

# R Module 9

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## Contents

<b>Data Setup</b>	<b>2</b>
<b>Analysis</b>	<b>3</b>
1. Moran's I for Numeric Variables in NC . . . . .	3
2. Moran's I for different Neighbor Rules . . . . .	4
3. Spatial Correlogram Plots . . . . .	5
4. LISA . . . . .	6

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## Data Setup

```
# Reproject the shapefile into UTM, a projected CRS
NC_utm <- st_transform(NC, st_crs(26917))
# Find centroids of each polygon
centroids <- NC_utm %>%
  st_geometry() %>%
  st_centroid()

# Calculate neighbors using Queen's rule
NC_queen <- poly2nb(pl = NC_utm,
  row.names = NC_utm$NAME,
  queen = TRUE)
NC_rook <- poly2nb(pl = NC_utm,
  row.names = NC_utm$NAME,
  queen = FALSE)
NC_k4 <- knn2nb(knearneigh(centroids, k = 4))
# Calculate max distance for dnearneigh
max.dist <- knn2nb(knearneigh(centroids, k = 1)) %>%
  nbdists(coords = centroids) %>%
  unlist() %>% max()
NC_d100 <- dnearneigh(x = centroids, d1 = 0, d2 = max.dist)
# Convert to weights list object
NC_queen_listw <- nb2listw(NC_queen)
NC_rook_listw <- nb2listw(NC_rook)
NC_k4_listw <- nb2listw(NC_k4)
NC_d100_listw <- nb2listw(NC_d100)
```

## Analysis

```
moran_MNEM2000 <- moran.test(  
  x = NC_utm$MNEM2000,  
  listw = NC_queen_listw,  
  alternative = "two.sided",  
  zero.policy = TRUE  
)
```

Measuring Moran's I for Manufacturing Jobs gives a statistic of **0.3884704** and a p-value of  $3.2387706 \times 10^{-10}$ , indicating that there is somewhat strong positive spatial autocorrelation.

### 1. Moran's I for Numeric Variables in NC

The variables in this analysis are MNEM1990, MNEM2000, TOTJOB1990, and TOTJOB2000.

```
vars <- list(MNEM1990 = "MNEM1990",  
            MNEM2000 = "MNEM2000",  
            TOTJOB1990 = "TOTJOB1990",  
            TOTJOB2000 = "TOTJOB2000")  
moran.list <- lapply(vars, function(x) {  
  data <- NC_utm %>% dplyr::pull(x)  
  moran.test(  
    x = data,  
    listw = NC_queen_listw,  
    zero.policy = T,  
    alternative = 'two.sided'  
  )  
})
```

```
list.statistics <- data.frame(matrix(ncol = 0, nrow = 4))  
  
list.statistics <- list.statistics %>%  
  mutate(  
    Rule = names(vars),  
    `Moran's I` = sapply(moran.list, function(x)  
      {x$estimate["Moran I statistic"]}),  
    `p-value` = sapply(moran.list, function(x)  
      {x$p.value}  
    )  
  )
```

Rule	Moran's I	p-value
MNEM1990	0.4283699	0.0000000
MNEM2000	0.3884704	0.0000000
TOTJOB1990	0.1642235	0.0042649
TOTJOB2000	0.1631958	0.0035879

## 2. Moran's I for different Neighbor Rules

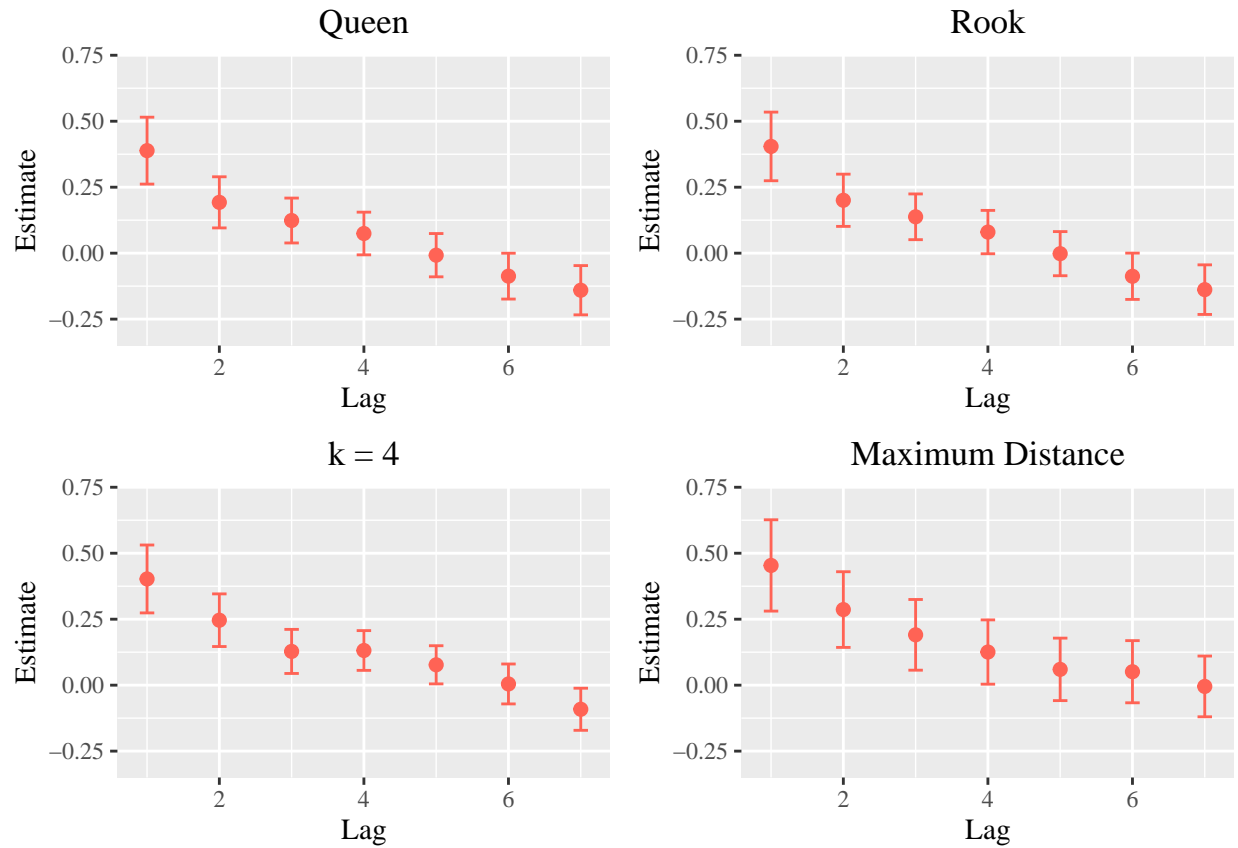
```
w_list <- list(
  Queen = NC_queen_listw,
  Rook = NC_rook_listw,
  `k = 4` = NC_k4_listw,
  `Maximum Distance` = NC_d100_listw
)
moran_diff_w <- lapply(w_list, function(w) {
  moran.test(
    x = NC_utm$MNEM2000,
    listw = w,
    alternative = "two.sided",
    zero.policy = TRUE
  )
})

diff_w_stats <- data.frame(matrix(ncol = 0, nrow = 4))
diff_w_stats <- diff_w_stats %>%
  mutate(
    Rule = names(w_list),
    `Moran's I` = sapply(moran_diff_w, function(x) {
      x$estimate["Moran I statistic"]
    }),
    `p-value` = sapply(moran_diff_w, function(x) {
      x$p.value
    }),
    `Avg. Neighbors` = sapply(w_list, function(x) {
      y <- x$neighbours
      len <- sapply(y, length)
      mean(len)
    })
  )
```

Rule	Moran's I	p-value	Avg. Neighbors
Queen	0.3884704	0e+00	4.90
Rook	0.4044841	0e+00	4.62
k = 4	0.4024844	0e+00	4.00
Maximum Distance	0.4536758	1e-07	2.80

I found it interesting that the Moran's I for the Rook and k = 4 cases were so similar; this might be because Rook searches for neighbors in four directions.

### 3. Spatial Correlogram Plots



The general spatial pattern with MNEM2000 is that counties become less similar the further out, but around order  $\sim 4$ , there is a small “bump”, showing that clusters have a significant distance between clusters. Interestingly, the estimate drops below 0 for most of the models, indicating there is a slight negative spatial autocorrelation at large scales (lags) – this could represent clusters of manufacturing jobs being surrounded by areas with far fewer jobs, and shows how industrial centers “clump” together. From a business standpoint, it would not make sense to build factories very close together, but rather provide access to a greater geographic area.

## 4. LISA

```
localMoran.queen <- localmoran(NC_utm$MNEM2000, NC_queen_listw)

sig <- c(.5, .1, .05, .01)

NC_utm <- NC_utm %>%
  mutate(Significance = localMoran.queen[,5],
         Quadrant_name = attr(localMoran.queen, "quadr")$mean,
         Quadrant = data.table::fifelse((Significance > sig),
                                       "Insignificant",
                                       Quadrant_name %>% as.character()))
```

