

Spatial Statistics – Global Measures of Spatial Autocorrelation

The global autocorrelation measures calculated below use 2012 presidential election vote share data gathered from 1,055 counties in the Midwest.

1. Join counts

```
> joincount.multi(midwest_sf$obama_dummy, nb2listw(midwest_queen, style = "B"))
```

	Joincount	Expected	Variance	z-value
0:0	2211.000	2053.737	251.996	9.9067
1:1	264.000	102.540	77.417	18.3505
1:0	602.000	920.723	384.040	-16.2639
Jtot	602.000	920.723	384.040	-16.2639

```
> joincount.mc(midwest_sf$obama_dummy, nb2listw(midwest_queen, style = "B"), nsim = 999)
```

Monte-Carlo simulation of join-count statistic

```
data: midwest_sf$obama_dummy
weights: nb2listw(midwest_queen, style = "B")
number of simulations + 1: 1000
```

Join-count statistic for 0 = 2211, rank of observed statistic = 1000, p-value = 0.001

alternative hypothesis: greater

sample estimates:

mean of simulation	variance of simulation
2054.5896	244.6771

Monte-Carlo simulation of join-count statistic

```
data: midwest_sf$obama_dummy
weights: nb2listw(midwest_queen, style = "B")
number of simulations + 1: 1000
```

Join-count statistic for 1 = 264, rank of observed statistic = 1000, p-value = 0.001

alternative hypothesis: greater

sample estimates:

mean of simulation	variance of simulation
102.4575	81.7815

## 2. Moran's I

```
> moran.test(midwest_sf$diff, nb2listw(midwest_queen))

Moran I test under randomisation

data: midwest_sf$diff
weights: nb2listw(midwest_queen)

Moran I statistic standard deviate = 33.657, p-value < 0.0000000
0000000022
alternative hypothesis: greater
sample estimates:
Moran I statistic      Expectation      Variance
      0.6179604836      -0.0009487666      0.0003381504

> moran.mc(midwest_sf$diff, nb2listw(midwest_queen), nsim = 999)

Monte-Carlo simulation of Moran I

data: midwest_sf$diff
weights: nb2listw(midwest_queen)
number of simulations + 1: 1000

statistic = 0.61796, observed rank = 1000, p-value = 0.001
alternative hypothesis: greater
```

## 3. Geary's C

```
> geary.test(midwest_sf$diff, nb2listw(midwest_queen))

Geary C test under randomisation

data: midwest_sf$diff
weights: nb2listw(midwest_queen)

Geary C statistic standard deviate = 32.476, p-value < 0.0000000
0000000022
alternative hypothesis: Expectation greater than statistic
sample estimates:
Geary C statistic      Expectation      Variance
      0.3774116574      1.0000000000      0.0003675191

> geary.mc(midwest_sf$diff, nb2listw(midwest_queen), nsim = 999)

Monte-Carlo simulation of Geary C

data: midwest_sf$diff
weights: nb2listw(midwest_queen)
number of simulations + 1: 1000

statistic = 0.37741, observed rank = 1, p-value = 0.001
alternative hypothesis: greater
```

4. The join-counts, Moran's I, and Geary's C test results all indicate that there is evidence of positive autocorrelation in the vote data. The Monte-Carlo simulations of the joint-count statistic suggest that there are far more "1:1" (Obama-Obama) and "0:0" (Romney-Romney) counts than one would expect to see under a pattern of spatial randomness. Moran's I is positive and significant, providing further evidence of clustering. Finally, Geary's C is less than one and significant, which is also a sign that similar vote shares are spatially close together.

```

# Load libraries -----
#make sure the below libraries are installed on your machine. Use
install.packages("LIBRARYNAME")
rm(list = ls()) #clear objects in memory
# Set working directory
setwd(paste0(here::here(), "/08_Global_Measures"))
# Load additional libraries
pacman::p_load(tidyverse, # for basic data manipulation, visualization
scales, # for formatting number output
patchwork, # for arranging ggplots in grids
sf, # simple features for spatial
ngeo, # nearest neighbors
summarytools, # for checking data frame characteristics
here, # for relative file paths
knitr, # for tables
kableExtra, # table styling
janitor) # for cleaning and tabulations
#load your spatial libraries
library(sp)
library(raster)
library(spdep)
options(stringsAsFactors = FALSE, scipen = 999)
# *****
# START OF ASSIGNMENT
# *****
# Assignment (Due April 6, 2020) -----
# Read in RDS
midwest <- readRDS("Raw_Data/Midwest_2012_election.RDS")
# Convert to simple features
midwest_sf <- st_as_sf(midwest, crs = 4269)
# Define neighbors
# Queen
midwest_queen <- poly2nb(midwest_sf)
# # Rook
# midwest_rook <- poly2nb(midwest_sf, queen = FALSE)
#
# # Identify the difference
# midwest_diff <- diffnb(midwest_queen, midwest_rook)
#
# # Get the centroids of the counties
# midwest_cent <- coordinates(midwest)
#
# #plot the three neighbor definitions
# par(mfrow = c(1, 3))
# plot(midwest, border = 8)
# plot(midwest_queen, midwest_cent, add = TRUE)
# title(main = paste("Midwest Counties Queen"))
#
# plot(midwest, border = 8)
# plot(midwest_rook, midwest_cent, add = TRUE)
# title(main = paste("Midwest Counties Rook"))
#
# plot(midwest, border = 8)
# plot(midwest_diff, midwest_cent, add = TRUE, col = 2)
# title(main = paste("Midwest Counties Difference"))
# Create dummy variable

```

```

midwest_sf <- midwest_sf %>%
mutate(obama_dummy = ifelse(WINNER == "Obama", 1, 0) %>%
as_factor())
# Compute global autocorrelation measures using the Midwest data
#1. Use join count based on a variable that takes 1 when Obamawon and 0
otherwise
joincount.multi(midwest_sf$obama_dummy, nb2listw(midwest_queen, style = "B"))
joincount.mc(midwest_sf$obama_dummy, nb2listw(midwest_queen, style = "B"),
nsim = 999)
#2. Use Moran's I
moran.test(midwest_sf$diff, nb2listw(midwest_queen))
moran.mc(midwest_sf$diff, nb2listw(midwest_queen), nsim = 999)
#3. Use Geary's C
geary.test(midwest_sf$diff, nb2listw(midwest_queen))
geary.mc(midwest_sf$diff, nb2listw(midwest_queen), nsim = 999)
#4. Interpret your results

```