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## GEOGRAPHICAL STATISTICS \& The Grid

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## OutLine

- About Geographically Weighted Regression (GWR)
- Chris Brunsdon, Department of Geography, University of Leicester
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- An example to illustrate a problem of using GWR with large datasets
- 'The solution'
- Rich Harris, School of Geographical Sciences, University of Bristol
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## Local Vs Global Statistics

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


## Geographically Weighted Regression (GWR)

- What is it?

Regression

- Extension of regression model
- Allows model to vary over space
- How it works...



## 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
- $\mathrm{X}_{1}$ : Proportion of persons of working age unemployed
- $\mathrm{X}_{2}$ : Proportion of households in public housing
- $X_{3}$ : Proportion of households that are lone parent households
- $X_{4}$ : Proportion of persons 16 or above that are single
- $X_{5}$ : Proportion of persons that are "white British"
- $\mathrm{n}=165,665$


## Spatial variation in the lone parent coefficient (West Midlands)



## Spatial variation in the lone parent COEFFICIENT (LONDON)



## Scaling Problems

- There are n regression models
- But actually there are many more:
- $\mathrm{n} \times \mathrm{g}$
- $g$ 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!

- If $\mathrm{n}=100,000$, 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 $O\left(n^{3}\right)$, the distance calculations are $O\left(n^{2}\right)$


## 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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## 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:<br>Data<br>GWRWorkshop<br>Data<br>Exercises<br>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'



- 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 services
- User 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. $\mathrm{n}<1000$ )
- 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
\}
- 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.


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