
col_file <- system.file("shapes/columbus.shp", package="spData")[1]</pre>
col_sf <- st_read(col_file, quiet=TRUE)</pre>
names(col_sf)
   [1] "AREA"
                      "PERIMETER"
                                    "COLUMBUS_"
                                                 "COLUMBUS_I" "POLYID"
##
                                                 "CRIME"
##
   [6] "NEIG"
                      "HOVAL"
                                    "INC"
                                                               "OPEN"
## [11] "PLUMB"
                      "DISCBD"
                                    "X"
                                                 "Y"
                                                               "NSA"
                                    "CP"
## [16] "NSB"
                      "EW"
                                                 "THOUS"
                                                               "NEIGNO"
## [21] "geometry"
library(tmap)
tm_shape(col_sf) + tm_fill("CRIME", style="pretty")
```



Create spatial weights

```
library(spdep)

## Loading required package: sp

## Loading required package: spData

nb_sf <- poly2nb(col_sf)
nb_sf

## Neighbour list object:
## Number of regions: 49

## Number of nonzero links: 236

## Percentage nonzero weights: 9.829238
## Average number of links: 4.816327

lw <- nb2listw(nb_sf, style="W")</pre>
```

ESDA

```
moran.test(col_sf$CRIME, lw)
##
## Moran I test under randomisation
##
## data: col_sf$CRIME
## weights: lw
##
## Moran I statistic standard deviate = 5.5894, p-value = 1.139e-08
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic
                                                 Variance
                           Expectation
         0.500188557
                          -0.020833333
                                              0.008689289
##
loc_I_sf <- localmoran(col_sf$CRIME, lw)</pre>
sum(loc_I_sf[,"Ii"])/Szero(lw)
## [1] 0.5001886
col_sf$Ii <- loc_I_sf[,"Ii"]</pre>
tm_shape(col_sf) + tm_fill("Ii", midpoint=0, style="pretty")
```



Spatial regression

```
library(spatialreg)
ev <- eigenw(lw)
err <- errorsarlm(CRIME ~ INC + HOVAL, data=col_sf, listw=lw, method="eigen", control=list(pre_eig=ev))
summary(err)
##
## Call:errorsarlm(formula = CRIME ~ INC + HOVAL, data = col_sf, listw = lw,
##
      method = "eigen", control = list(pre_eig = ev))
##
## Residuals:
##
        Min
                 1Q
                       Median
                                    ЗQ
                                            Max
## -34.65998 -6.16943 -0.70623
                               7.75392 23.43878
##
## Type: error
## Coefficients: (asymptotic standard errors)
              Estimate Std. Error z value Pr(>|z|)
## (Intercept) 60.279469 5.365594 11.2344 < 2.2e-16
## INC
             ## HOVAL
             ## Lambda: 0.54675, LR test value: 7.2556, p-value: 0.0070679
## Asymptotic standard error: 0.13805
      z-value: 3.9605, p-value: 7.4786e-05
```

```
## Wald statistic: 15.686, p-value: 7.4786e-05
##
## Log likelihood: -183.7494 for error model
## ML residual variance (sigma squared): 97.674, (sigma: 9.883)
## Number of observations: 49
## Number of parameters estimated: 5
## AIC: 377.5, (AIC for lm: 382.75)
Impacts
lag <- lagsarlm(CRIME ~ INC + HOVAL, data=col sf, listw=lw, method="eigen", control=list(pre eig=ev))
summary(impacts(lag, R=2000, evalues=ev), short=TRUE, zstats=TRUE)
## Impact measures (lag, evalues):
##
           Direct Indirect
                                Total
## INC
      -1.1008955 -0.7176834 -1.8185788
## HOVAL -0.2795832 -0.1822627 -0.4618459
## Simulation results (asymptotic variance matrix):
## -----
## Simulated standard errors
           Direct Indirect
##
                              Total
## INC 0.31314711 0.3894641 0.5783900
## HOVAL 0.09558312 0.1284327 0.1997202
## Simulated z-values:
          Direct Indirect
## INC -3.553192 -1.951256 -3.237635
## HOVAL -2.904591 -1.536780 -2.378340
## Simulated p-values:
       Direct
                  Indirect Total
## INC 0.00038059 0.051027 0.0012052
## HOVAL 0.00367733 0.124347 0.0173908
Write GAL file
td <- tempdir()</pre>
tf <- file.path(td, "col queen.gal")
write.nb.gal(nb_sf, tf)
Using reticulate to run Python from R in an R markdown notebook
```

```
library(reticulate)
use_python(python='/usr/bin/python3')
py_config()

## python: /usr/bin/python3
## libpython: /usr/lib64/libpython3.7m.so
## pythonhome: /usr:/usr
```

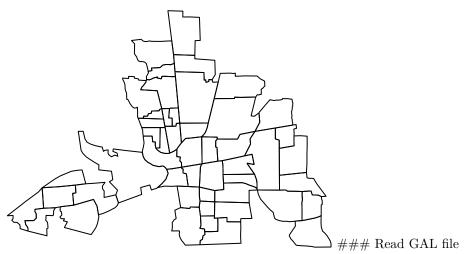
```
3.7.4 (default, Jul 9 2019, 16:48:28) [GCC 8.3.1 20190223 (Red Hat 8.3.1-2)]
## version:
## numpy:
                    /usr/local/lib64/python3.7/site-packages/numpy
## numpy_version: 1.16.0
## python versions found:
## /usr/bin/python3
## /usr/bin/python
pkgr <- import("pkg resources")</pre>
np <- import("numpy")</pre>
pkgr$get_distribution("numpy")$version
## [1] "1.16.0"
libpysal <- import("libpysal")</pre>
pkgr$get_distribution("libpysal")$version
## [1] "4.0.1"
gpd <- import("geopandas")</pre>
pkgr$get_distribution("geopandas")$version
## [1] "0.4.0"
col_ps <- gpd$read_file(col_file)</pre>
```

Passing shapely geometries to sf

```
shapely <- import("shapely")
pkgr$get_distribution("shapely")$version

## [1] "1.6.4.post2"

oo <- st_as_sf(data.frame(unlist(lapply(col_ps$geometry, shapely$wkt$dumps))), wkt=1)
plot(st_geometry(oo))</pre>
```



nb_ps <- libpysal\$weights\$Queen\$from_dataframe(col_ps)
nb_gal_ps <- libpysal\$io\$open(tf)\$read()</pre>

ESDA

```
esda <- import("esda")
pkgr$get_distribution("esda")$version

## [1] "2.0.0"
y <- np$array(col_ps[,"CRIME"])
mi <- esda$Moran(y, nb_ps, two_tailed=FALSE)
mi$I

## [1] 0.5001886

mi <- esda$Moran(y, nb_gal_ps, two_tailed=FALSE)
mi$I

## [1] 0.5001886

loc_I_ps <- esda$Moran_Local(y, nb_ps)
col_ps["Is"] <- loc_I_ps$Is</pre>
```

Spatial regression

```
spreg <- import("spreg")
pkgr$get_distribution("spreg")$version

## [1] "1.0.4"

x <- np$array(col_ps[, c("INC", "HOVAL")])
y <- matrix(y, ncol=1)
mlerr_ps <- spreg$ML_Error(y, x, nb_ps)</pre>
```

Comparison

```
rbind(R=coefficients(err)[c(2:4,1)], PySAL=c(mlerr_ps$betas))

## (Intercept) INC HOVAL lambda

## R 60.27947 -0.9573053 -0.3045593 0.5467531

## PySAL 60.27947 -0.9573053 -0.3045593 0.5467530
```

Python directly in R markdown

Read shapefile

```
import numpy as np
import libpysal as libpysal
import geopandas as gpd
col_ps = gpd.read_file('/home/rsb/lib/r_libs/spData/shapes/columbus.shp')
```

Create weights

```
nb_ps = libpysal.weights.Queen.from_dataframe(col_ps)
nb_ps.cardinalities
## {0: 2, 1: 3, 2: 4, 3: 4, 4: 8, 5: 2, 6: 4, 7: 6, 8: 8, 9: 4, 10: 5, 11: 6, 12: 4, 13: 6, 14: 6, 15: 5
ESDA
import esda as esda
```

mi = esda.Moran(col_ps[['CRIME']].values, nb_ps, two_tailed='false')

0.5001885571828611

[0.54675303]])

Spatial regression

mi.I

##

```
import spreg as spreg
mlerr_ps = spreg.ML_Error(col_ps[['CRIME']].values, col_ps[['INC', 'HOVAL']].values, nb_ps)

## /usr/local/lib64/python3.7/site-packages/scipy/optimize/_minimize.py:761: RuntimeWarning: Method 'bo'
## "defaulting to absolute tolerance.", RuntimeWarning)
mlerr_ps.betas

## array([[60.2794697],
## [-0.95730534],
## [-0.30455926],
```