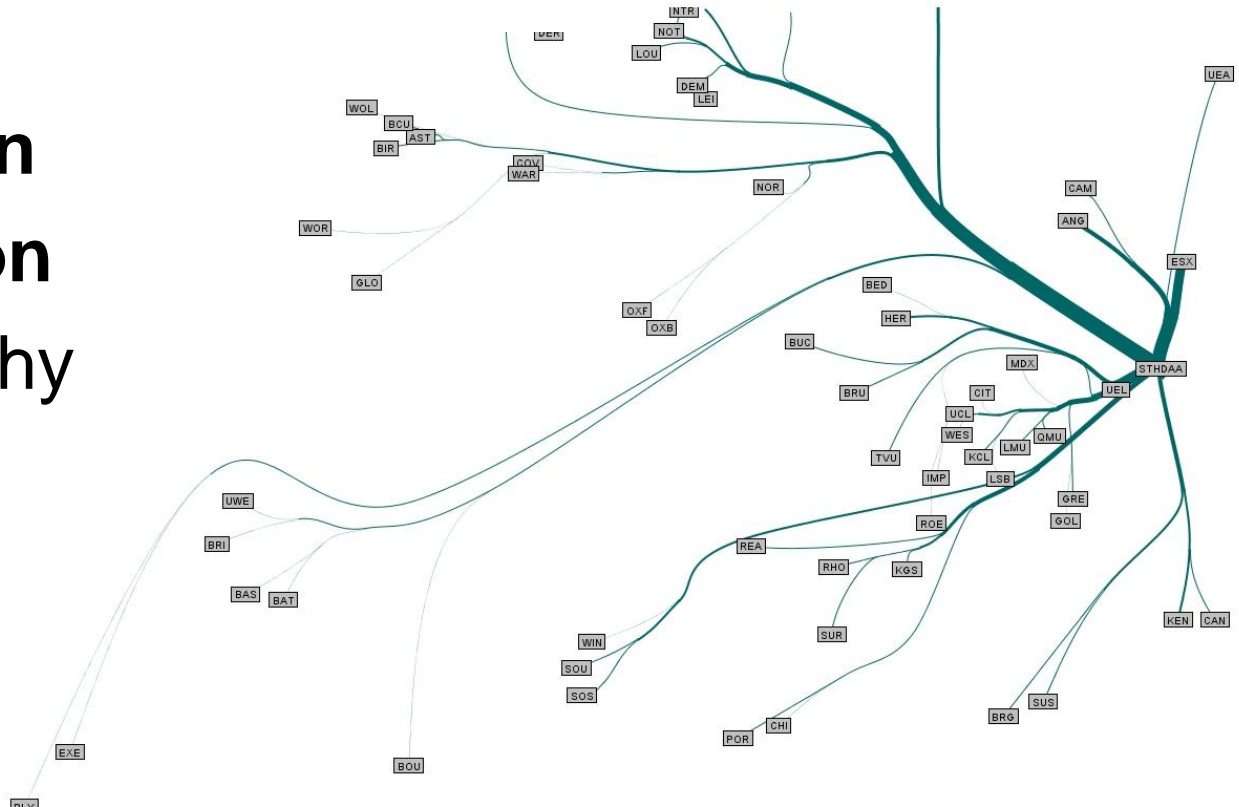


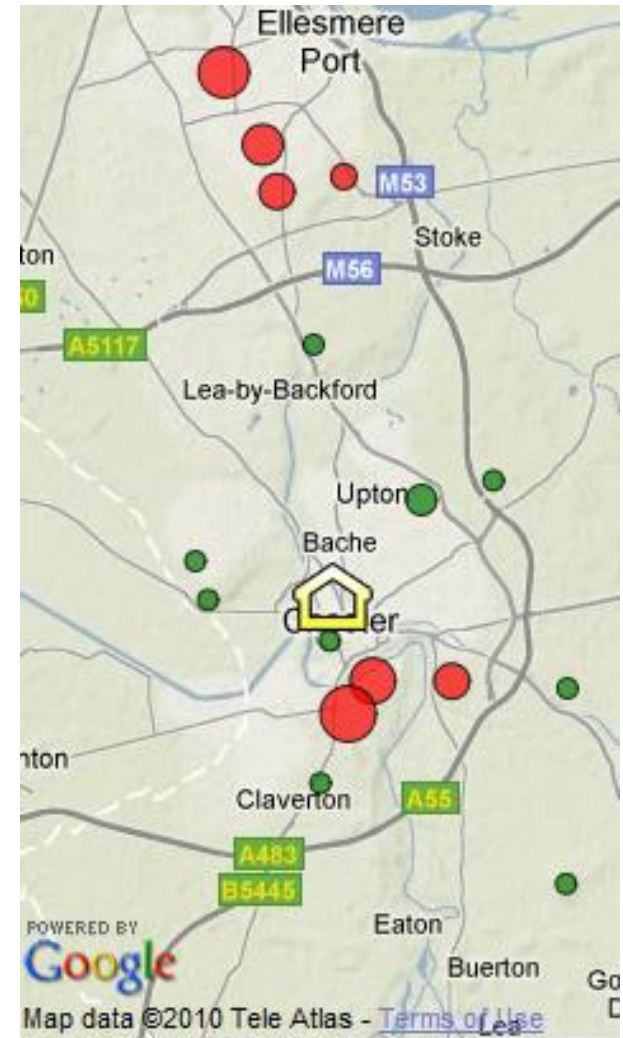
Spatial Interaction Models for Higher Education

Oliver O'Brien
Alex Singleton
UCL Geography



Contents

- Theory
 - Spatial Interaction Models
 - Geodemographics
- The Project
 - Putting them together
 - Simplifications
- Results
 - Interesting Anomalous Cases
 - Refinements



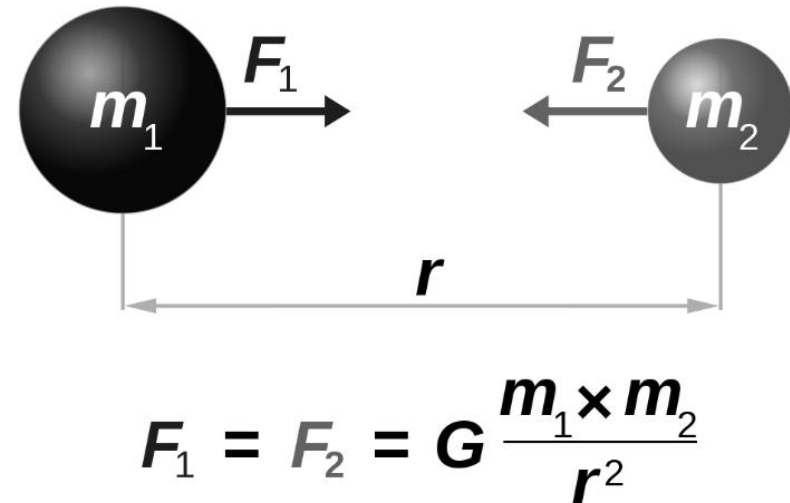
Spatial Interaction Models

- Modelling the flows from specific origin(s) to destination(s)
 - Commuting to work
 - Shopping at retail centres
 - *Exploring urban retail phase transitions*
(Dearden & Wilson)
 - NHS G.P. Provision
 - Summer holidays

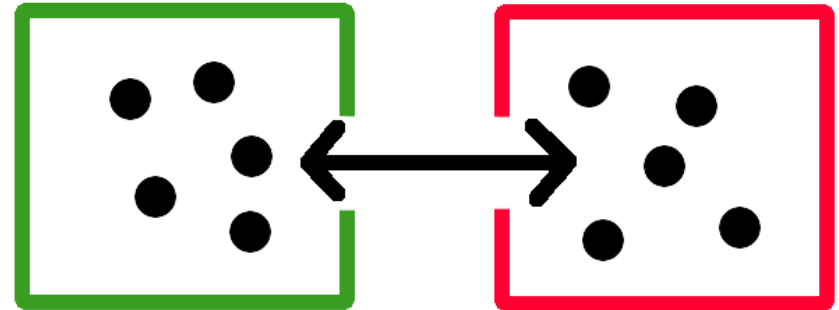


Spatial Interaction Modelling

- A classic gravity model
 - Analogous to Newton's Law of Universal Gravitation
- Distance (or cost) decay is always a key component
 - Tobler's "first law of geography"



Spatial Interaction Modelling



- $F_{12} = G m_1 m_2 r_{12}^{-2}$
- $S_{ij} = k O_i D_j d_{ij}^{-\beta}$ “unconstrained”
- $S_{ij} = A_i B_j O_i D_j e^{-\beta c_{ij}}$ “doubly constrained”
 - Can also derive it from entropy-maximising theory
 - A_i depends on B_j which depends on A
 - Solve iteratively

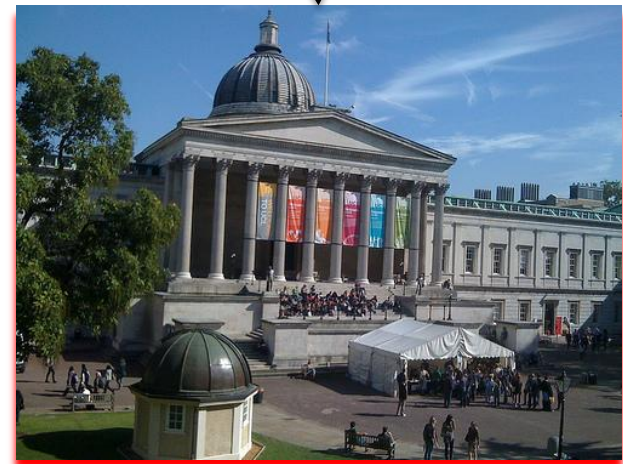
Constraining the Model

- Doubly constrained model
 - A fully closed system
 - e.g World Travel
- Singly constrained model
 - A finite origin population or destination population
 - e.g. Retail - finite number of shoppers, but shopping centre will never want to be “full” and turning them away – particular if capacity is measured in \$\$\$.
- Partially constrained model
 - A combination of the two
 - Some destinations full, others have spare capacity.
 - e.g. NHS doctor’s surgeries in a local authority.



Spatial Interaction Modelling for Higher Education

- The flows are from schools and F.E. colleges to universities
- Timescales are “different”
 - Flow is normally termly or one-way rather than daily or weekly
- Distances are “different”
 - Often intercity rather than intracity
- Distance is less important
 - Going to the “right” university is important for most people



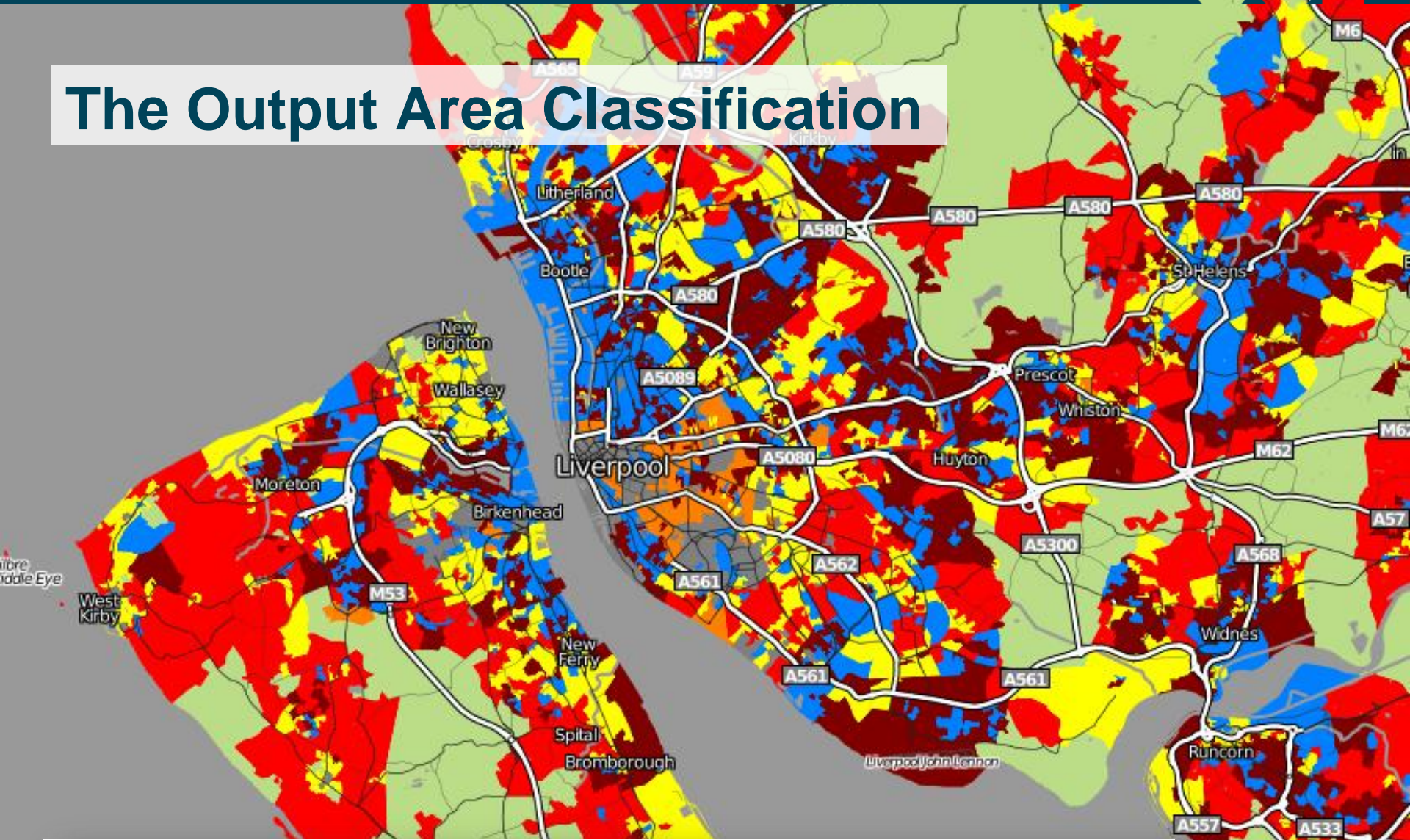
Partially Constrained Model

- Appropriate for modelling flows to higher education
 - More school pupils than university places but not every course at every university is fully subscribed
 - Have both “Selective” and “Recruiting” universities
 - Universities have quotas rather than operating in a fully unconstrained market
 - Many more universities have become selective recently
- Can treat singly-constrained and doubly-constrained flows separately
 - mark each flow appropriately in each iteration during the model run as the destinations “fill up”

Geodemographics

- Demographic characteristics (age, ethnicity, housing type, occupation, marital status, facilities)
- Interested in how geodemographics affect the patterns of university choice
- Using the Output Area Classification (Vickers)
 - Generalised (not education specific)
 - Available for each output area (typically 10 postcodes)
- Other UK geodemographic classifications
 - Mosaic (by Experian), Acorn

The Output Area Classification

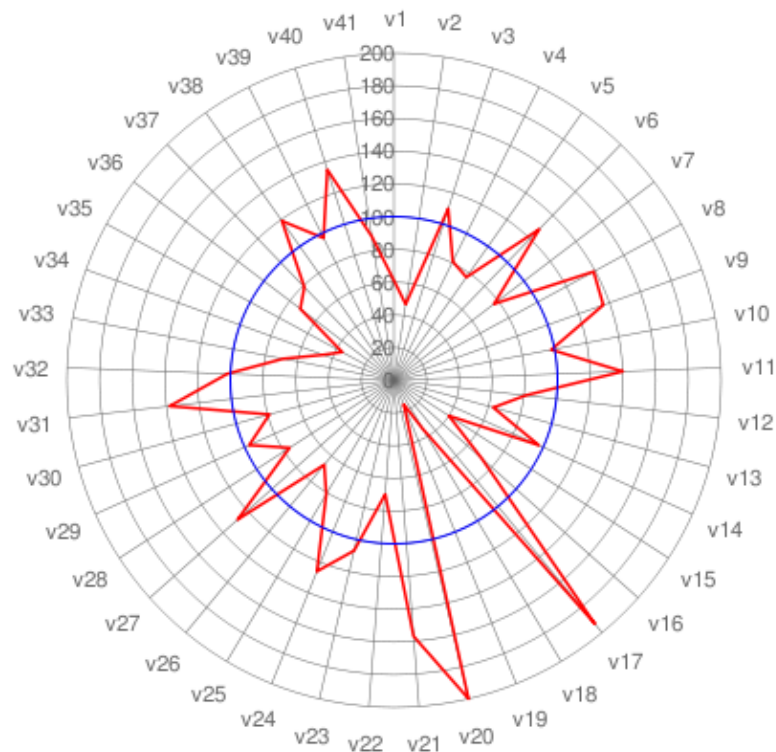


The OAC Map: Output Area Classifications for the U.K.

Blue Collar Communities	City Living	Countryside	Prospering Suburbs	Constrained by Circumstances	Typical Traits	Multicultural
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Output Area Classification 2A1

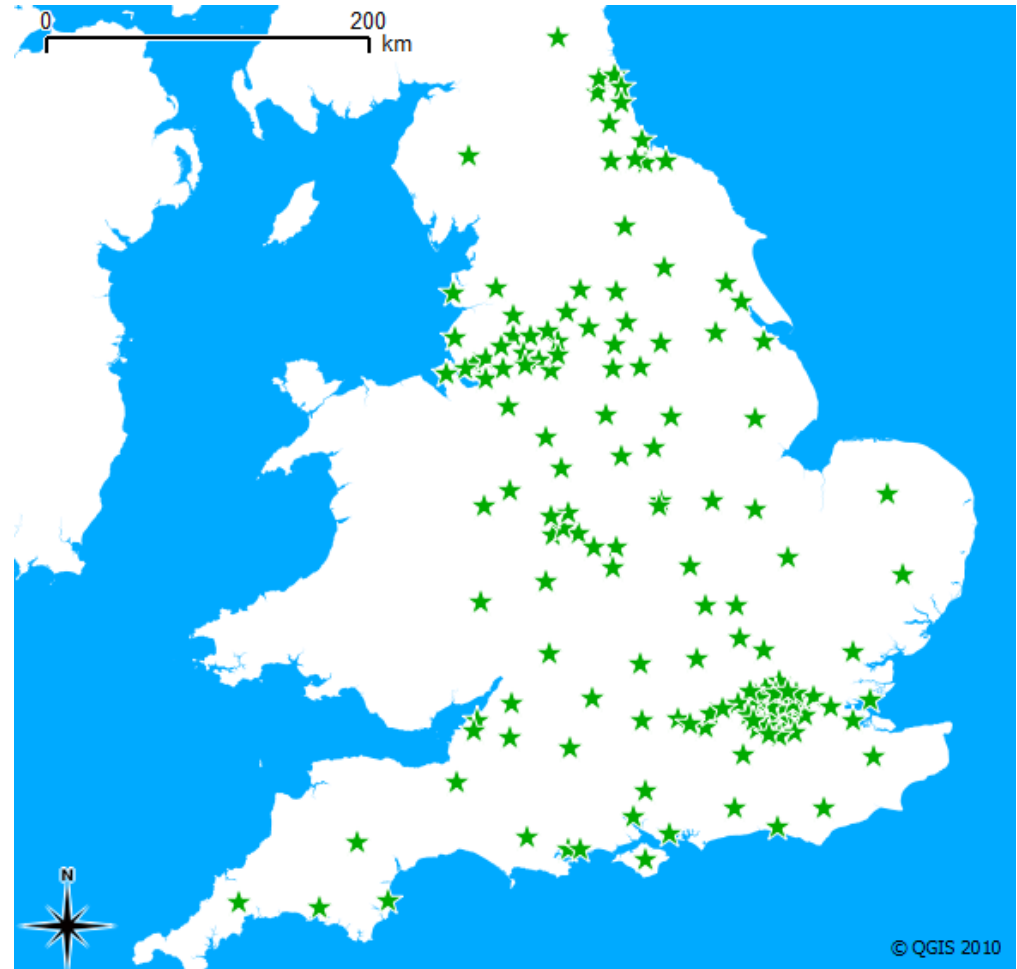
“City Living – Settled in the City 1”



- v1 Age 0-4
- v2 Age 5-14
- v3 Age 25-44
- v4 Age 45-64
- v5 Age 65+
- v6 Indian, Pakistani or Bangladeshi
- v7 Black african, Black Caribbean or Other Black
- v8 Born Outside the UK
- v9 Population Density
- v10 Divorced
- v11 Single person household (not pensioner)
- v12 Single pensioner household
- v13 Lone Parent household
- v14 Two adults no children
- v15 Households with non-dependant children
- v16 Rent (Public)
- v17 Rent (Private)
- v18 Terraced Housing
- v19 Detached Housing
- v20 All Flats
- v21 No central heating

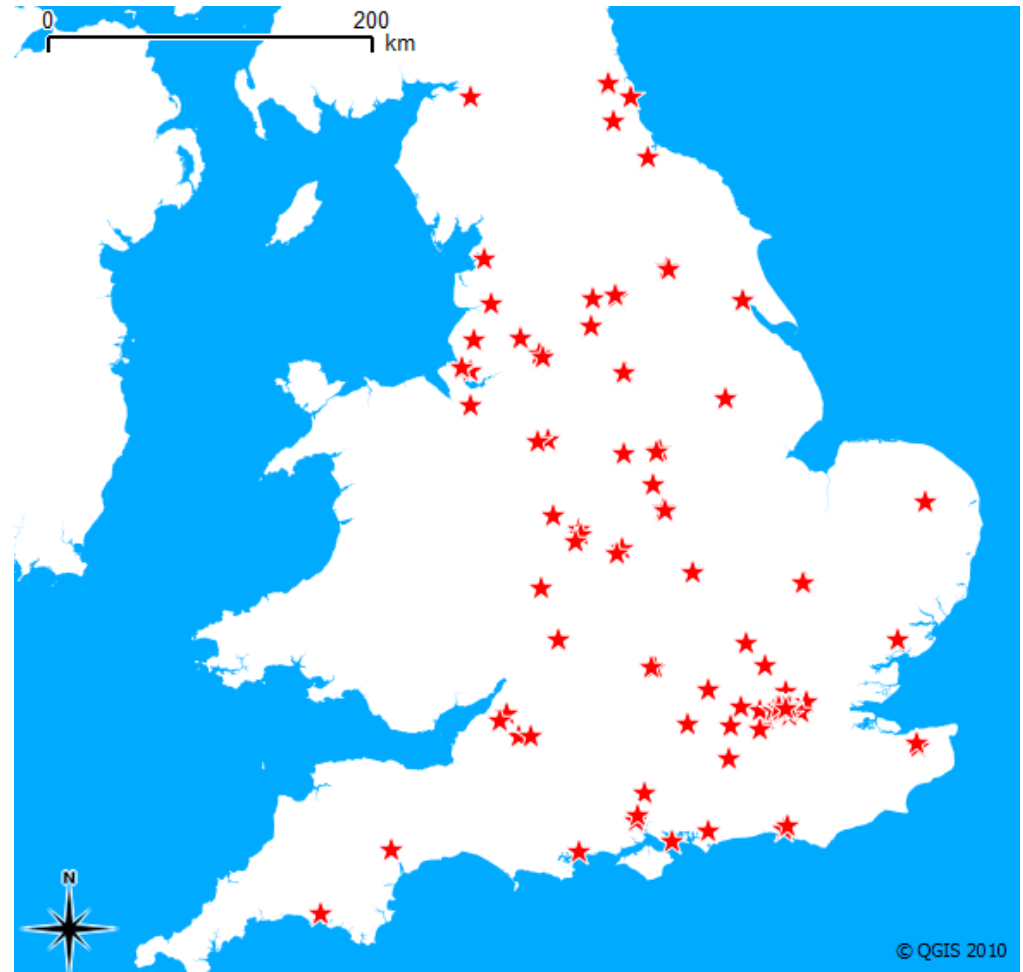
The Data – Origin Side

- National Pupil Database (NPD)
 - Home OAs (state only)
 - Used school OA for private schools
 - Includes attainment
- Individual Learning Records (ILR)
 - For sixth-form colleges
 - Home postcodes
 - Includes attainment
- OAC



The Data – Destination Side

- HESA Individual Student Records
 - Subjects
 - Home postcodes
 - A-Level point score
 - Nearly everything needed for modelling the flows, but excludes those who didn't go to university
 - Crucially, no theoretical capacity information



A Great Model – Modelling Reality

- Paper by Wilson (2002)
- $S_{ij} = A_i^{km} e_i^{km} P_i^k (W_j^{mh})^{\alpha^{km}} \exp(-\beta^{km} c_{ij}^k)$
- This is the singly-constrained form
 - Finite number of school students go to university
 - No restriction on places at university
 - Doubly-constrained version is quite similar to look at
- W is the “attractiveness” of the institution

A Great Model – Modelling Reality

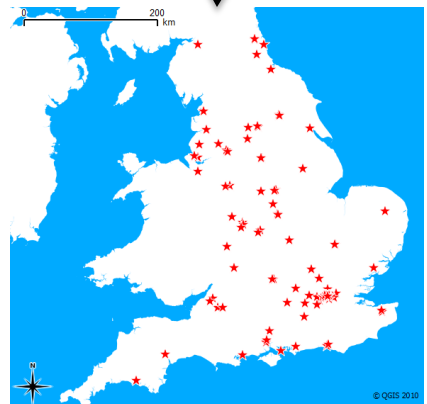
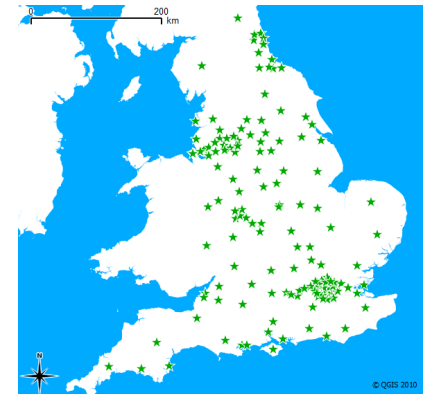
- 150 universities
- 3000 secondary schools + 500 F.E. Institutions
- 10 UCAS principal subject topics
 - e.g. Axxx – Medicine & Dentistry
- Multitude of possible attainments
 - A Level points scores, vocational qualifications, IB
 - Attainments are a useful additional factor for attractiveness

A Good Model – Simplifications

- In order to produce meaningful data on (relatively) small numbers (~300,000 annually) of students
 - use coarse categories
 - streamline the variables used
- Otherwise, the results would be a massive matrix with almost every value a fraction of a single person

A Good Model – Spatial Simplifications

- Assume universities are single-site
 - Generally using the “administrative HQ”
 - Some universities are fairly equally split
 - e.g. Angla Ruskin in Cambridge, Chelmsford
 - Ignore the Open University
- Assume English closed system
 - English schools and English universities only
- Make distance proportional to travel cost
- Assume schools and F.E. Institutions are a single institution at their LA’s centroid
- **149 “super schools”**



A Good Model – Origin Simplifications

- Ignore school types
- For pupils without postcode information assume the pupil's geodemographic is the same as the school's
- Assume pupils don't go to schools in a different local authority to that they live in
- Binary classification of attainment – “good”/”bad”
 - Based on A-level or equivalent points
- **2 attainment types**

A Good Model – Origin Simplifications

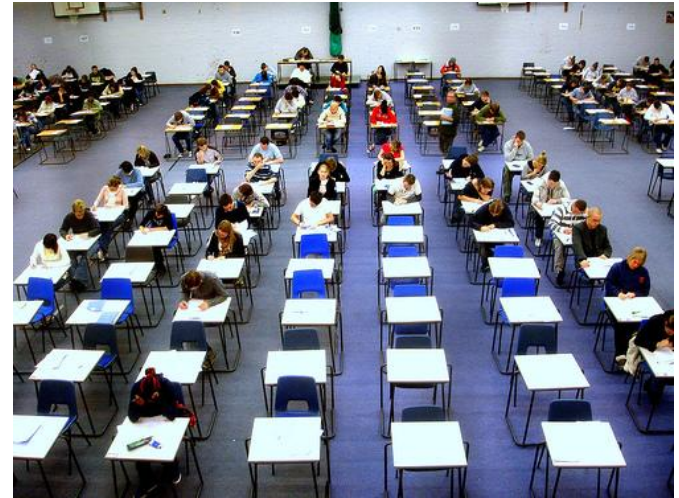
- Use the seven geodemographic “supergroups” from the Output Area Classification
 - Be aware of possible correlations between geodemographic and other factors included separately in the model, such as attainment
 - Very different overall numbers (and proportions) of each demographic go to universities
- **7 demographics**

A Good Model – Destination Simplifications

- Ignore subjects
 - Assume all universities offer all subjects and admissions criteria does not differ
 - But some universities are selective for some subjects (e.g. Medicine) and recruiting for some subjects (e.g. Physics)
 - The nearest few universities to someone may not offer the subjects that the person wants to study
- Ignore universities with a specialist subject focus
 - University for the Creative Arts
 - London School of Economics
 - These are also generally “small” universities
- **89 universities, 1 “subject”**

A Good Model – Destination Simplifications

- Binary classification of attainment requirement
 - “good only”
 - “any”
- Account for students not going to university by a special catch-all “university of last resort”
 - No “distance” element
 - Adjust attractiveness of this university to see the relative popularity of the other universities in the model



A Good Model – Destination Simplifications

- Attractiveness
 - Very subjective – different people like different things
 - Was originally modelled as a university “type”
 - Ancient, 19th century, Red brick, Plate glass, Post-1992
 - Funding type: Big research-focused institution & hospital, big research-focused, big teaching-based, small teaching
 - But difficult to categorise type and its relative effect on each of the origin geodemographics
 - Using Times Higher Education Score (range 200-1000)
 - Factor to modify its influence if necessary
- Attractiveness becomes less important and location more important, as more of the flows become doubly constrained (i.e. more universities fill to capacity)



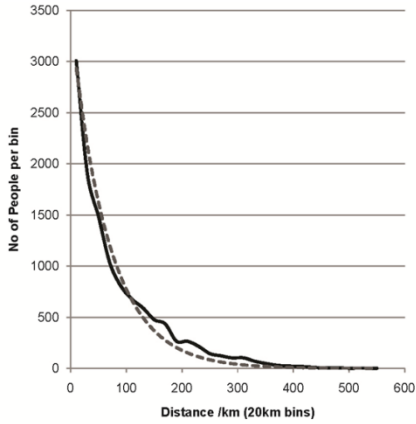
Simplified Form

- From:
$$\mathbf{S}_{ij} = \mathbf{A}_i^{km} \mathbf{e}_i^{km} \mathbf{P}_i^k (\mathbf{W}_j^{mh})^{\alpha^{km}} \exp(-\beta^{km} \mathbf{c}_{ij}^k)$$
- To:
$$\mathbf{S}_{ij} = \mathbf{A}_i^k \mathbf{P}_i^k (\mathbf{W}_j^h)^\alpha \exp(-\beta^k \mathbf{d}_{ij})$$
 - No subject consideration
 - No “demand” factor
 - Cost is replaced by distance
 - Numbers of i, j locations greatly reduced
 - Attractiveness is not dependent on geodemographic
- Similar for the doubly-constrained version

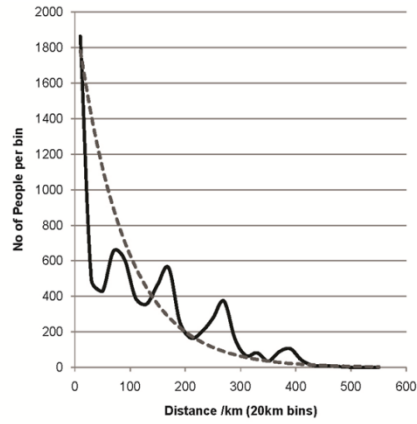
Calibrating the Model

- Find values for the constants in the equations
- A_i & B_j values are “balancing constants”
 - they converge on the correct values during iteration
- Calculate the β^k distance-decay with known flows
 - Overall distance decay for all pupils
 - Break down by geodemographic
 - Very unequal numbers within each geodemographic
 - Compare distance decay functions

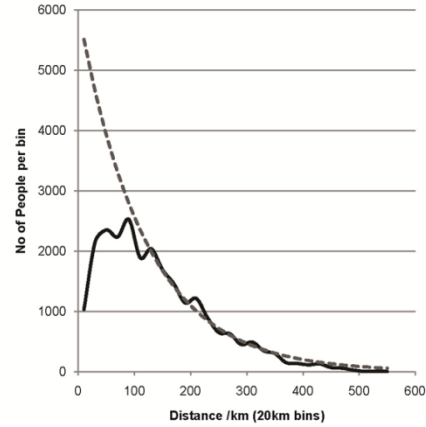
Calibrating the Model – Beta Decay



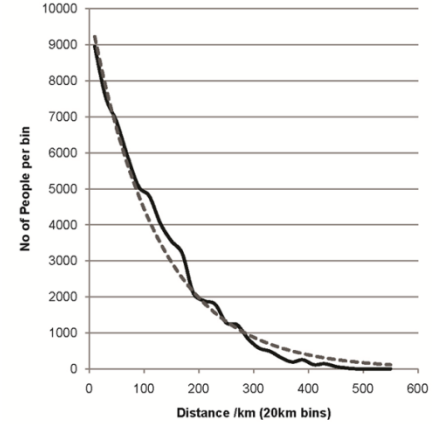
Blue Collar Communities



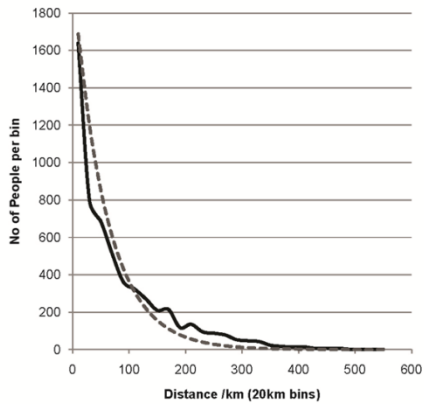
City Living



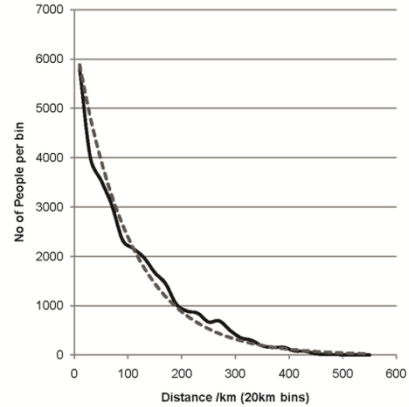
Countryside



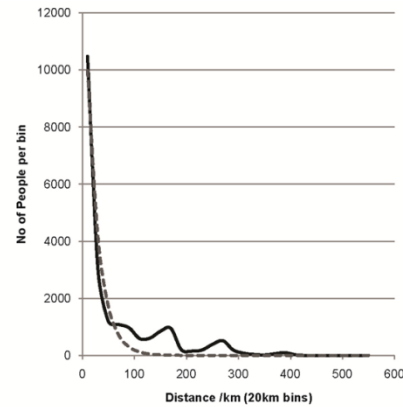
Prospering Suburbs



Constrained by Circumstances

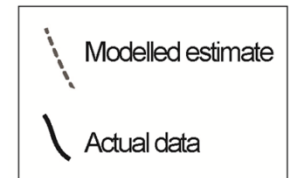


Typical Traits



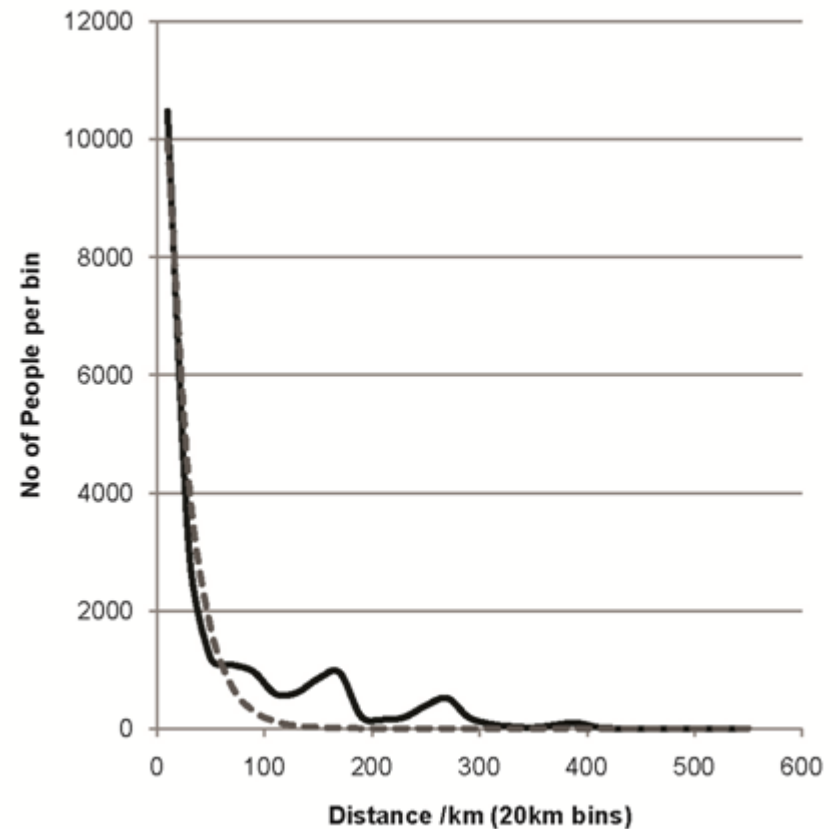
Multicultural

Legend



Calibrating the Model – Beta Decay

- London to Birmingham: 160 km
- London to Manchester: 260 km
- Distinctive pattern seen for the City Living & Multicultural demographics



Modelling

- Java

```
private void init()
{
    setupProperties();
    setupModel();
    m_resultsTable = new SIMResultsTable(ORIGIN_GROUPS, DESTINATION_GROUPS);
    setupGUI();
}

private double getSValueSC(OriginGroup oi, DestinationGroup dj)
{
    double distance = oi.getDistance(dj);
    return A.get(oi) * oi.getPopulation() * dj.getAttractiveness(oi) * Math.exp(-
}

private double getSValueDC(OriginGroup oi, DestinationGroup dj)
{
    double distance = oi.getDistance(dj);
    return A.get(oi) * B.get(dj) * oi.getPopulation() * dj.getCapacity() * Math.e
}

/*
 * Set the values of Bj to 1 initially.
 */
private void initialiseBs()
{
    Iterator<DestinationGroup> dIt = DESTINATION_GROUPS.values().iterator();
```

- Iterative process to calculate the normalising constants which depend on each other
 - Typically takes a minute to calculate the results

Results

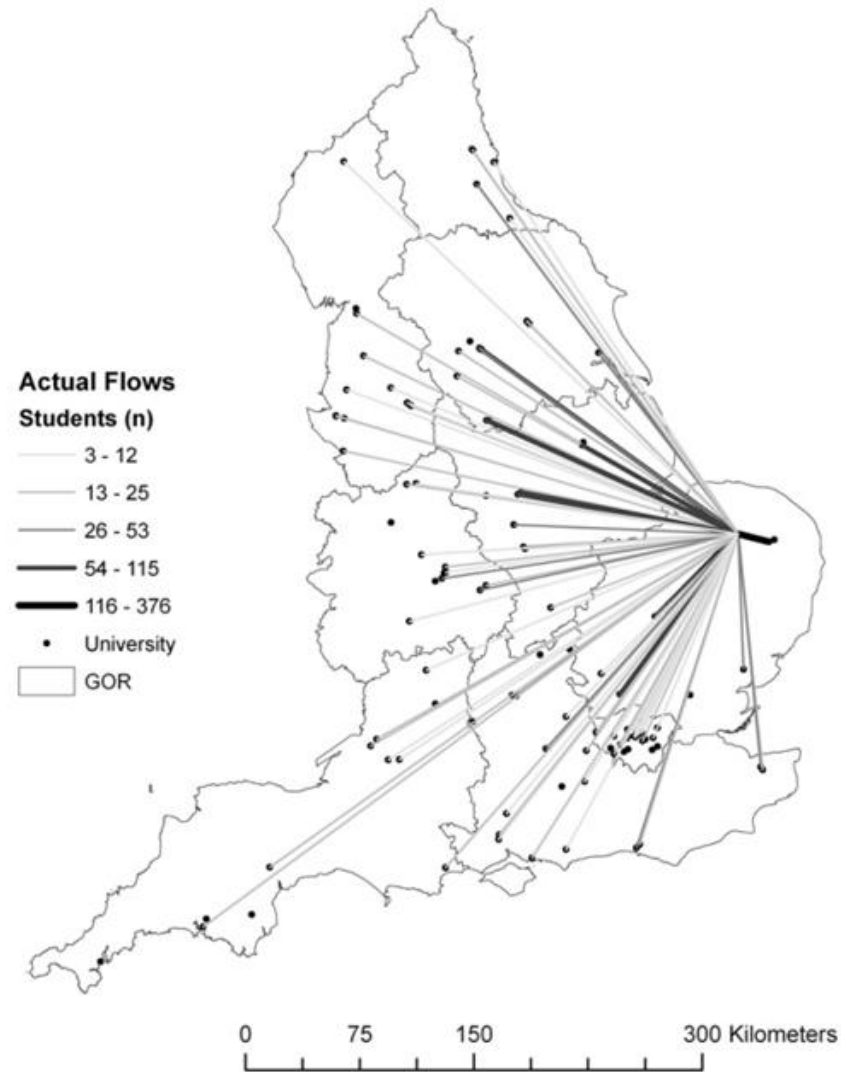
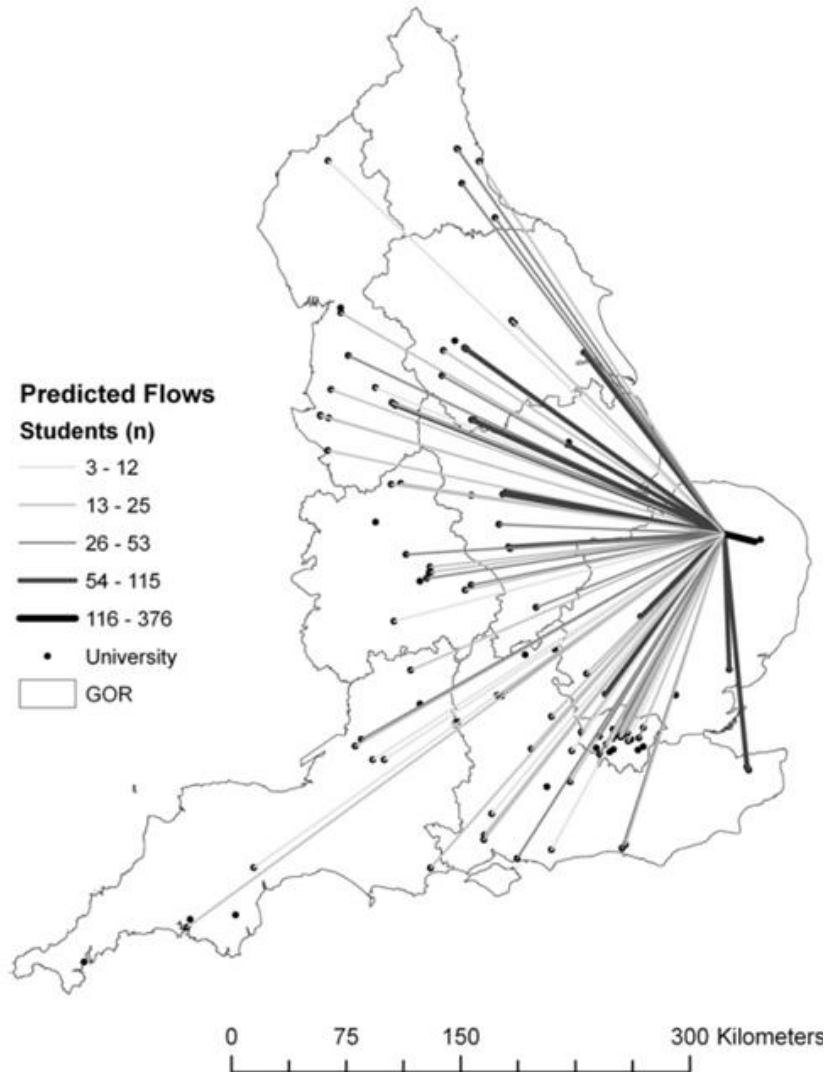
Partially Constrained Spatial Interaction Model

Flow Differences Show Predicted Flows Include Unknown OACs Excluding Unknown OACs

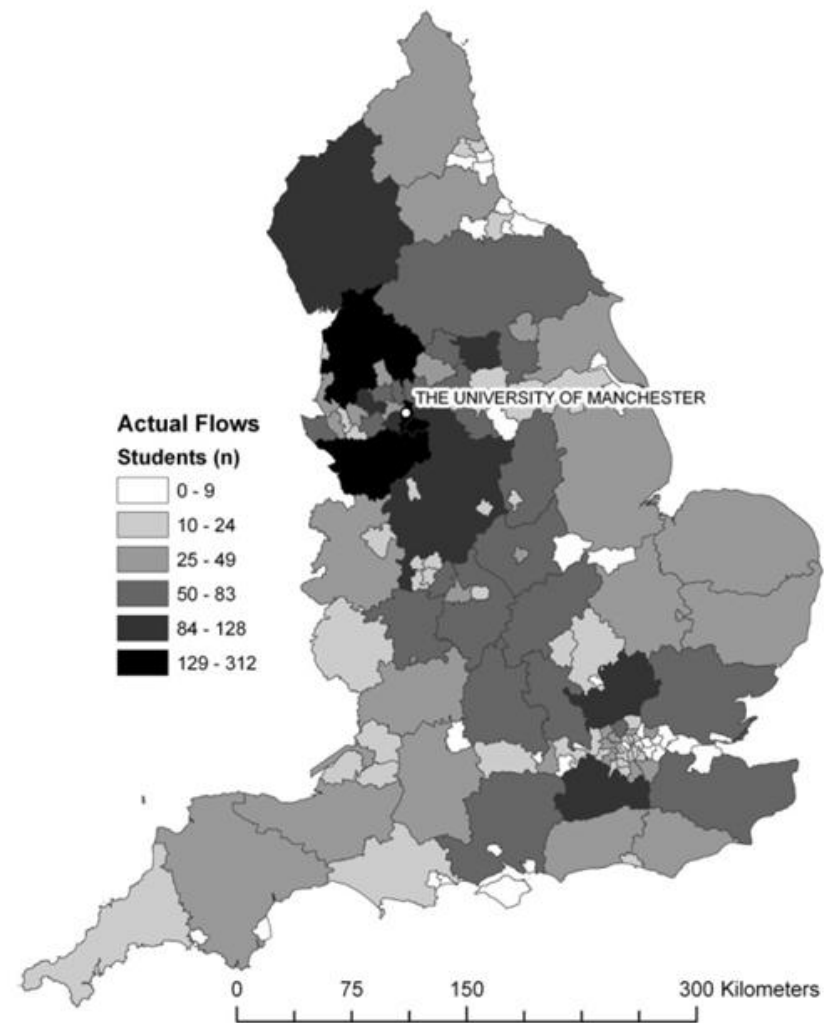
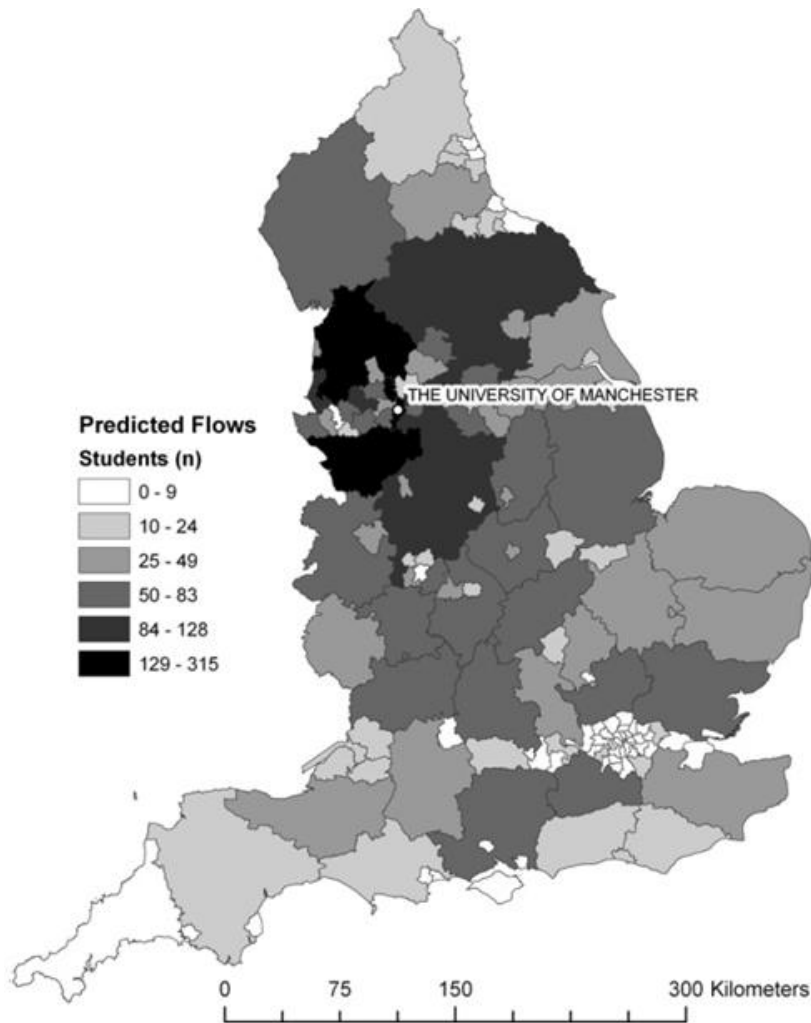
University ->	ANG	AST	BAS	BAT	BCU	BED	BIR	BOL	BOU	BRA	BRG	BRI	BRU	BUC	CAM	CAN	CHE	CHI	CIT	CLA	CMB	COV	DEM	DER	DUR
BIRM3M (Birming...	0	0	-2	0	-1	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	-1	0
BIRM4H (Birmingh...	3	-28	2	2	-42	1	-30	2	1	-1	1	0	-1	2	-3	2	5	1	0	7	1	-8	-2	3	4
BIRM4M (Birmingh...	1	-0	-1	-1	-77	-0	-7	-0	2	2	1	0	1	1	0	1	1	1	1	4	0	-21	1	-1	0
BIRM5H (Birmingh...	1	0	1	2	6	0	16	0	-0	1	0	4	1	-0	2	0	-0	0	0	1	0	2	7	2	1
BIRM5M (Birmingh...	2	13	1	0	28	1	-4	1	1	2	-0	0	1	1	0	0	3	0	1	4	-1	8	12	9	0
BIRM6H (Birmingh...	2	-15	2	5	-38	1	-17	1	0	-0	1	5	1	1	2	-0	2	1	1	5	0	-12	5	-1	3
BIRM6M (Birmingh...	6	13	3	0	-37	2	-9	3	6	5	3	0	3	3	0	2	9	1	2	12	1	3	24	9	0
BIRM7H (Birmingh...	0	-123	0	-2	107	0	-100	0	-2	-11	-2	-4	-2	0	-8	0	-2	0	-2	0	0	-54	-51	-2	-2
BIRM7M (Birmingh...	-5	-27	-1	-1	174	0	-30	0	-3	-4	-2	0	0	-3	0	0	-1	0	0	-1	0	-156	-45	-10	0
BLAB1H (Blackbur...	0	0	0	0	0	0	-1	0	0	-2	0	0	0	0	0	0	0	0	0	-2	-1	0	0	0	-1
BLAB1M (Blackbu...	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	-9	0	-1	0	0	0
BLAB2H (Blackbur...	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
BLAB2M (Blackbu...	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
BLAB3H (Blackbur...	0	0	0	-1	0	0	2	0	0	1	0	-1	0	0	0	0	1	0	0	-0	0	-1	1	0	3
BLAB3M (Blackbu...	0	0	0	0	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	-1	0	0	-1	0	0
BLAB4H (Blackbur...	0	0	0	0	-2	0	1	0	0	-1	0	-2	0	0	-1	0	-1	0	0	-10	0	-1	-2	0	-5
BLAB4M (Blackbu...	-1	0	0	0	0	0	0	-2	0	0	0	0	0	0	0	0	-2	0	0	-22	0	0	0	0	0
BLAB5H (Blackbur...	0	0	0	0	0	0	1	1	0	1	0	0	0	0	0	0	1	0	0	6	0	0	1	1	3

- Simple Java GUI to show the matrix of results
 - visually spot good/poor matches
 - refine model parameters
 - rerun

Results – Norfolk Schools to Universities



Results – Schools to University of Manchester

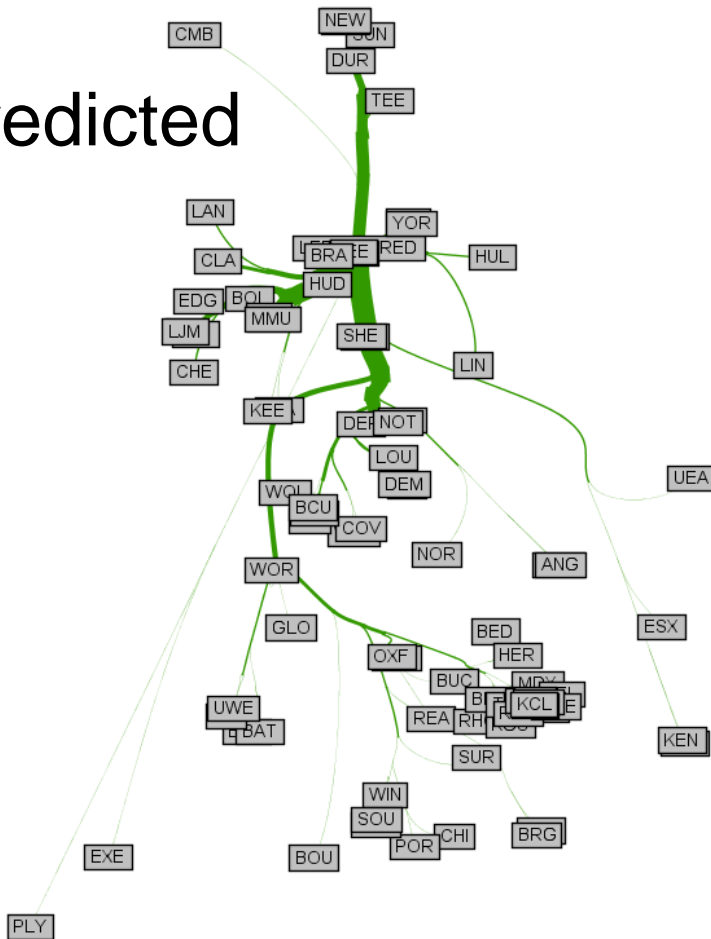


Results – Flow Maps

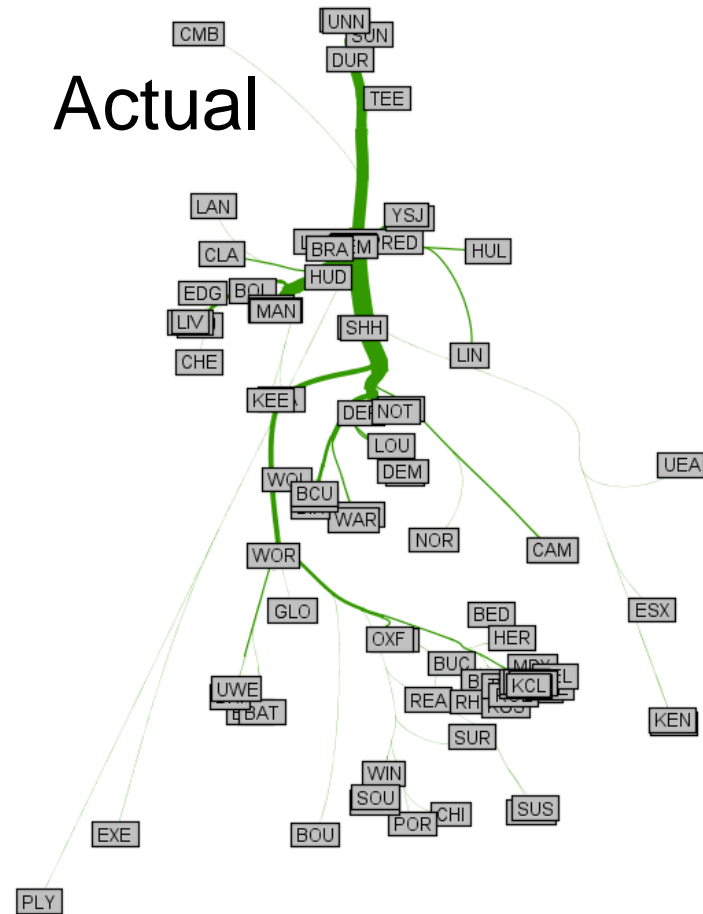
- *FlowMapLayout* Java application
 - Developed at Stanford for InfoVis 2005
- A more graphic & flexible presentation of the flows
 - Pseudo-spatial
 - Lengths and directions of the connecting lines are not meaningful

Results – Leeds Schools to Universities

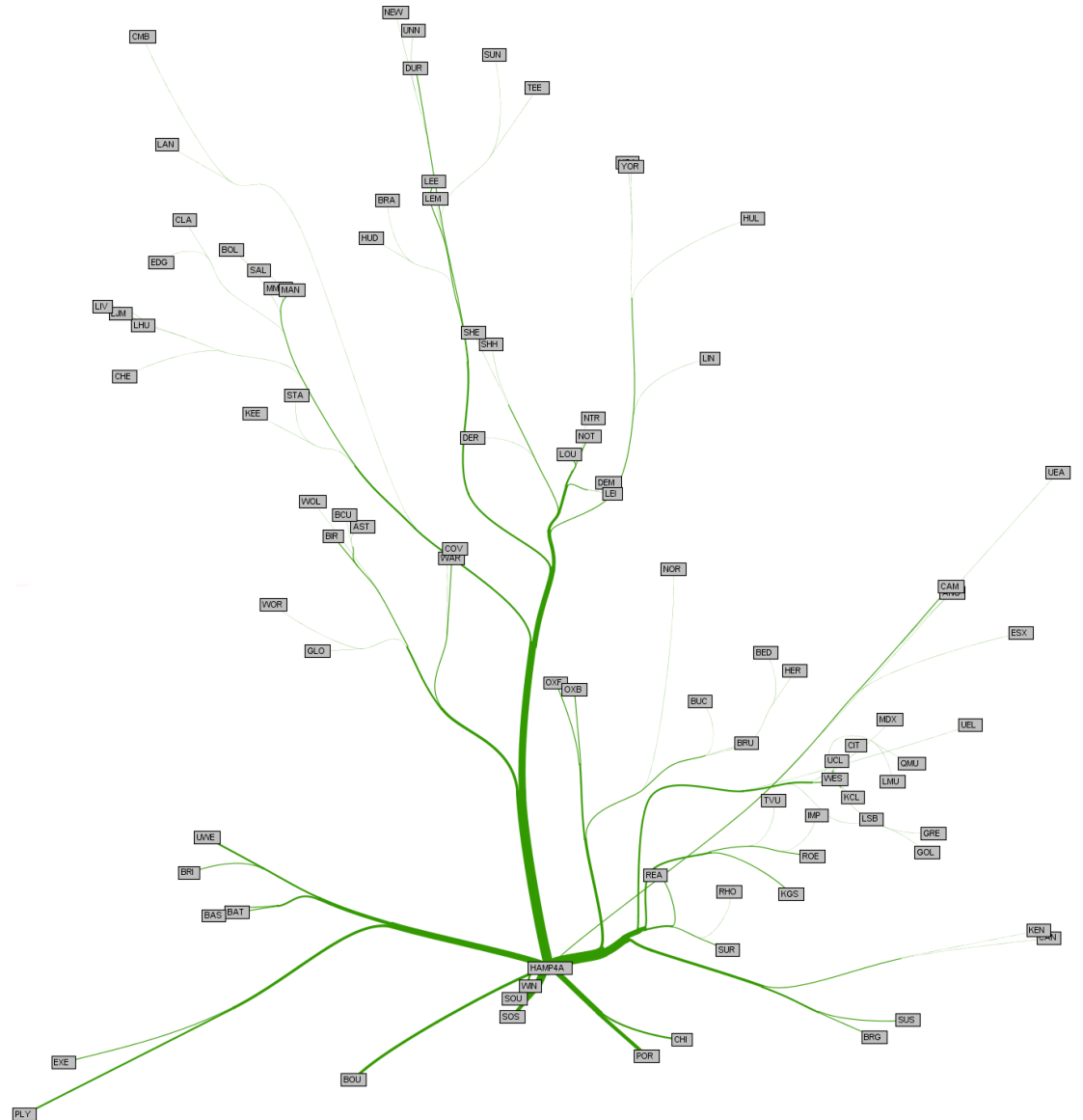
Predicted



Actual



Results – Hampshire Schools to Universities



Some Interesting Anomalous Results

- Flows significantly higher than expected
 - Flows from North-West London to Manchester
 - Essex to Exeter and Exeter to Essex
 - Small-distance flows to modern “metropolitan” universities, particularly paired with older institutions, such as Sheffield & Sheffield Hallam
- Flows significantly lower than expected
 - Yorkshire to/from Lancashire
 - Essex to/from Kent

Model Refinements

- Creating a genuinely partially-constrained model
 - University capacities are not known, instead we assume the actual enrolled numbers are all at capacity
 - This results in a completely doubly-constrained model, unless the “not at university” option is made attractive.
 - Possible solution would be to increase all capacities by a small % and adjust “not at university” attractiveness to rebalance the numbers
- The local “Metropolitan university” issue
 - Model adjusted to reduce the distances for these flows

Further Considerations

- Cost vs Distance – d_{ij} vs c_{ij}
 - Not necessarily linearly related for “life changing” spatial flows as such universities
- Straight-line distance is too simple
 - Natural barriers (e.g. mountain ranges, water)
 - Fast intracity & intercity transport networks
- Subject-specific analysis may be more revealing
 - e.g. flows for medicine courses only
- Including Scottish/Wales data
 - Potentially interesting with different fee requirements

References & Acknowledgements

- Wilson, A.G. (2000). The widening access debate: student flows to universities and associated performance indicators. *Environment and Planning A* **32** pp 2019-31
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http://graphics.stanford.edu/papers/flow_map_layout/
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- Vickers, D.W. and Rees, P.H. (2007). Creating the National Statistics 2001 Output Area Classification. *Journal of the Royal Statistical Society, Series A*
- The paper for this project is currently in review.

Graphics Acknowledgements

- Gravity model graphic: Wikipedia
- Liverpool Street Station commuters: steve_way on Flickr
- Aberystwyth University examination hall: jackhynes on Flickr

Q&A

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