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GEOGRAPHICAL STATISTICS & THE GRID

Rich Harris, Chris Brunson and Daniel Grose (Universities of Bristol, Leicester & Lancaster)

OUTLINE

About Geographically Weighted Regression (GWR)

- Chris Brunsdon, Department of Geography, University of Leicester
- cb179@le.ac.uk
- An example to illustrate a problem of using GWR with large datasets
- 'The solution'
 - Rich Harris, School of Geographical Sciences, University of Bristol
 - rich.harris@bris.ac.uk

LOCAL VS GLOBAL STATISTICS

o Global

- similarities across space
- single-valued statistics
- non-mappable
- GIS "unfriendly"
- search for regularities
- aspatial

o Local

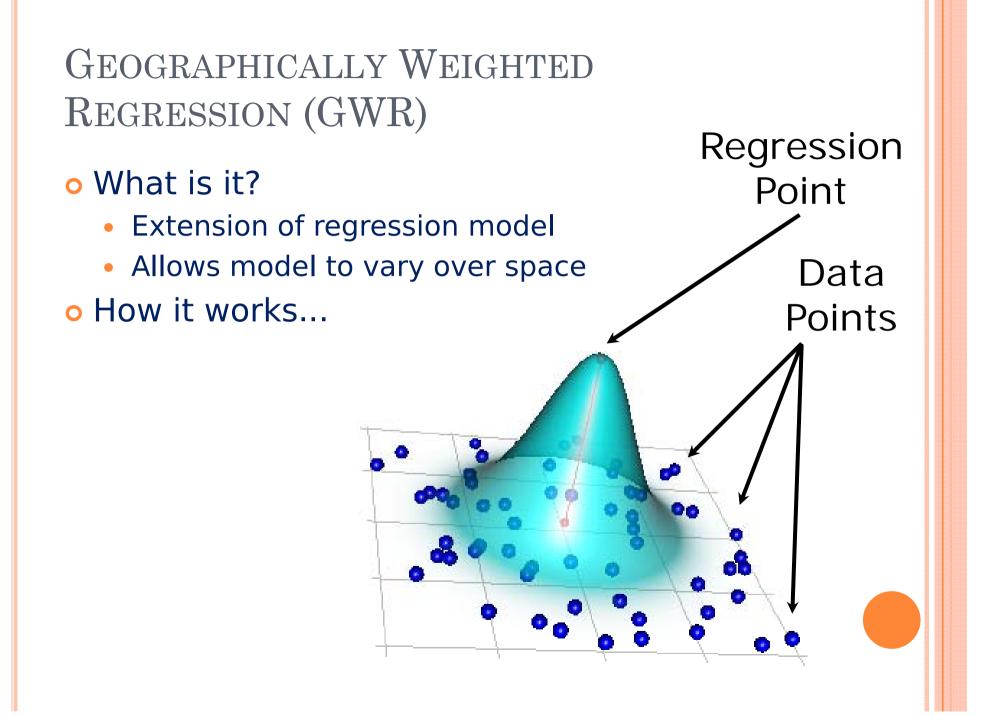
- differences across space
- multi-valued statistics
- mappable
- GIS "friendly"
- search for exceptions
- spatial

 Local statistics are spatial disaggregations of global statistics

WHY MIGHT RELATIONSHIPS VARY SPATIALLY?

Sampling variation

- Relationships intrinsically different across space e.g. differences in attitudes, preferences or different administrative, political or other contextual effects produce different responses to the same stimuli.
- Model misspecification suppose a global statement can ultimately be made but models not properly specified to allow us to make it. Local models good indicator of how model is misspecified.



IN GWR WE CAN ALSO...

- estimate local standard errors
- calculate local leverage measures
- perform tests to assess the significance of the spatial variation in the local parameter estimates
- perform tests to determine if the local model performs better than the global one

BUT

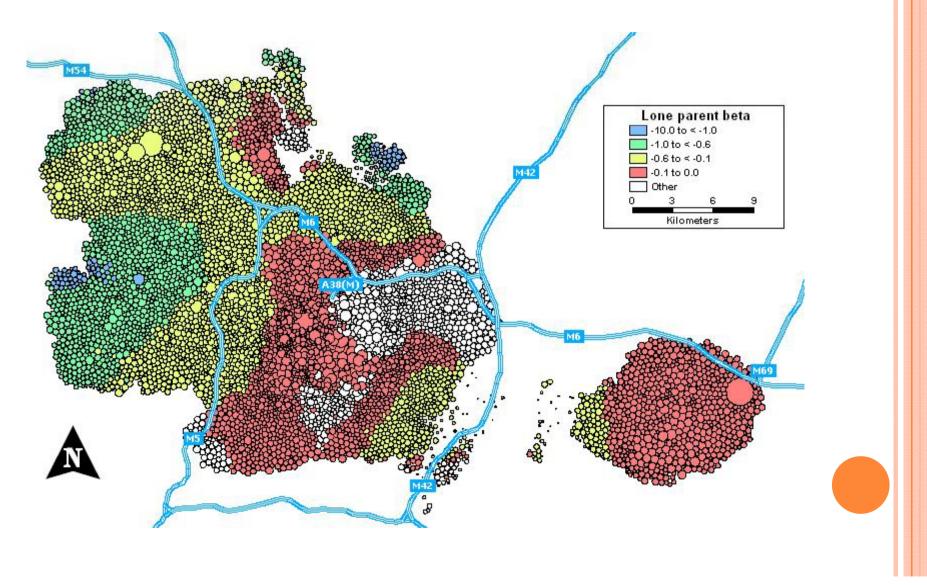
- Computationally very demanding
- Need to fit weighted regression models in several places
- Sometimes not viable on a single computer
- How do we address this problem?

EXAMPLE

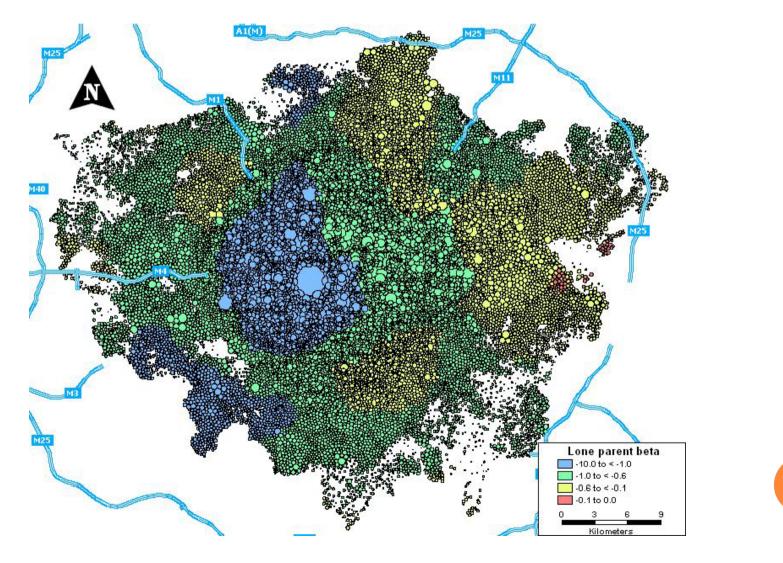
• Y: Proportion of households without a car

- X₁: Proportion of persons of working age unemployed
- X₂: Proportion of households in public housing
- X₃: Proportion of households that are lone parent households
- X₄: Proportion of persons 16 or above that are single
- X₅: Proportion of persons that are "white British"
- o n = 165,665

SPATIAL VARIATION IN THE LONE PARENT COEFFICIENT (WEST MIDLANDS)



SPATIAL VARIATION IN THE LONE PARENT COEFFICIENT (LONDON)



SCALING PROBLEMS

There are n regression modelsBut actually there are many more:

- n× g
- g is the number of iterations to optimize the bandwidth, b

 And you also need to calculate the distance matrix, D: (n²)

TAKES A LONG TIME!

o If n = 100,000, on a single processor

- Would take about half a day to calculate D
- Would take about a fortnight to find b
- But GWR is intended for exploratory analysis!
- The main bottleneck is the calibration of b
 - Because the regression calculations are O(n³), the distance calculations are O(n²)

FORTUNATELY

- The regression models are fitted entirely independent of each other. The results are pooled and compared *at the end*.
- The process is sequential but it can also be embarrassingly parallel
 - For GWR, the (distance-weighted regression) function stays the same, only the data are changing.
 - Each spatial subset is handled separately from the next.
- True of many methods of spatially localized analysis
- Suitable for a computational grid
 - The UK's National Grid Service (NGS)

MULTIR AND PARALLEL R

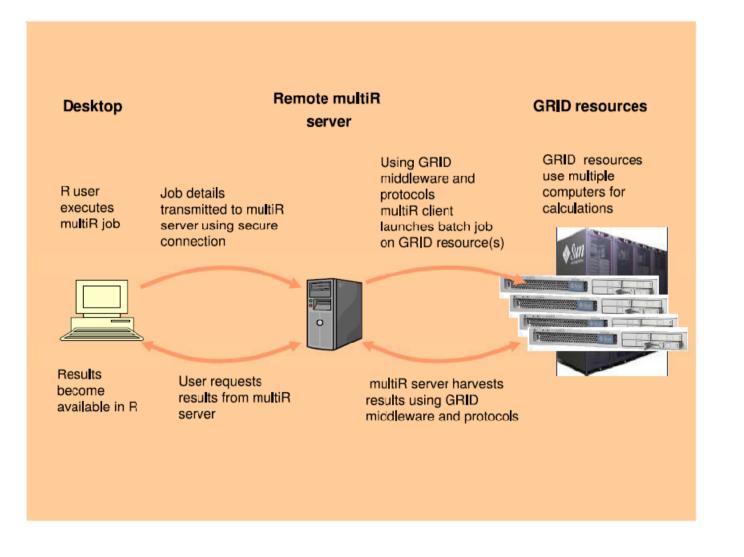
- R is a free software environment for statistical computing and graphics
 - http://www.r-project.org/
- There is an implementation of GWR in R
 - The spgwr package
- In R, there is a method to invoke a function a number of times with varying argument values
 - sapply
- The idea is to invoke the function on different processors running on the NGS.
- multiR (thanks to Daniel Grose) is both an 'add in' to R and a server (currently at Lancaster) which provides middleware between desktop R and the NGS.

EXAMPLE

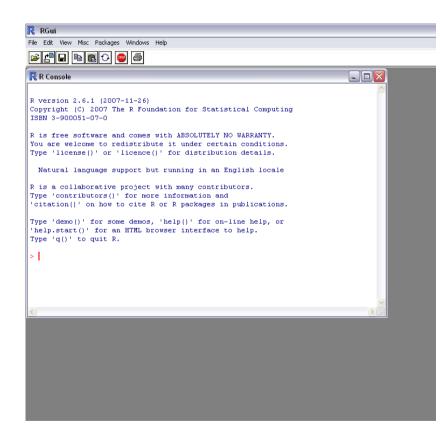
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- o n = 165,665

THE THREE TIER CLIENT/SERVER ARCHITECTURE EMPLOYED BY MULTIR



THIS IS R (IN WINDOWS)



Available free from

 http://cran.rproject.org/

• We have a tutorial

- www.esrcsocietyto day.ac.uk
- type 'Grid Enabled Spatial Regression Models' into the Search

FITTING A GWR MODEL IN R

```
> library(spgwr)
```

```
> load("carsmsoa.RData")
```

```
> names(car.msoa)
```

...

```
[1] "Name" "Borough" "ProfMan" "Renting"
"HHNoCar" "Easting" "Northing"
```

```
> coords = cbind(car.msoa$Easting,
    car.msoa$Northing)
```

```
> bandwidth = gwr.sel(HHNoCar ~ Renting, data =
    car.msoa, coords)
```

Bandwidth: 25571.63 CV score: 4.278767

```
Bandwidth: 41334.45 CV score: 4.373939
```

```
Bandwidth: 1467.076 CV score: 2.922873
```

```
> gwr.model1 = gwr(HHNoCar ~ Renting, data =
    car.msoa, coords, bandwidth)
```

FITTING A GRID RUN GWR MODEL

```
> library(spgwr.dist)
```

- > load("D:\\Data\\GWR-Workshop\\Data\\Exercises\\carsmsoa.RData")
- > names(car.msoa)
- [1] "Name" "Borough" "ProfMan" "Renting"
 "HHNoCar" "Easting" "Northing"
- > coords = cbind(car.msoa\$Easting, car.msoa\$Northing)
- > session = multiR.session.dlg()
- > bandwidth = gwr.sel.dist(session, HHNoCar ~
 Renting, data = car.msoa, coords,
 max.processors = 50)
- > gwr.model2 = gwr.dist(session, HHNoCar ~
 Renting, data = car.msoa, coords, bandwidth,
 max.processors = 50)

'THE SESSION'

🏸 Session Creation Dialo 🗖 🗖 🔀	
Host Name	stats-grid.hpc.lancs.ac.uk
Port Number	50201
Certificate File	C:/OpenSSL/bin/usercer
Key File	C:/OpenSSL/bin/userkey
CA File	C:/OpenSSL/bin/escienc
Number of Days	1
Create	Cancel

 It is loading and dealing with the various security certificates that are needed to use the NGS

OBTAINING THE CERTIFICATES

This is required for all NGS servicesUser certificates:

- Apply at https://ca.grid-support.ac.uk/
 - The certificate is issued for a web browser: the one you apply from needs to be the same as the one which will receive it.
- Then need to:
 - Export the certificate from the browser
 - use OpenSSL toolkit to convert into two files (the certificate and its key file) and to generate a proxy certificate
 - See www.grid-support.ac.uk/content/view/67/184/
- You also need to apply for an NGS account

 http://www.grid-support.ac.uk/content/view/221/171/
- CA certificate:
 - Download from http://www.gridsupport.ac.uk/content/view/182/184/

CONCLUSIONS AND CAVEATS

- There is no point in using Grid GWR for small data sets (e.g. n < 1000)
- And you should not expect instant results with large datasets (it still takes a second or so for each regression fit and you don't have that many processors)
- You cannot presently disconnect from R when running a Grid GWR but that should follow (and return later to collect the results)
- The software are still being tested

POTENTIAL

• multiR is more generic than GWR

Imagine

```
function1 = function(its_parameters) {
```

```
what it does
```

}

o Then

```
some_results = multiR(session, function1,
    list=(the_data))
```

In other words, the function is running in parallel on the NGS

```
    Applications include spatial statistics,
geostatistics, 'hot spot analysis', simulation,
etc.
```

ACKNOWLEDGMENTS

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