



# ECON6027 6a

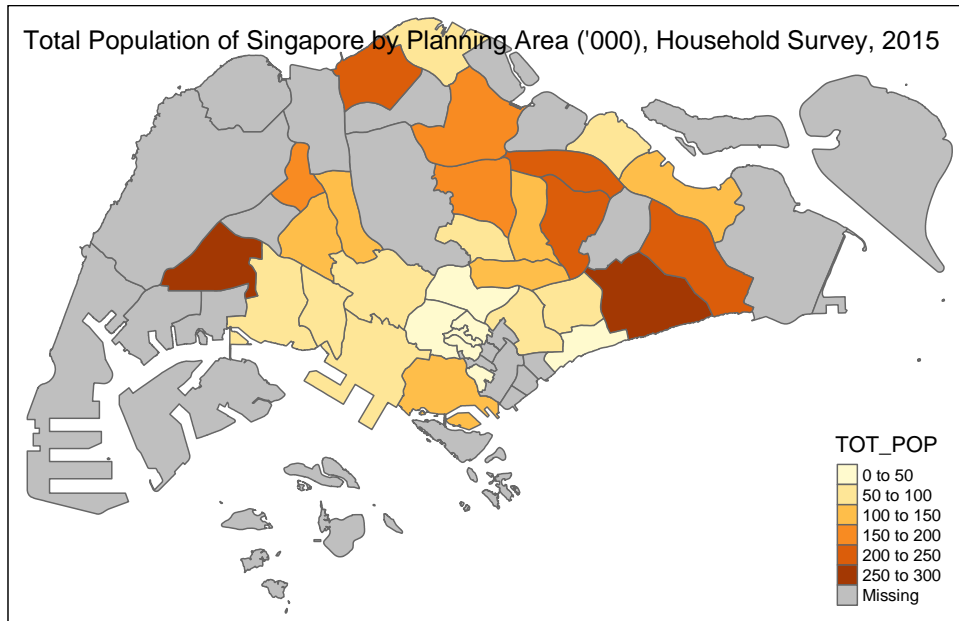
Areal units and neighbours



# Required R packages

- sf
- spdep

# Areal/lattice/regional data



- These are spatial data where the **domain D** is “**fixed and discrete**” (non-random and countable).
  - Eg: postal codes, GRCs, planning areas, remotely sensed data reported by pixels (such as data coming from satellites).
- Spatial locations with areal data are often referred to as “**sites**” or “**areal units**”.
- One of the main differences between point data and areal data is that, in practice areal data are **spatially aggregated** over areal regions. (Mathematically this refers to an integration of a continuous spatial attribute).
  - yield measures on an agricultural plot
  - event counts (such as deaths, crimes, voter turnout, etc.) for various sites (such as postal codes, regions, states, etc.)
- Spatial aggregation is becoming increasingly common due to the growing need to confidentiality and privacy of data records.

# Areal/lattice/regional data

- If areal units are **irregular**, a more precise term would be "regional data".
- Given the discrete nature of the collection of sites, areal data can be **exhaustive** (another differentiating feature compared to point data or geospatial data).
  - For example voter turnout data provide the number for every electoral unit and the issue of predicting the number for any other are does not arise.



# Spatial autocorrelation and neighbours

- Spatial relationships are best modelled based on the principle of spatial neighbours.
- We assume that the influence of spatial neighbours among  $n$  spatial units can be quantified using a ***spatial weight***.



# Neighbours...

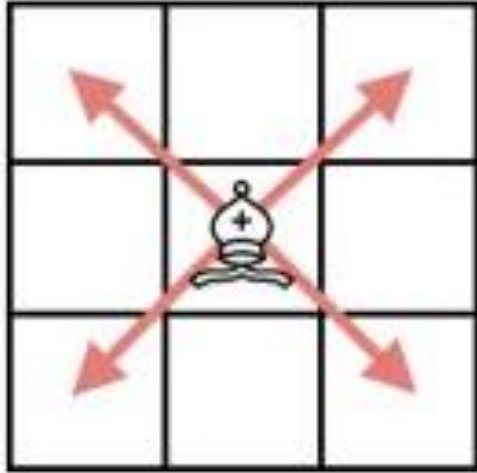
- Once you have divided the complete domain into the areal units, next step is to generate a matrix (n X n) that indicates how each areal unit is related to one another.
- This is the so-called "**spatial weights**" matrix (aka connectivity matrix).
- Two areal units may be neighbours of each other based on
  - Distance (geographic, economic or social distance)
  - Nearest neighbour
  - Contiguity (sharing borders)
  - Or a combination

$$W_n = \begin{bmatrix} w_{11} & \dots & w_{n1} \\ \dots & w_{ij} & \\ w_{1n} & & w_{nn} \end{bmatrix}$$

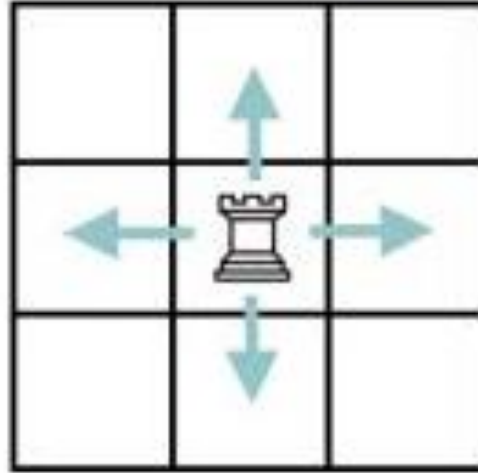
$$\text{where } w_{ij} = \begin{cases} 1 & \text{if } j \in N(i) \\ 0 & \text{o/w} \end{cases}$$



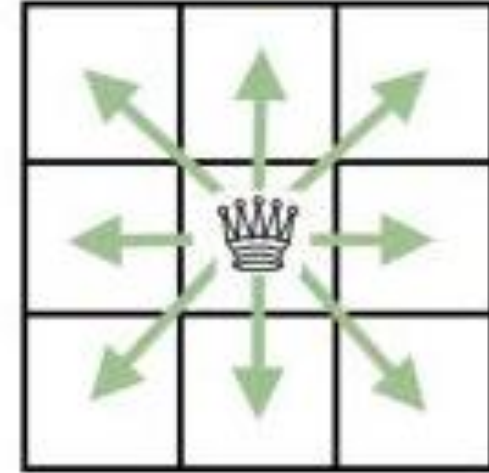
Bishop Contiguity



Rook Contiguity



Queen Contiguity



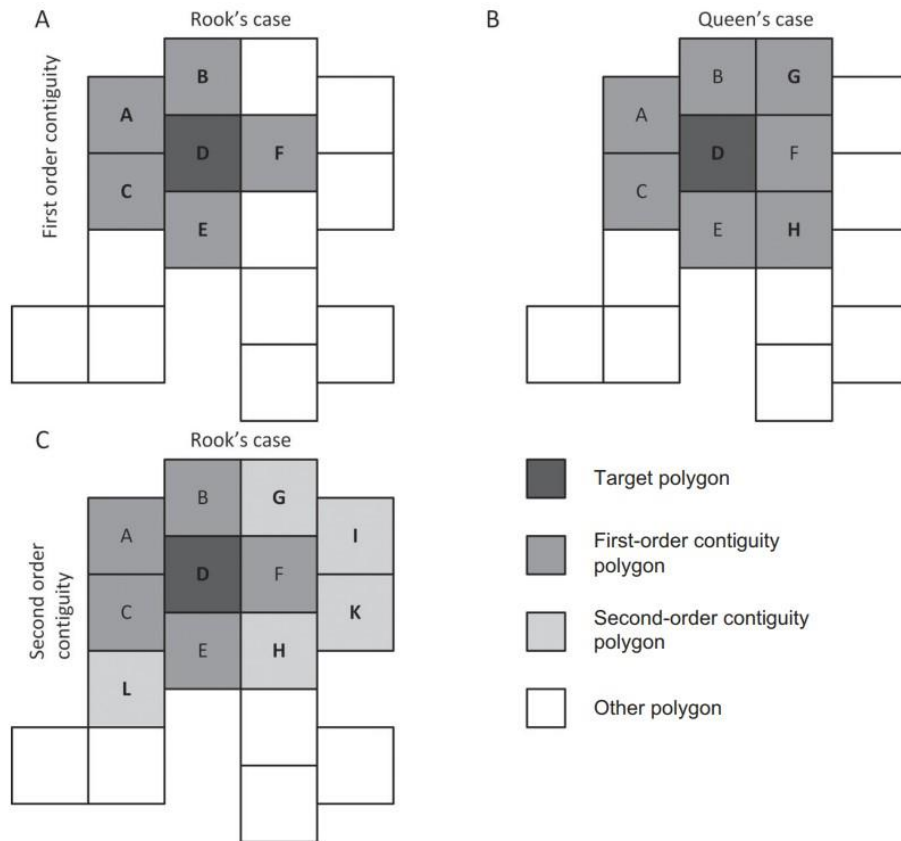
# Contiguity criteria on a “regular” lattice

If data are observed on a regular rectangular lattice, the contiguity neighbours can be defined using,

- Rook criterion (cardinal neighbours)
- Queen criterion (cardinal and ordinal neighbours)
- Bishop criterion (less popular)
- Circular
- Group interaction

# Contiguity criteria on regional data

However, in practice we must deal with irregularly spaced areal units such as planning areas, regions, countries etc.







# W matrix example

- See uk.xlsx for an example of a W matrix for UK regions based on contiguity.
- The fact that northern Ireland doesn't have any contiguity poses a (computational) problem since the  $\text{rank}(W) < n$ .
- The given W matrix is a symmetric matrix which means neighbours are mutual, however, this need not be the case.
- It is normal for weights matrices to be row standardised in applications (especially in spatial econometrics) as:

$$w_{ij}^* = \frac{w_{ij}}{\sum_j w_{ij}}$$

- The implication of row standardisation is that,

$$\sum_i \sum_j w_{ij}^* = n$$

Where n is the number of areal units.

# GAL file

- The portable form of a weights matrix that indicate information on neighbouring areal units are *usually* (but not necessarily) saved as a GAL file.
- A .gal file is a file produced by the software GeoDa.
- A .gal file is useful since it allows flexibility in the creation and the updating of neighbours, however, may not be feasible for large n.
- For small datasets, we can create a .gal file from scratch ourselves without having the need to use GeoDa.

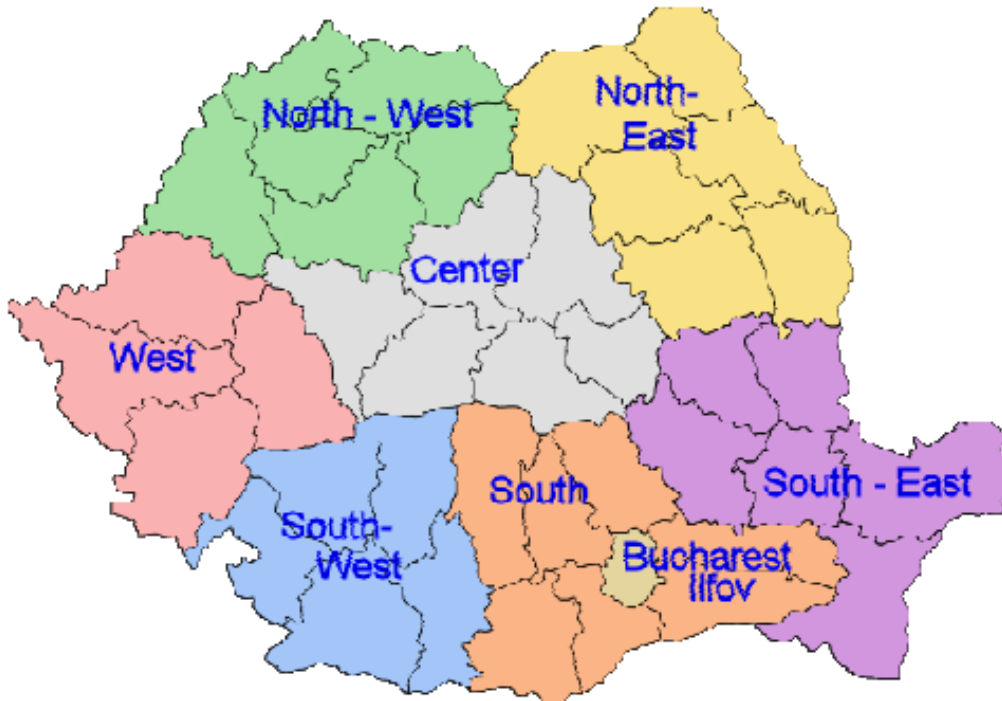
# Create your own GAL file (without GeoDa)

- You can save the information on neighbours and save this text file with the extension .GAL.
- A .GAL file must have the following format:
  - Header: line 1, starts with a mandatory 0 and the number of areal units (and optional column name)
  - A region identifier will be assigned to each areal unit from 1 to n.
  - Line 2: the region identifier of region 1 followed by the number of neighbouring regions of regions 1.
  - Line 3: region identifiers of the neighbours of region 1.
  - And so on...
- You need to **leave an empty line at the end.**

# GAL file example for regions of Romania

The figure shows the boundaries of the 8 Romanian NUTS2 regions.

1. North-west
2. Centre
3. North-east
4. South-east
5. South
6. Bucharest
7. South-west
8. West





Copy and paste the following data in a text file and save as a \*.GAL file (you need to leave an empty line at the end.).

Instructions on how to change the file extension to \*.gal:

<https://www.wikihow.com/Change-a-File-Extension>

**(Windows is preferred)**

0 8 romania rom\_regions

1 3

3 2 8

2 6

1 3 4 5 7 8

3 3

1 2 4

4 3

2 3 5

5 4

2 4 6 7

6 1

5

7 3

2 5 8

8 3

1 2 7

# Activity A

Create a GAL file for the 8 South Asian countries.

- Two countries are to be considered close neighbours if they share a common land border or if the shortest distance separated by water is less than 100km.
- Use the indexing found here:  
[https://en.wikipedia.org/wiki/South\\_Asia](https://en.wikipedia.org/wiki/South_Asia)

# Read neighbours from an existing GAL file

```
> (uk_nb = read.gal("uk_cont.gal"))
```

Neighbour list object:

- Number of regions: 12
- Number of nonzero links: 42
- Percentage nonzero weights: 29.16667
- Average number of links: 3.5

# R class “nb”

```
> class(UK_nb)
```

```
[1] "nb"
```

nb is a neighbour list object which is how neighbourhood information is handled in the R ecosystem.

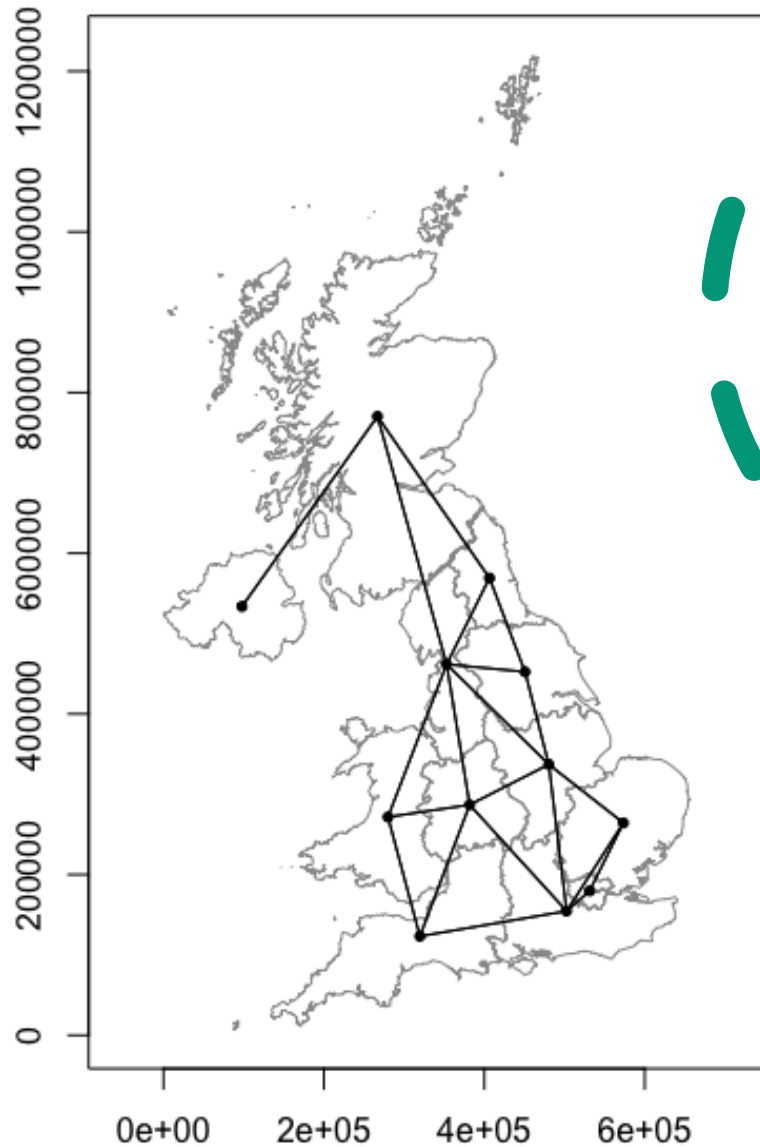


# We also need a shapefile to anchor the external “nb” object to...

```
> (uk_boundaries =  
st_read("NUTS_Level_1__January_2018__Boundaries.  
shp"))
```

- Make sure the areal units are in the same order as the “nb” object.
- For your convenience, in this case, I have made sure that the object ID of the shapefile and the gal file are the same.

nb from GAL file



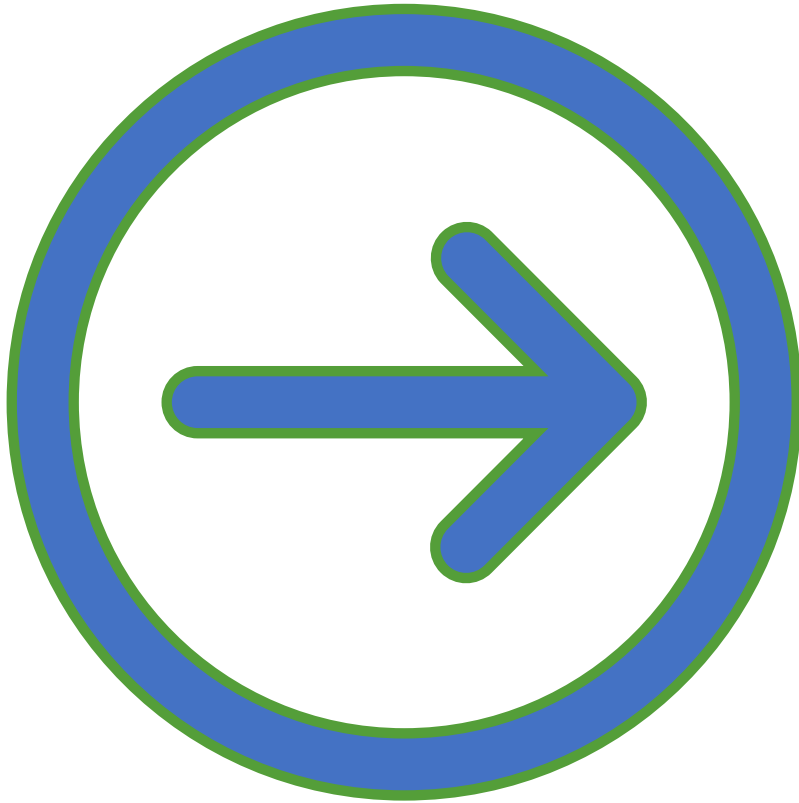
## Plot nb

```
> UK_coords =  
st_centroid(st_geometry(UK_  
boundaries))
```

```
>
```

```
plot(st_geometry(UK_boundar  
ies), border="grey60",  
axes=T, reset = F, main="nb  
from GAL file")
```

```
> plot(UK_nb, UK_coords,  
pch=19, cex=0.6, add=T)
```



# Create GAL file from nb

```
> write.nb.gal(UK_nb,  
"UK_nb_new.gal", oldstyle=F)
```

oldstyle: if TRUE, first line of file contains only number of spatial units, if FALSE (recommended), uses newer GeoDa style.



# Create neighbours based on spatial polygons

We will work with Singapore planning areas

```
> SG = st_read("MySingapura.shp")
```



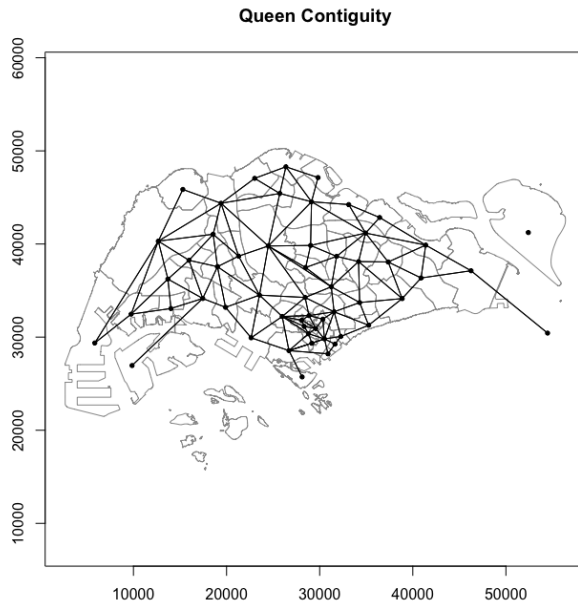
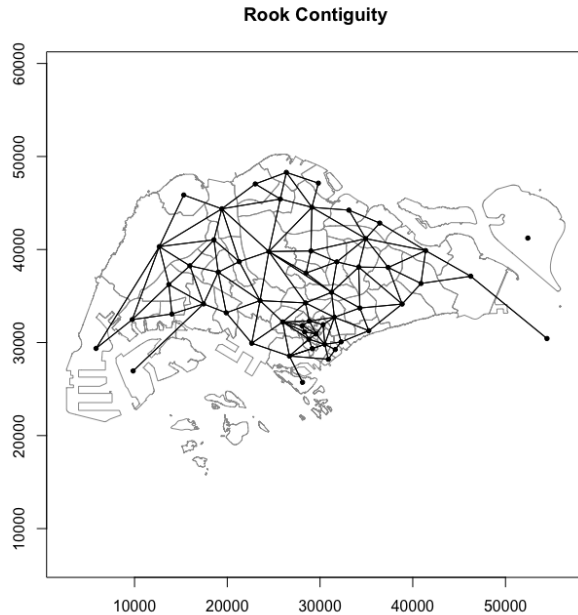
# 1. Creating contiguity neighbours from spatial polygons

Create **queen contiguity** neighbours from SG.

```
> (SG1_nb = poly2nb(SG) )
```

In “poly2nb”, queen=T by default, if we make queen=F, then we get the **rook contiguity**.

```
> (SG2_nb = poly2nb(SG, queen=F) )
```



# Plot...

```
SG_coords =  
st_point_on_surface(st_geometry(SG))
```

```
plot(st_geometry(SG), border="grey60", axes=T,  
main="Queen Contiguity")
```

```
plot(SG1_nb, SG_coords, pch=19, cex=0.6, add=T)
```

```
plot(st_geometry(SG), border="grey60", axes=T,  
main="Rook Contiguity")
```

```
plot(SG2_nb, SG_coords, pch=19, cex=0.6, add=T)
```



No  
difference

...

---

```
> isTRUE(all.equal(SG1_nb,  
SG2_nb, check.attributes =  
F))
```

---

```
[1] TRUE
```

---

For Singapore planning areas  
queen contiguity is the same as  
rook contiguity.

## 2. “k” Nearest neighbours

```
> IDs = row.names(SG)
```

k-nearest neighbours

```
SG3_nb =  
knn2nb(knearneigh(SG_coords, k =  
1), row.names = IDs)
```


```
SG4_nb =  
knn2nb(knearneigh(SG_coords, k =  
2), row.names = IDs)
```

```
SG5_nb =  
knn2nb(knearneigh(SG_coords, k =  
3), row.names = IDs)
```





Plot...



```
plot(st_geometry(SG), border="grey60",  
axes=T, main="k=1")
```

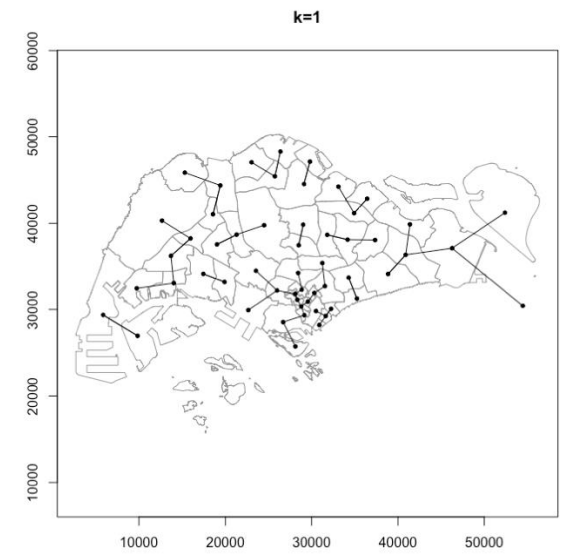
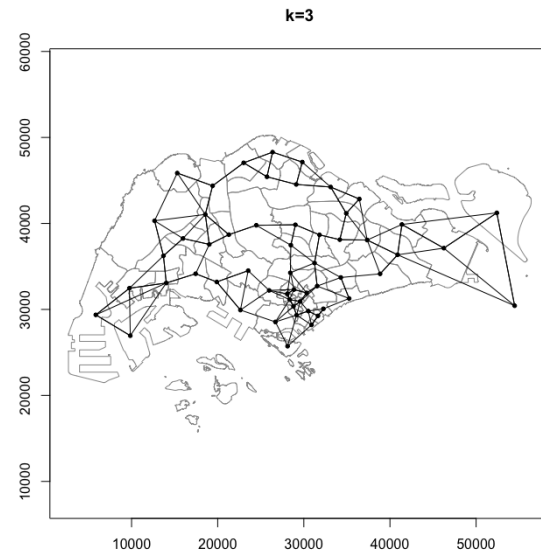
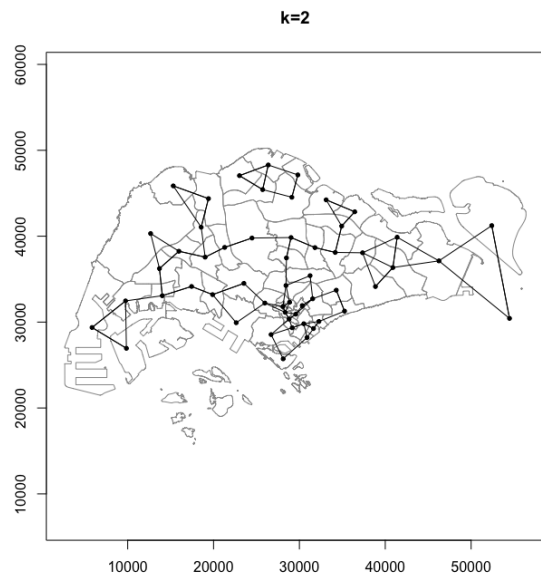
```
plot(SG3_nb, SG_coords, pch=19, cex=0.6,  
add=T)
```

```
plot(st_geometry(SG), border="grey60",  
axes=T, main="k=2")
```

```
plot(SG4_nb, SG_coords, pch=19, cex=0.6,  
add=T)
```

```
plot(st_geometry(SG), border="grey60",  
axes=T, main="k=3")
```

```
plot(SG5_nb, SG_coords, pch=19, cex=0.6,  
add=T)
```



### 3. Neighbours within a certain metric distance


```
> st_crs(SG)$units  
[1] "m"
```

```
SG6_nb = dnearneigh(SG_coords, d1 =  
0, d2 = 1000, row.names = IDs)
```

```
SG7_nb = dnearneigh(SG_coords, d1 =  
0, d2 = 5000, row.names = IDs)
```

```
SG8_nb = dnearneigh(SG_coords, d1 =  
0, d2 = 10000, row.names = IDs)
```

Note that distance-based neighbours can leave some spatial units neighbour-less and the degree of connectedness can exponentially increase as you increase the distance.



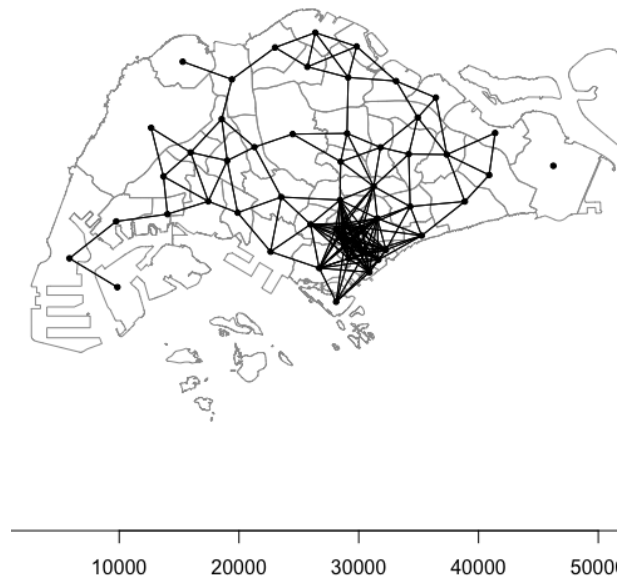
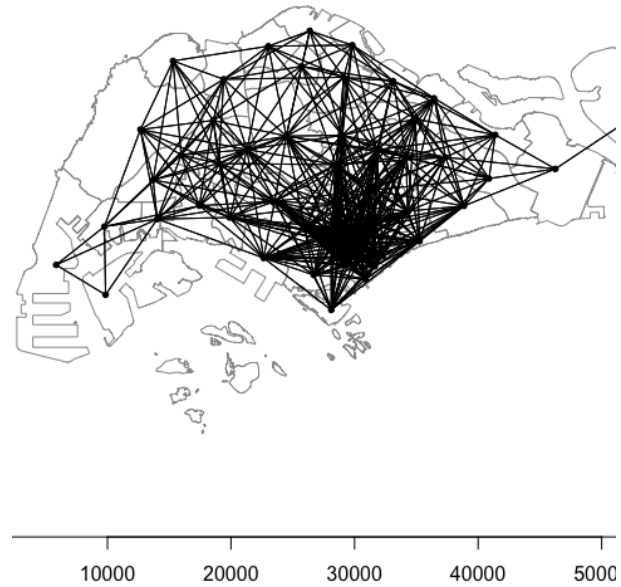
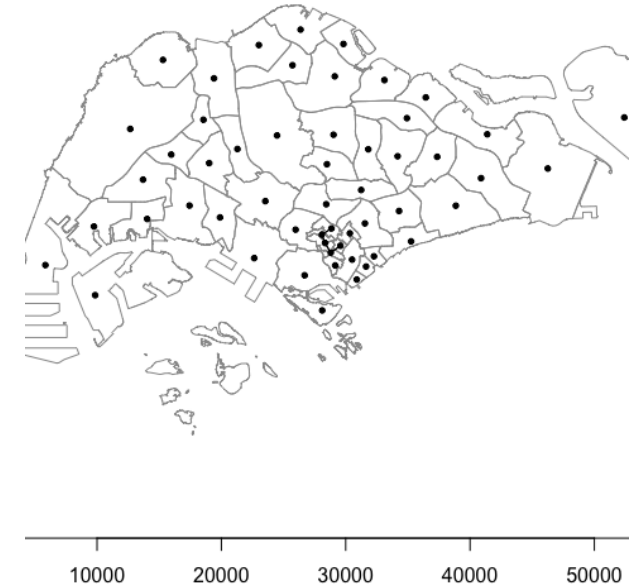
```
plot(st_geometry(SG), border="grey60",  
axes=T, main="Within 1km")  
plot(SG6_nb, SG_coords, pch=19, cex=0.6,  
add=T)
```

```
plot(st_geometry(SG), border="grey60",  
axes=T, main="Within 5km")  
plot(SG7_nb, SG_coords, pch=19, cex=0.6,  
add=T)
```

```
plot(st_geometry(SG), border="grey60",  
axes=T, main="Within 10km")  
plot(SG8_nb, SG_coords, pch=19, cex=0.6,  
add=T)
```



Plot...

**Within 5km****Within 10km****Within 1km**



## Activity B

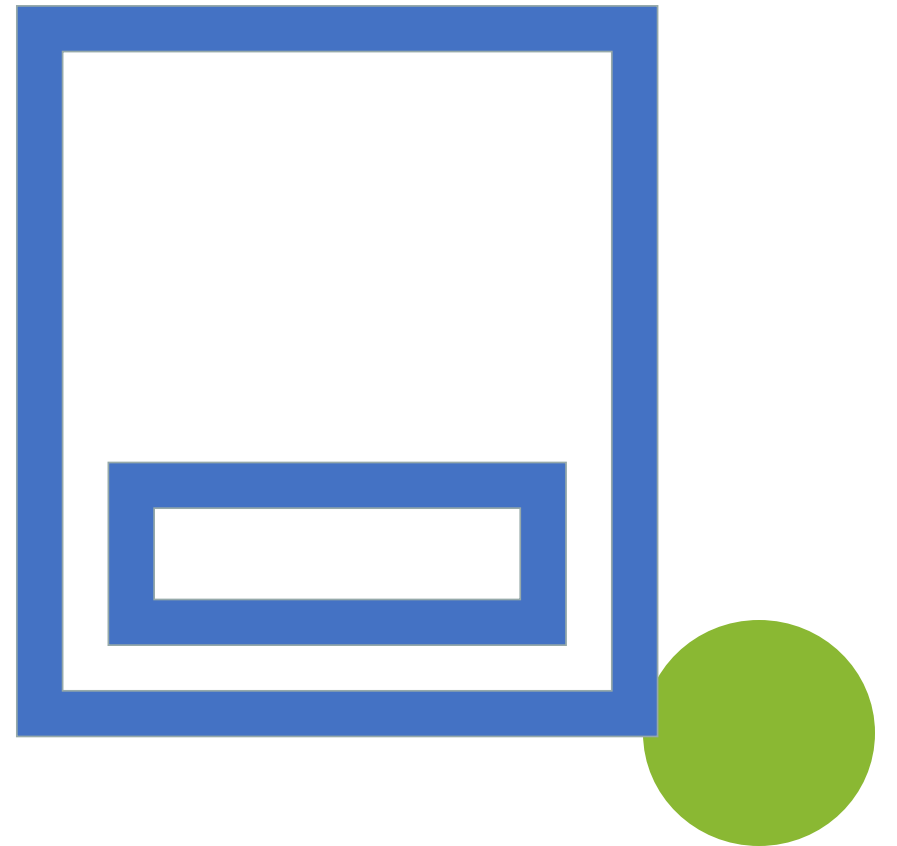
- Create nb files based on
  - Contiguity criterion
  - k-nearest neighbours and
  - Distance based (use three appropriate distances)Using the UK shapefile
- Create a GAL file using the "nb" object based on queen contiguity criterion. Coerce a relationship between Scotland and Northern Ireland by manipulating the GAL file. Read the GAL file back into your R session and inspect the new "nb" object.



# W Matrix

# Create a weights list object

- The **nb list** object that we created before contains information on how areal units are related to one another
- It must be **converted to a weights matrix before it can be used in statistical analyses.**
  - See the "uk.xlsx" file
- We can do this by using the function `nb2listw()`
- This function converts the nb list to a weights list object that contains information about the type of the weights matrix to be used in the analysis.





# Example of Weights list object

```
> (UK_lw = nb2listw(UK_nb))  
Characteristics of weights list object:  
Neighbour list object:  
Number of regions: 12  
Number of nonzero links: 42  
Percentage nonzero weights: 29.16667  
Average number of links: 3.5
```

Weights style: W (default)

Weights constants summary:

|   | n  | nn  | S0 | S1       | S2       |
|---|----|-----|----|----------|----------|
| W | 12 | 144 | 12 | 7.751111 | 50.16889 |

# Different weights “styles”



- Style options
  - style="B" is the basic binary coding,
  - style="**W**" is **row standardised/normalised** (sums over all row links to  $n$ ),
  - style="C" is globally standardised (sums over all links to  $n$ ),
  - style="U" is C divided by the number of neighbours (sums over all links to unity).
- Row normalised W matrix
  - Row standardising can turn a symmetric weights matrix into an asymmetric one.
  - There is also the chance of boosting the weightage of the units near the boundary.
  - In many econometric applications it is the norm to use style="W" matrix as it has the property of  $\text{sum}(W)=n$  which is quite useful in numerical optimisations and other computations related to estimation of parameters and test statistics. (See next chapter).

# Inspect weights list

```
> names(UK_lw)
[1] "style"      "neighbours" "weights"
> names(attributes(UK_lw))
[1] "names"      "class"      "region.id"  "call"      "GeoDa"
> summary(unlist(UK_lw$weights))
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.1667 0.2000  0.2667  0.2857 0.3333  1.0000
```

# Zero policy

---

By default, `zero.policy=F`, which means there cannot be “loners”.

---

You can disable this by setting `zer.policy=T` but in many econometric applications such a  $W$  matrix raises more problems than can be answered, for example, the effective sample size, a.k.a degrees of freedom in tests model identification issues, etc.



# Zero Policy example

```
> (SG1_lw = nb2listw(SG1_nb))  
Error in nb2listw(SG1_nb) : Empty neighbour sets found
```

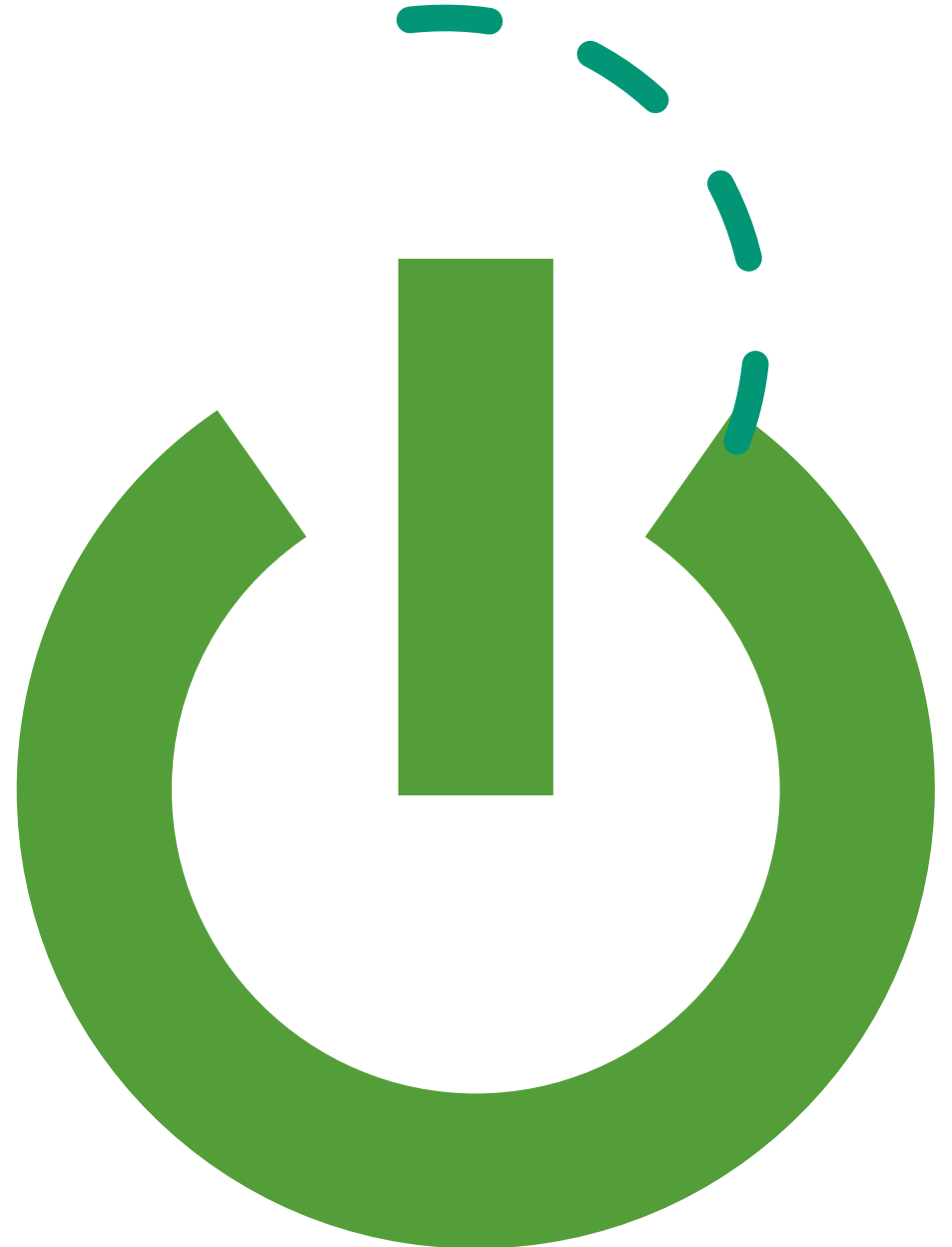
```
> (SG1_lw = nb2listw(SG1_nb, zero.policy=T))  
Error in print.listw(x) : regions with no neighbours  
found, use zero.policy=TRUE
```

```
> print(SG1_lw, zero.policy=T)  
Characteristics of weights list object:  
Neighbour list object:  
Number of regions: 55  
Number of nonzero links: 258  
Percentage nonzero weights: 8.528926  
Average number of links: 4.690909  
1 region with no links: 38
```

Weights style: W

Weights constants summary:

| n | nn | S0       | S1       | S2 |
|---|----|----------|----------|----|
| W | 54 | 29       | 16       | 54 |
|   |    | 26.00677 | 229.5735 |    |



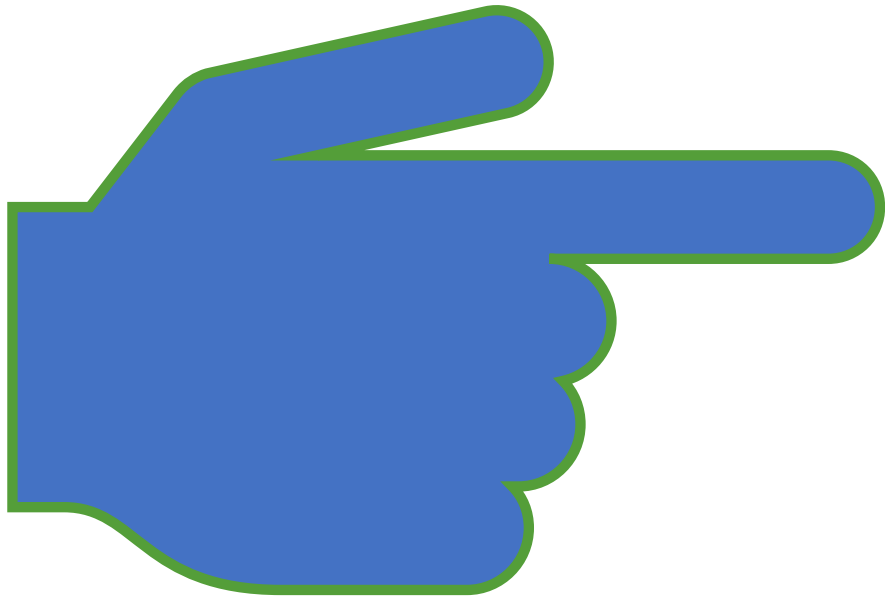
# Take home points

- Areal data and its properties
- Neighbours and weights matrix
  - Distance-based
  - Contiguity
- .GAL file: portable weights matrix
- Create neighbours from spatial polygons
- Convert nb list to weights list



# Important R functions

- `read.gal()`
- `write.nb.gal()`
- `poly2nb()`
- `knearneigh()`
- `dnearneigh()`
- `nb2listw()`



# References

- ***Spatial Analysis*** by Tonny Oyana 2<sup>nd</sup> edition, Chapter 7.
- ***Applied Spatial Data Analysis with R*** by Roger S. Bivand, Edzer Pebesma, and Virgilio Gómez-Rubio, 2<sup>nd</sup> edition, (2013), Chapter 9.
- `> vignette(package="spdep")`
- [https://r-spatial.github.io/spdep/articles/nb\\_sf.html](https://r-spatial.github.io/spdep/articles/nb_sf.html)