

# Complexity in Land Use- Transport Modelling:

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 @jmmichaelbatty

<http://www.complexcity.info/>

<http://www.spatialcomplexity.info/>

# My Main Themes

- Complexity:

Models were simple and are now complicated or complex

- Data Has Changed: A New Take on Big Data

Big Data: data is extensive but also more flawed, trickier

- Models Have Changed: Extensive Places, Many Interactions, Many Users

Models are bigger, more complex, many more types

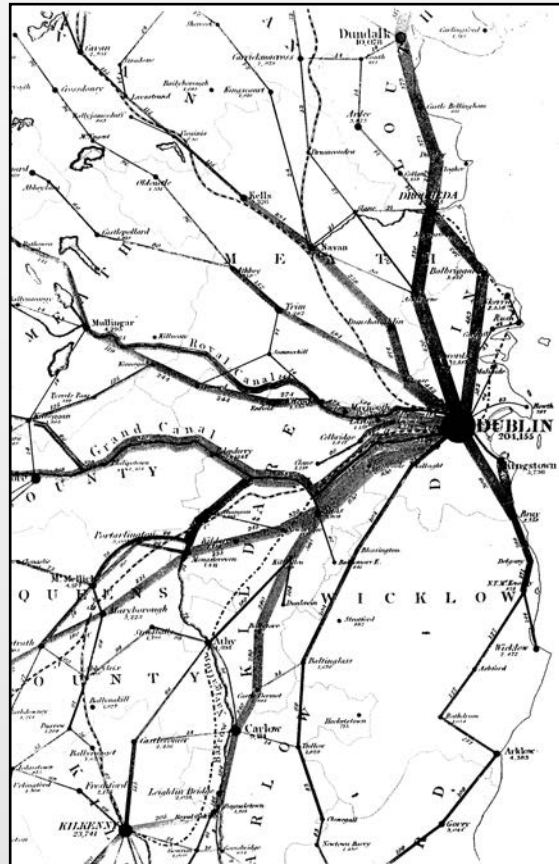
- Here Bigness In Modelling Means Building Our Model For Everywhere and Everyone

Our model is a LUTI model for England and Wales

- Let Me Explain What We Are Doing

# Let Me Begin with Some Examples of Big Data:

Dublin 1837, Ireland 1888, London 1955



Harness, 1837



Ravenstein 1888

blog.bigdatatoolkit.org

# BigDataToolkit



## Big Data Problems have been around longer than you think

The Strata Conference is in town and one presentation that caught my eye was titled The Great Railway Caper: Big Data in

big data, data processing, problems, shortest path

**Read More**

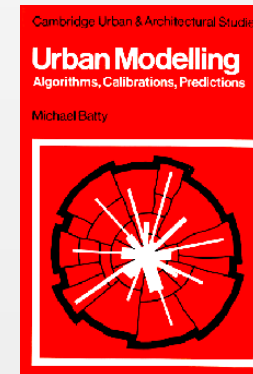
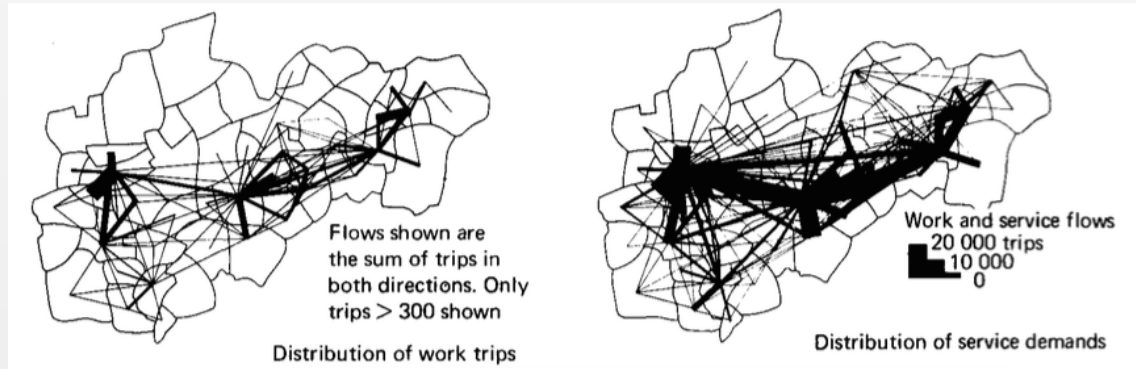


<https://www.youtube.com/watch?v=pcBJfkE5UwU>



# Always Had Big Data: Visualising Flows

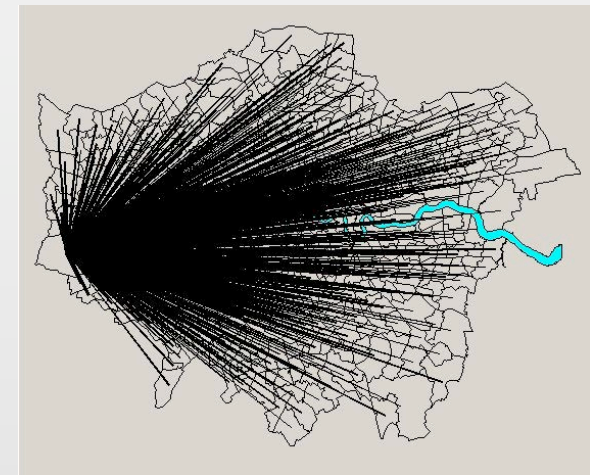
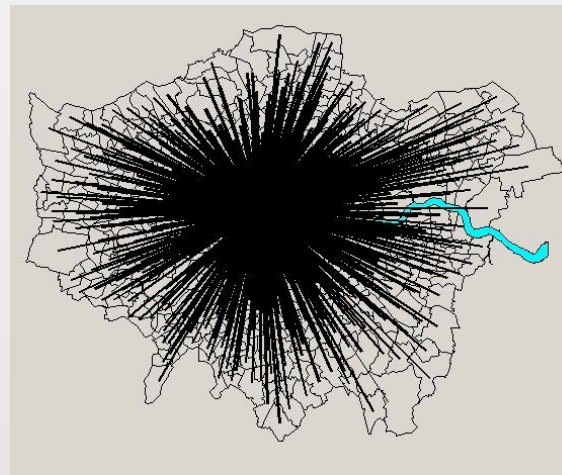
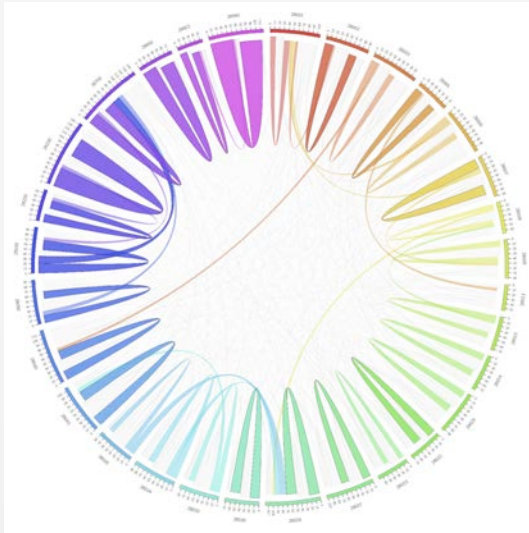
An early model circa 1967-8 Central and NE Lancs



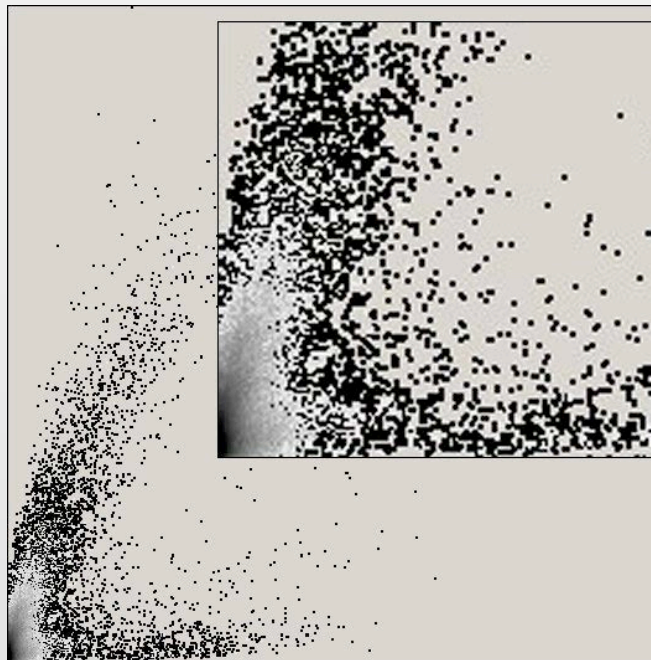
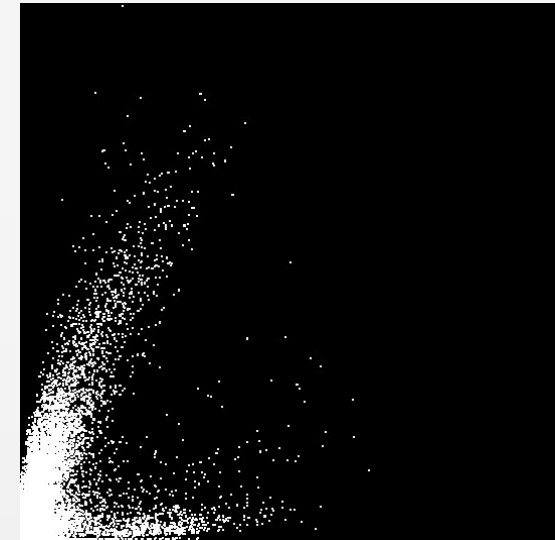
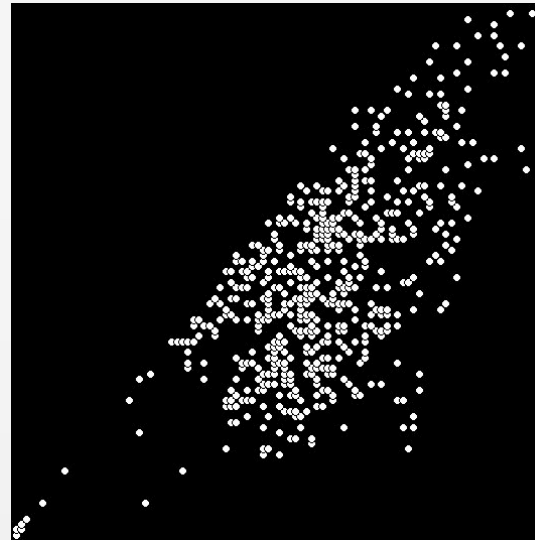
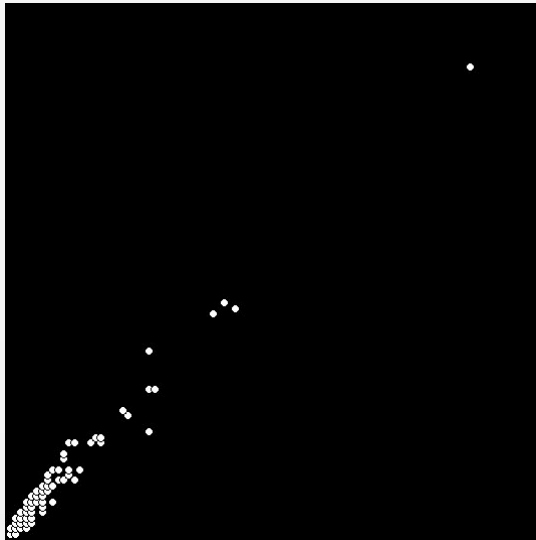
M. Batty (1976)  
**Urban Modelling**  
Cambridge UP



$n^2=33^2=1089$ , not so big but hard to visualise



$n^2=633^2=400,689$ , bigger but impossible to visualise



Even our statistics breaks down when we get large numbers like over several thousand as you can see on the left and above right for 400K data points where the pattern is highly convoluted. This is from a gravity model.

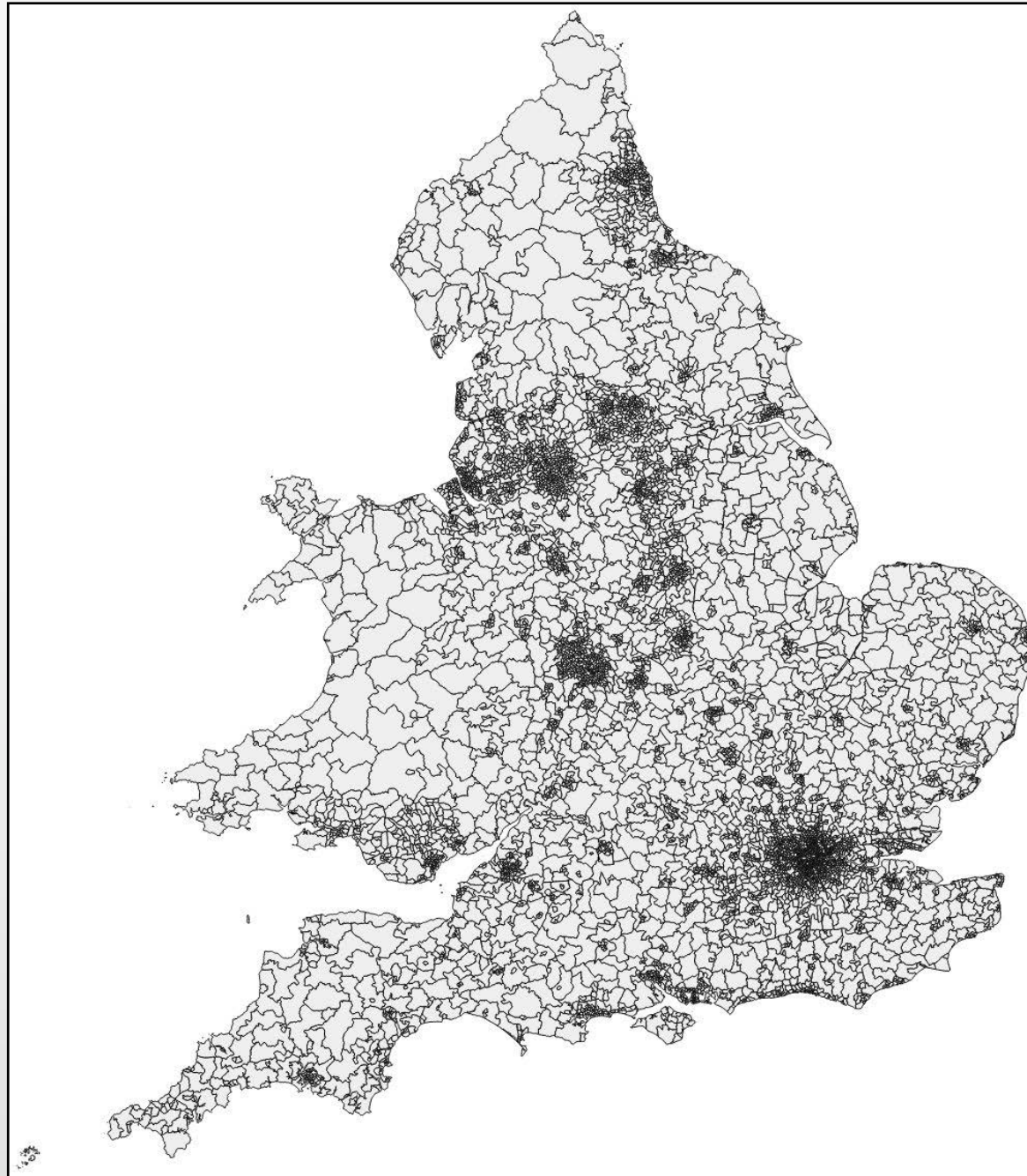


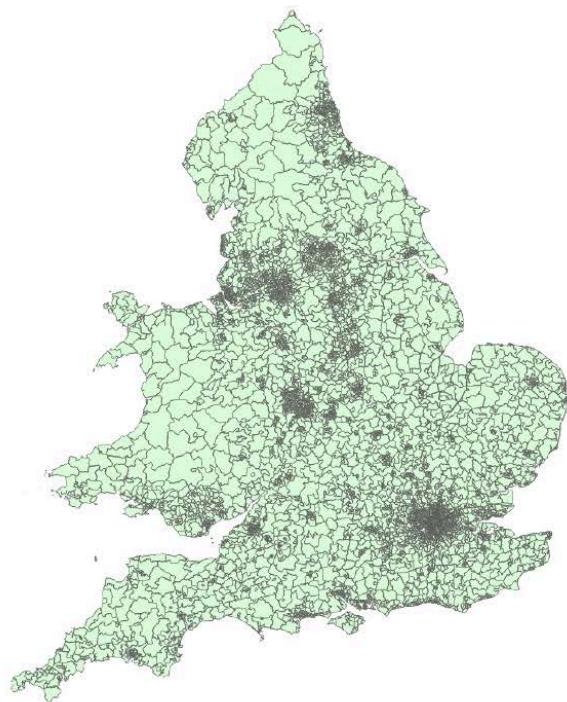
# Big Data Leads to Big(ger) Models

Now what happens when we really do scale up to the level of MSOAs of which there are 7201 in the UK – do we partition the spatial system and argue we don't need to scale up to  $n^2=7201^2=51,854,401$ ?

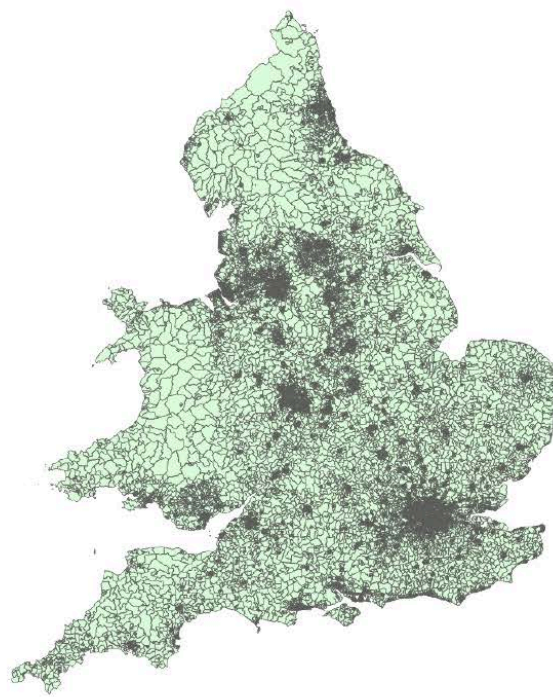
Circa 52 million points is an issue but our models run in a matter of seconds but that is a lot of data to store – ok it is sparse but sparsity isn't structured so we can't easily partition and in any case we want to compute any possible flow e.g. between, say central London and say Newcastle. Here is the problem scaled up and this is what we are grappling with at present.



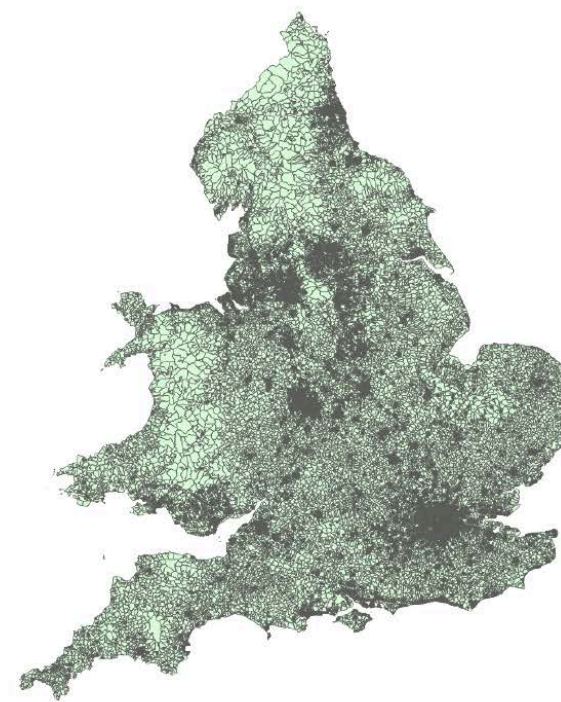




(a) MSOA (A=7201)



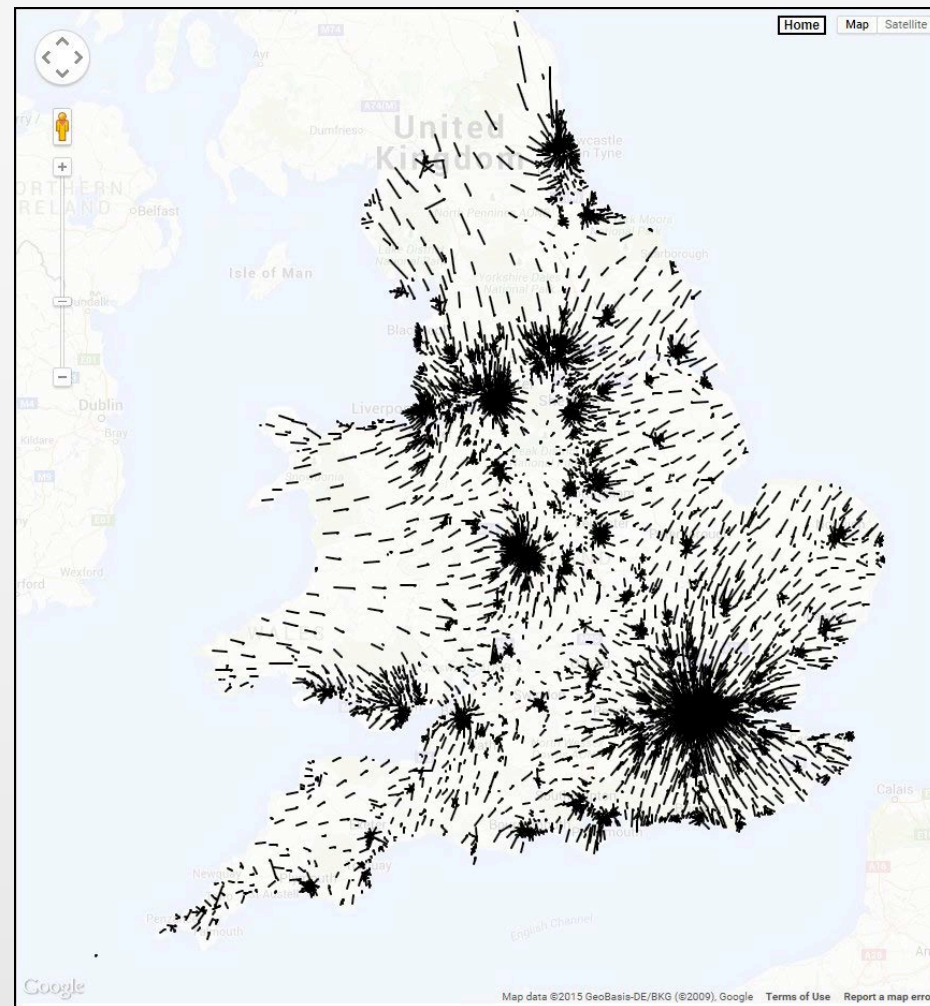
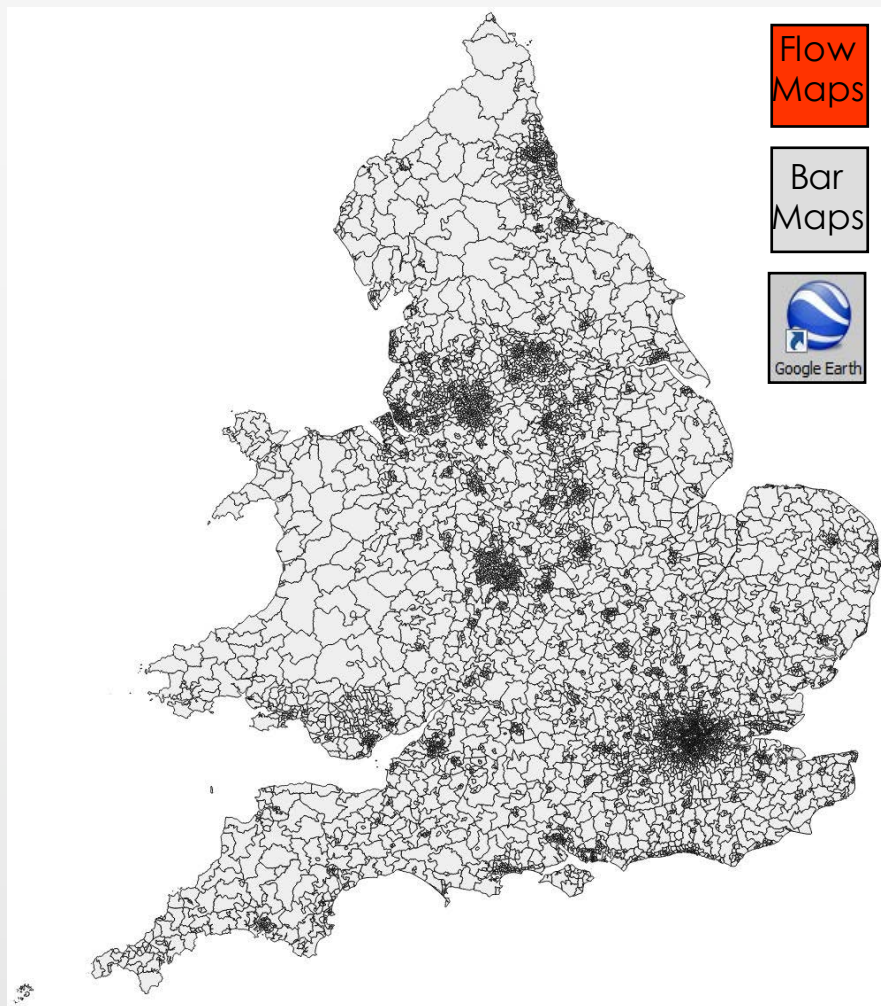
(b) LSOA (A=34753)



(c) OA (A=181408)

ONS Geographies for MSOA, LSOA and OA levels.





$$[x_i, y_i] = [x_i, y_i], \left[ \left[ x_i + \frac{\sum_j T_{ij} [x_i - x_j]}{n} \right], \left[ y_i + \frac{\sum_j T_{ij} [y_i - y_j]}{n} \right] \right]$$

## An Idea of the Model:

Essentially a couple of nested spatial interaction-discrete choice-like models – one for residential, the other for the retail sector, cross sectional static

$$\begin{aligned} T_{ij}^k &= O_i \frac{D_j^{obs} \exp(-\beta^k d_{ij}^k)}{\sum_k \sum_j D_j^{obs} \exp(-\beta^k d_{ij}^k)} & k=1, k=2, k=3 \\ & & road, rail, bus \\ P_j &= \sum_k \sum_i T_{ij}^k \\ S_{j\ell}^k &= P_j \frac{W_\ell^{obs} \exp(-\lambda^k d_{j\ell}^k)}{\sum_k \sum_\ell W_\ell^{obs} \exp(-\lambda^k d_{j\ell}^k)} \\ O_\ell &= \sum_k \sum_j S_{j\ell}^k \end{aligned}$$



# The Web and the Desktop: Users are also Data

We are building this aggregate LUTI model of the UK – well E&W at present – we will add Scotland before long – which is of the nature we have been implying –

Without going into details, the model takes a few seconds to run – it will take a lot longer when finished as we will add sectors and of course the number of big data we have to hold in RAM might be very large – currently we need to hold 3 x 2 such 52 million sized matrices – we may need to go up to multiples of this and that will involve a lot of packing and moving in and out of core, I think

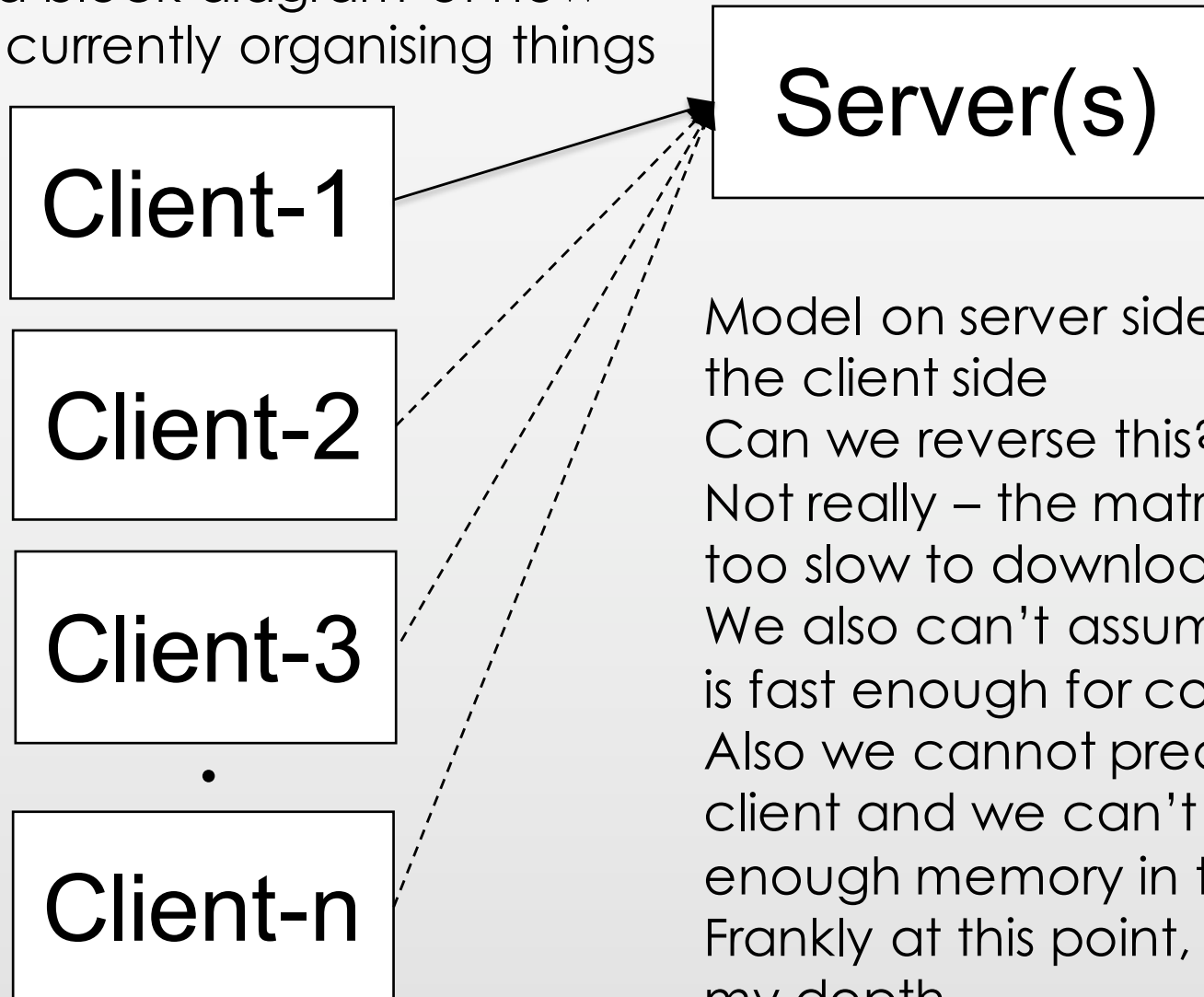
But the real issue is users – if our model is this large, and we have many users, then our data problem is exploded by the users –

Our big data is our original and predicted data from the model, times the number of users. Why are users data ? Well because they are using data differently – they are making their own predictions and thus scaling up the data.

We could have one model for each user but we don't know who the users are? We thus want them to access this on the web.... Let me see if I can demo it from here – in fact I am hoping you have some web links .....if you haven't then you can do it on your phone and then it really does look dreadful –

# The Web and the Desktop: Users are also Data

Here is a block diagram of how we are currently organising things



Model on server side; Maps on the client side  
Can we reverse this?  
Not really – the matrices are too slow to download to client?  
We also can't assume the client is fast enough for computation.  
Also we cannot predict the client and we can't assume enough memory in the client  
Frankly at this point, I am out of my depth

# So Let Me Give You A Demo Of What We Are Doing

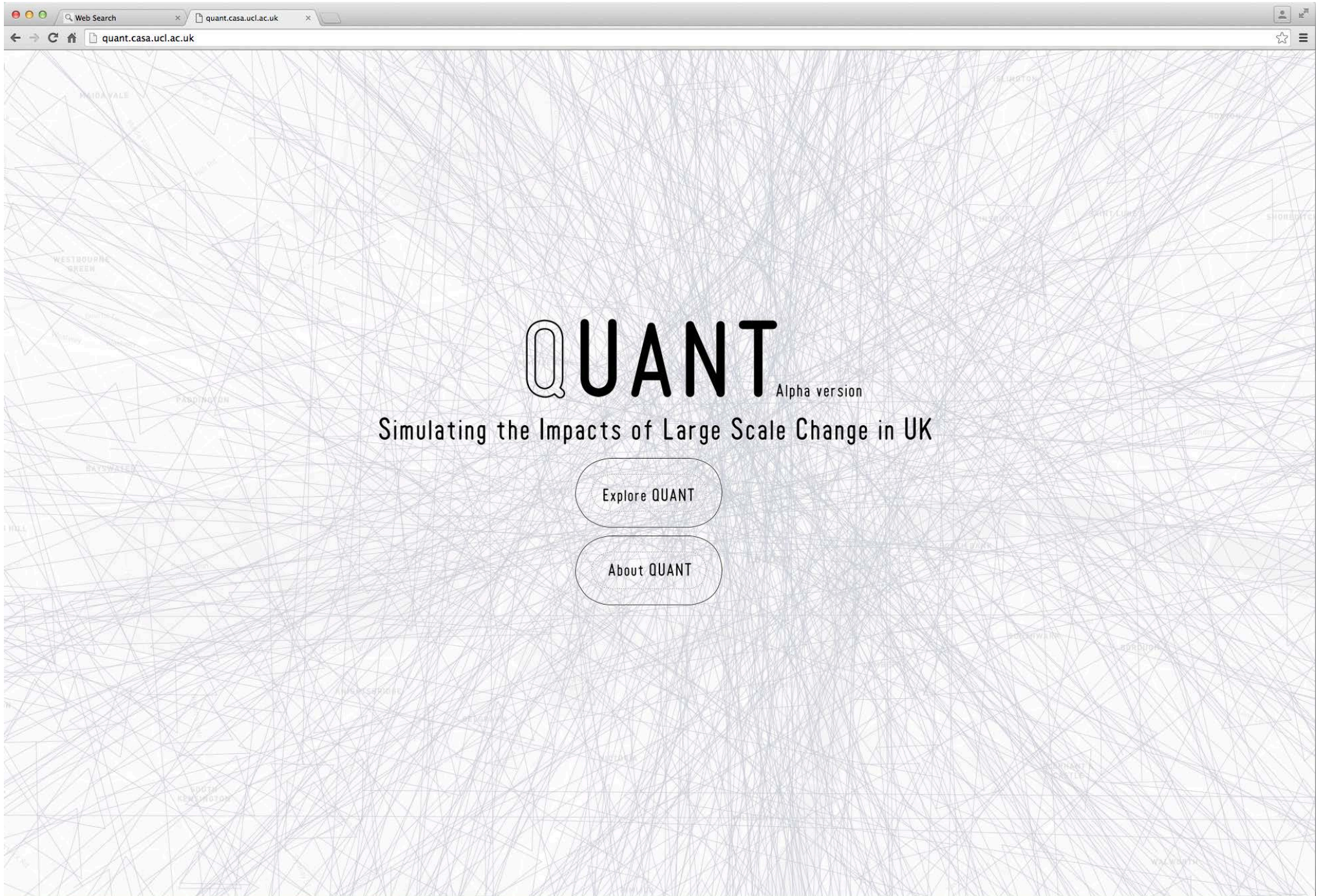
The web link is

<http://quant.casa.ucl.ac.uk/>

For those who want to try it out

There is always ***a caveat emptor*** with demo models like this. We are building it as a demo for the FCC and it not supposed to be a flashy app or anything – it is something that local authority planners and consultants might use as a resource along with many other complementary decision support tools







Web Search quant.casa.ucl.ac.uk

quant.casa.ucl.ac.uk

## About

QUANT release 5 simulates the impact of changes in population, employment, and travel costs associated with movements on the transport network in UK Cities.

QUANT uses a simple model of how workers choose the places where they live with respect to attractive those places are and the travel costs from their workplaces.

QUANT visualises employment, working population, and journeys to work from the 2011 Population Census and then compares these with predictions from the model. The process of running the model and making comparisons with what we observe is called calibration and this fine tunes the model to simulate the data as closely as possible

## How to use QUANT?

### Choose your location of interest...

You can choose an area of the country by pointing at the map or default to the entire country.

### Explore the Data...

You can visualize the data as a series of maps and/or other graphical outputs which you can load in any order.

### Run the Model...

We first ask you to initiate the calibration of the model and this then happens automatically. The travel parameter is then fixed and this ensures that we get a good match between the observations and the model predictions. After the model has been calibrated, you can see the outcome in a series of maps and graphical outputs that mirror those you have used if you explore the data

Simulating the impact of Large Scale Change in UK

About QUANT



# QUANT

Alpha version

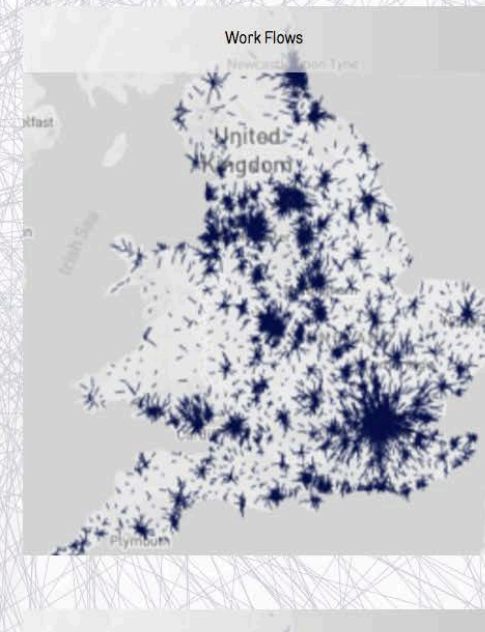
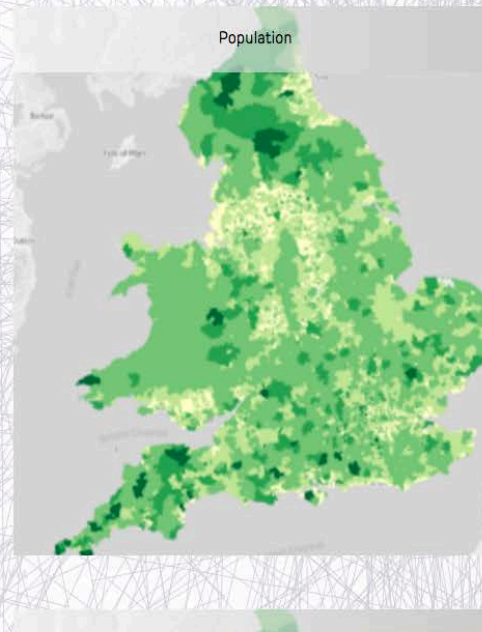
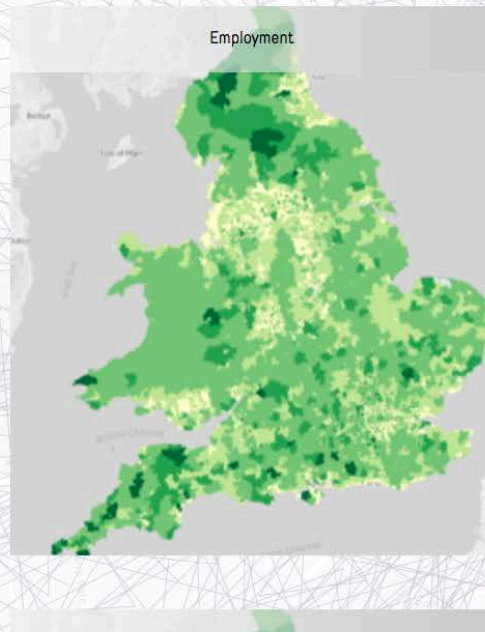


Explore Data

Run Model

Set Scenarios

Home





# QUANT

Alpha version

About

CATAPULT  
Future Cities

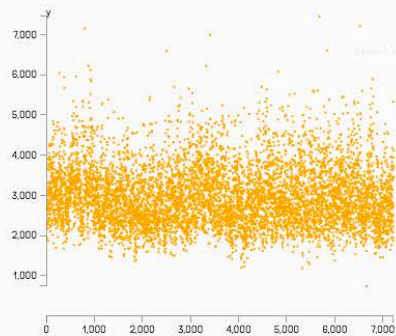
CASA

## Observed Population



This is the total working population in each zone of the model. This is the employment which lives in residential zones.

Ranked



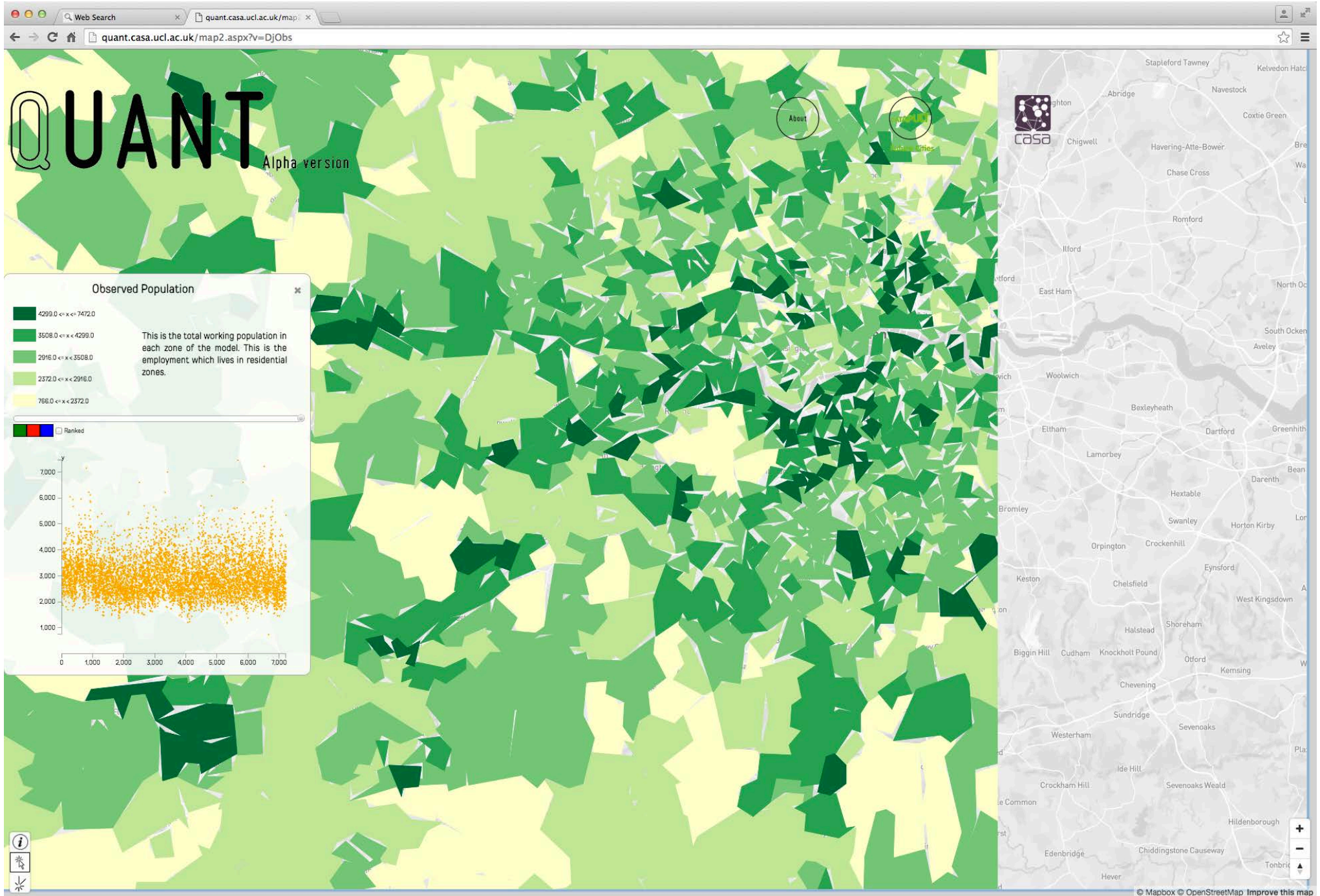
© Mapbox © OpenStreetMap Improve this map



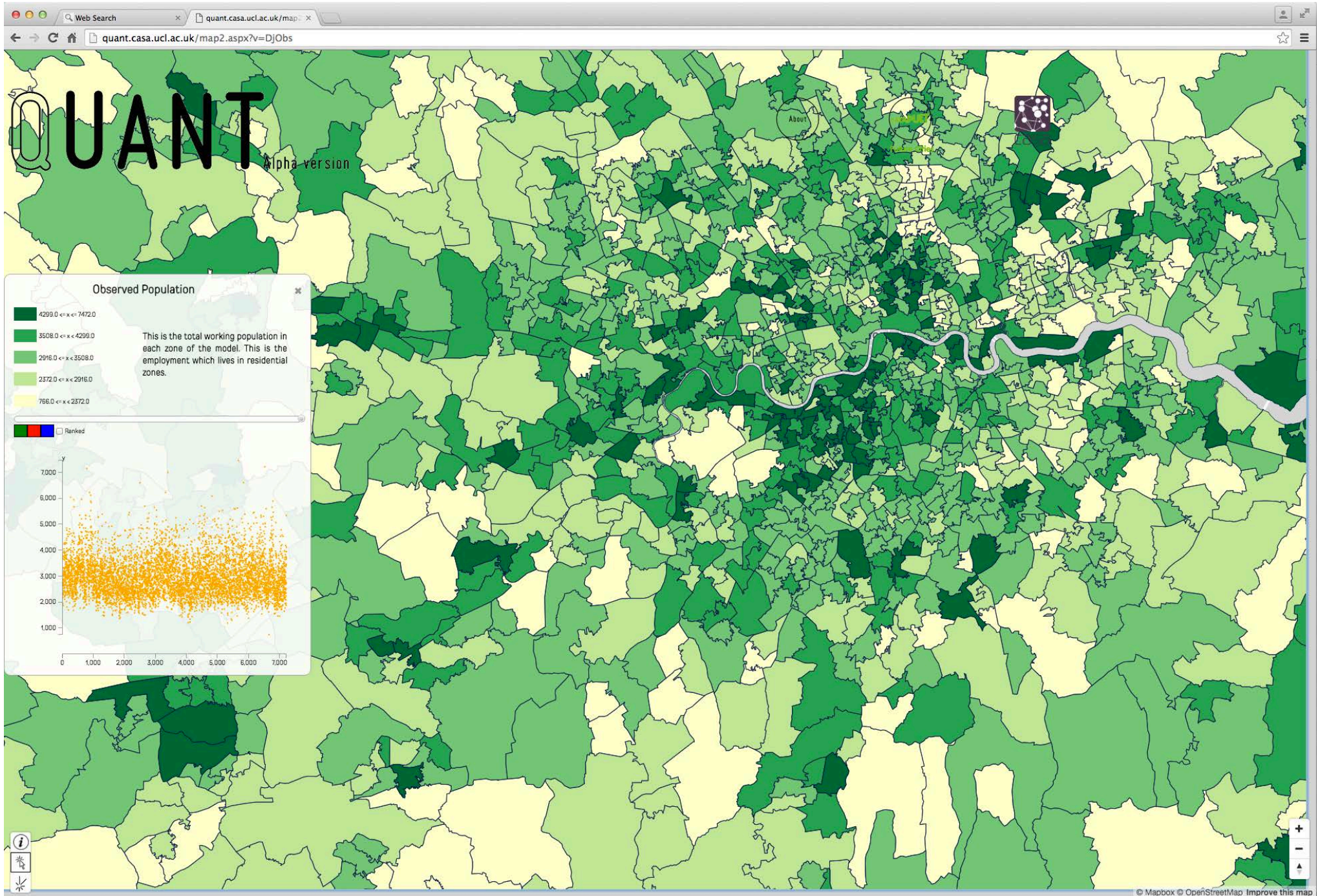
Centre for Advanced Spatial Analysis



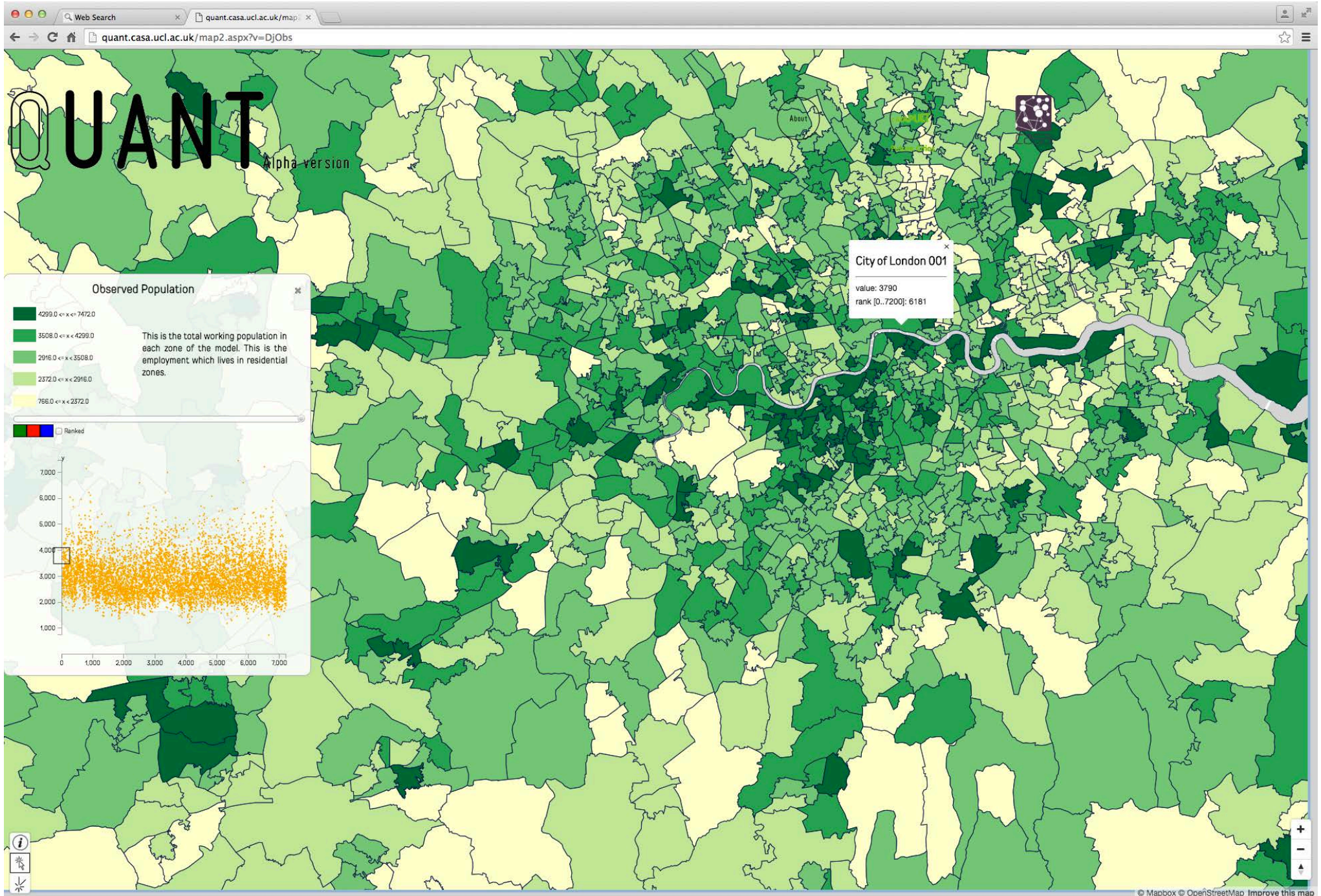








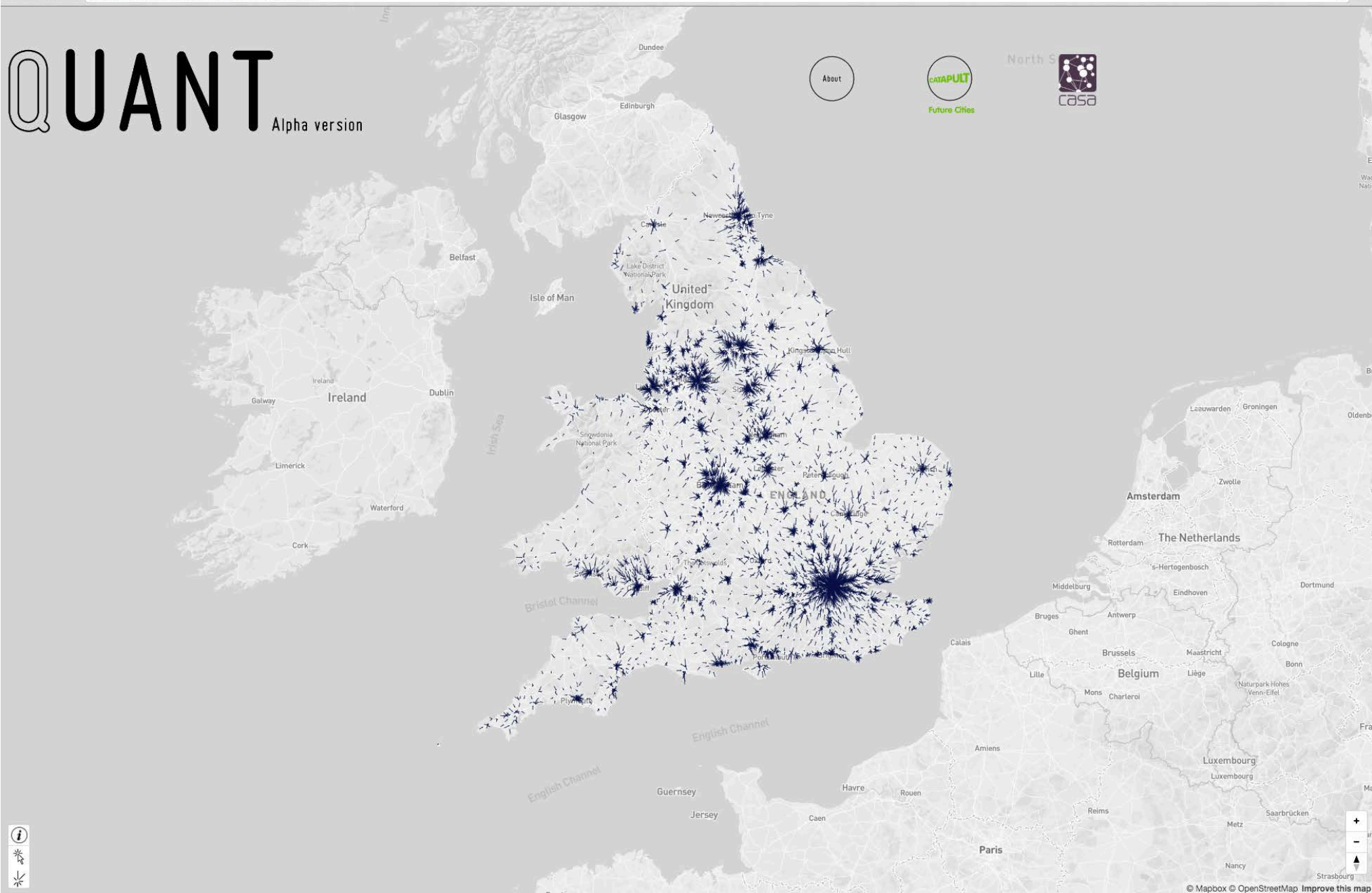






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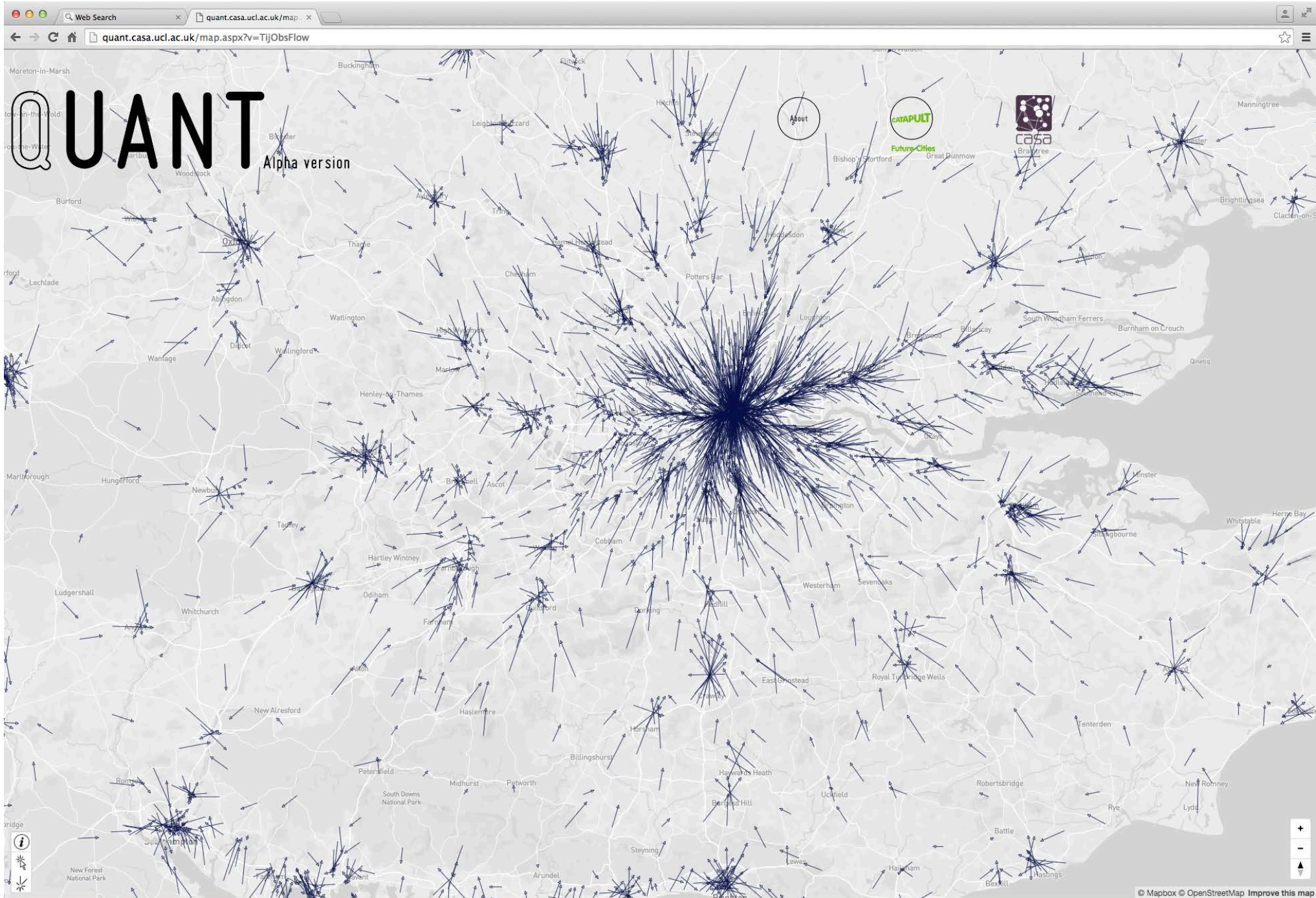


© Mapbox © OpenStreetMap Improve this map

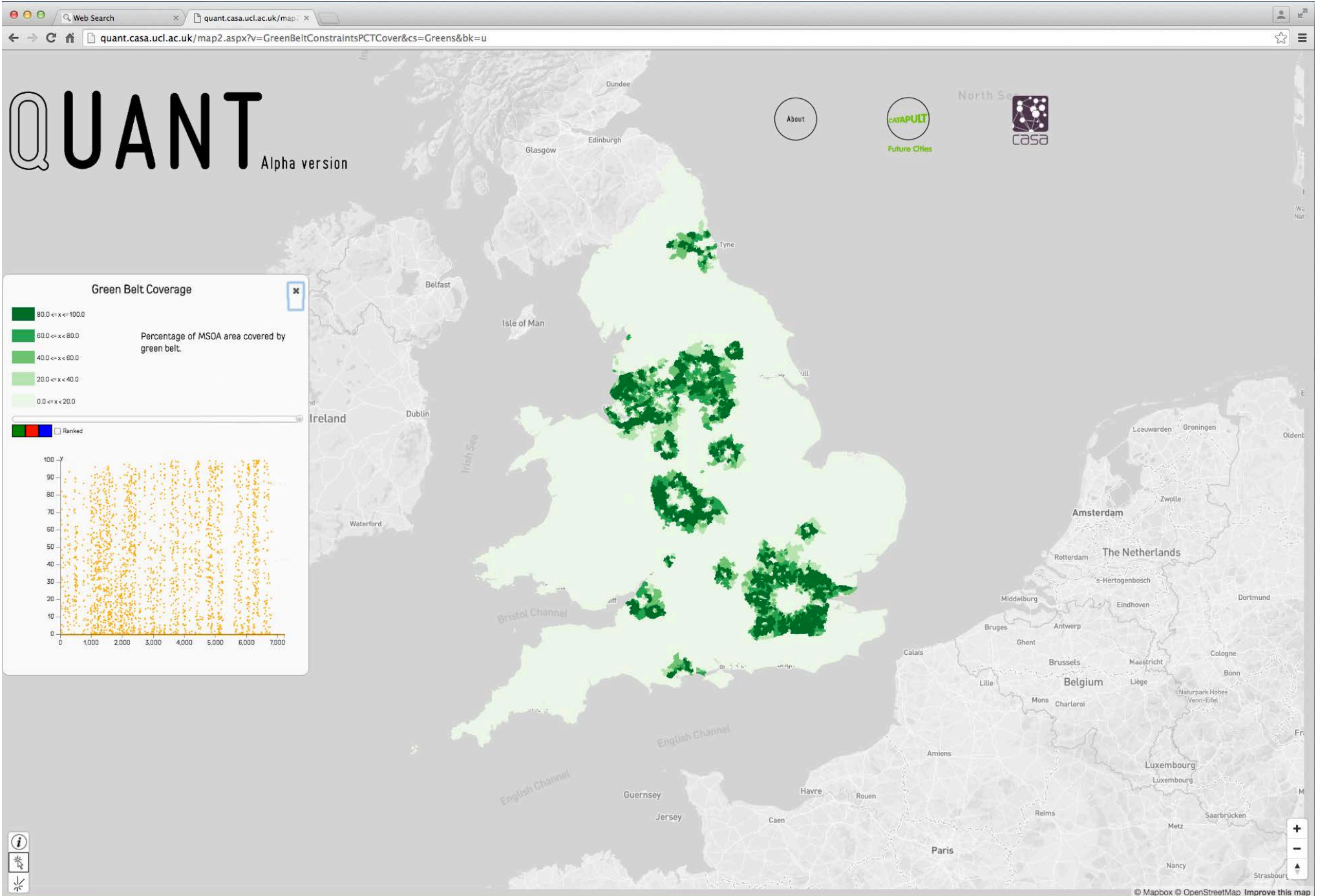


Centre for Advanced Spatial Analysis

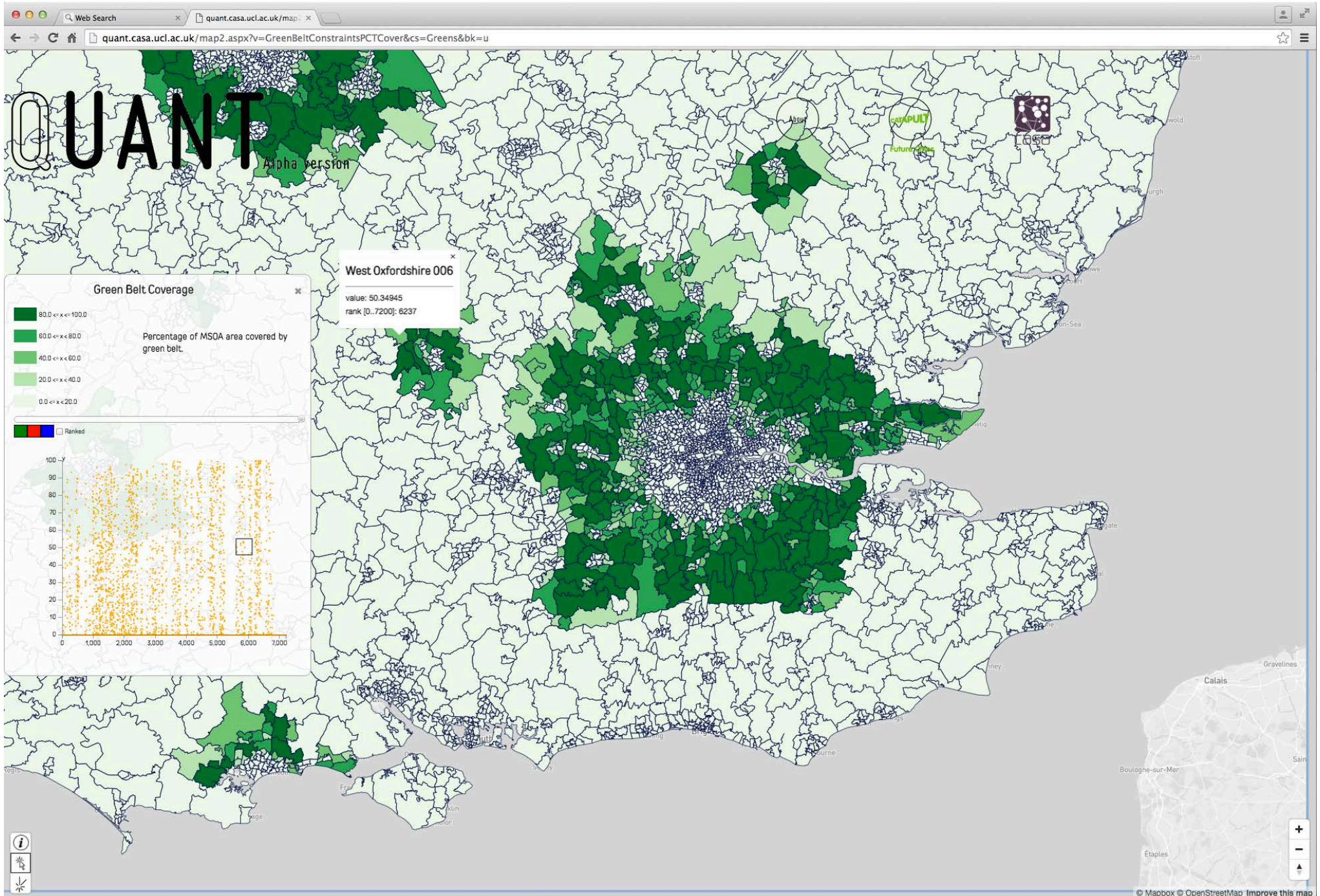




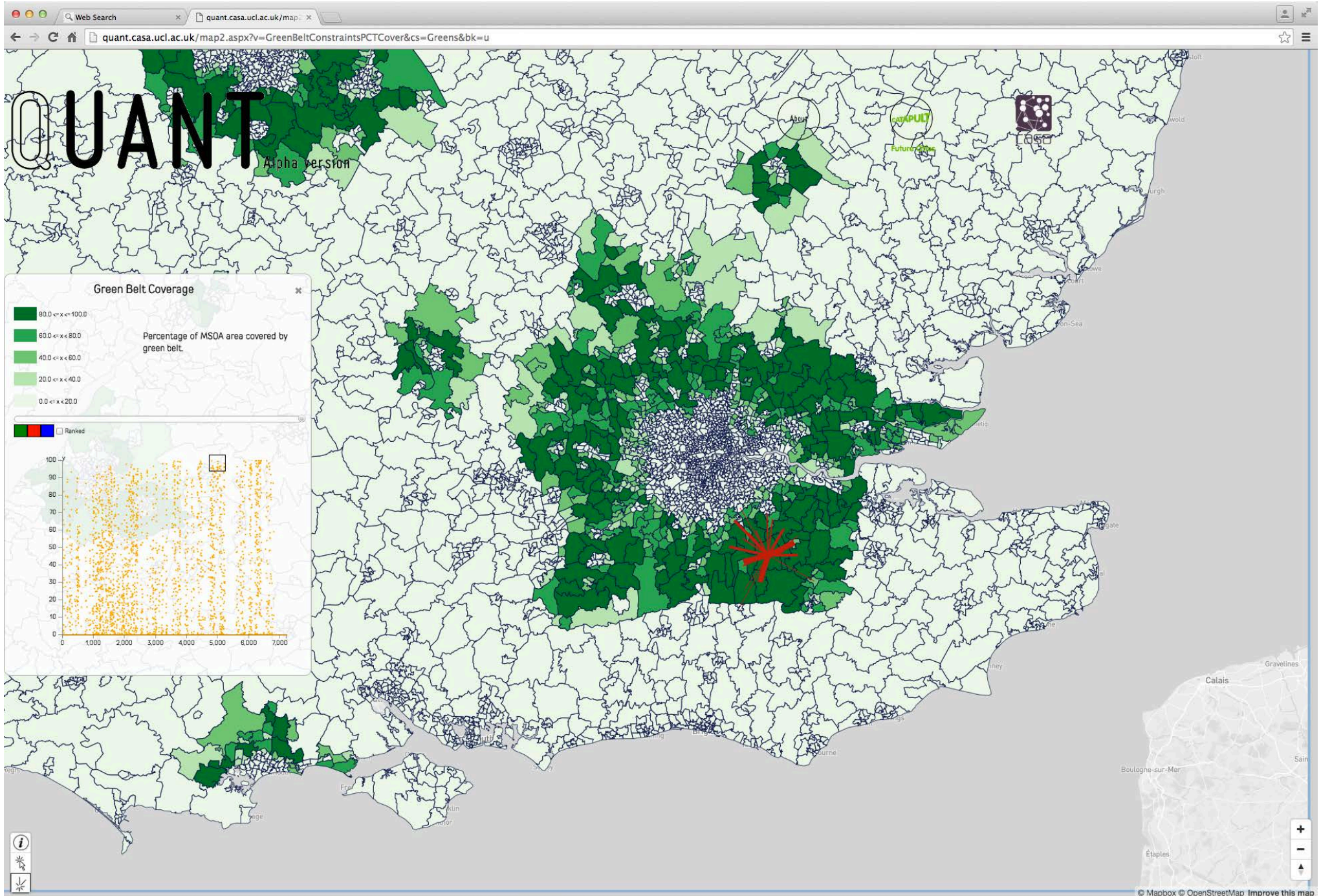








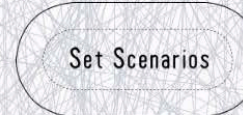




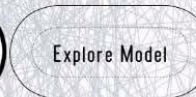


# QUANT

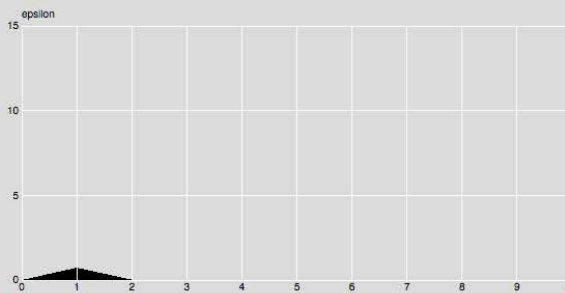
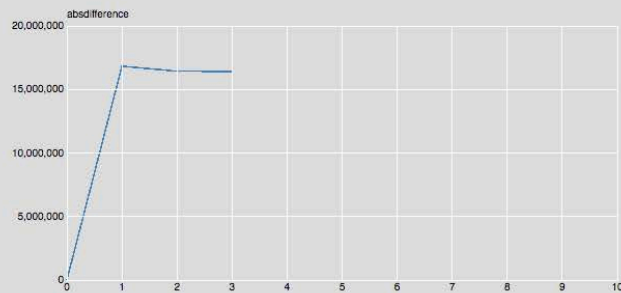
Alpha version



Before Running the model please calibrate the travel parameter to make sure the model is a good fit.



Finished



# QUANT

Alpha version

Explore Data

Run M

enarios

Home

Before  
para

travel  
fit.

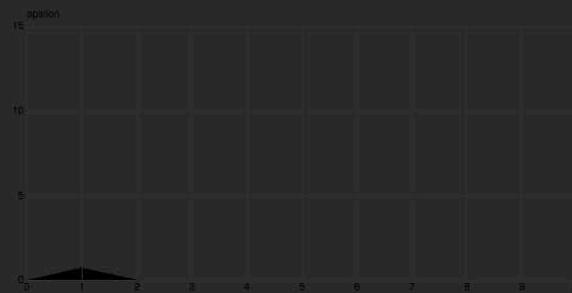
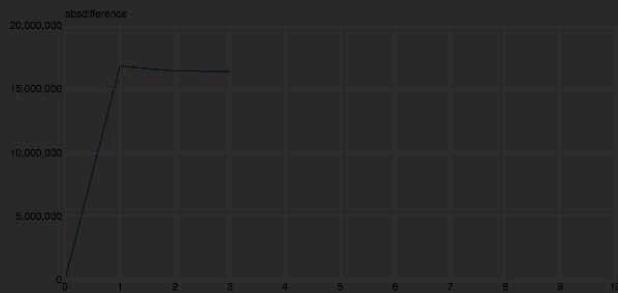
Con

odel

## Model Statistics

Statistic or Parameter	Value
Beta Distance Parameter	0.105535313
Total Integer Differences	0
Sparsity	1
Observed Mean Trip Length	14.8332882
Predicted Mean Trip Length	14.9799633
Total Trips, Total Population	21625060
Total Mean Absolute % Population Difference	0.238348529
Total Mean Absolute % Flow Difference	3.428195E+20
Mean Observed Population	3003.06348
Mean Predicted Population Densities	3003.064
Mean Observed Destinations	12.5258255
Mean Predicted Destinations	14.8611374
Mean Observed Trips	0.417034239
Mean Predicted Trips	0.4170357
Correlation Observed Predicted Destinations	0.7260883
Correlation Destinations	0.965938568
Correlation Trips	0.8063401
Sorenson-Dice Index Population	0.8929917
Sorenson-Dice Index Population Density	0.954629
Sorenson-Dice Index Flows	0.61706

Finished





# QUANT

Alpha version



Explore Data

Run Model

Set Scenarios

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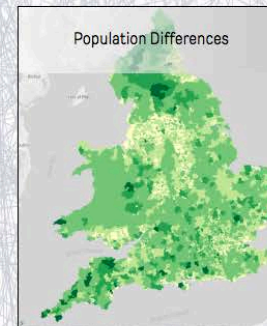
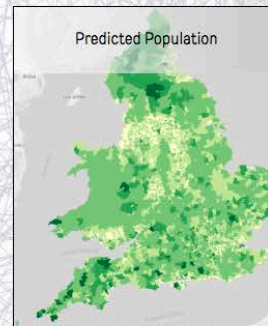
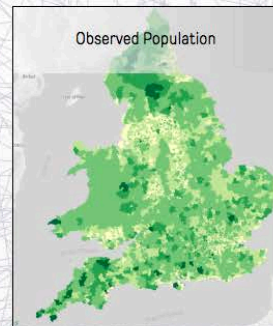
Maps

Graphs

Charts

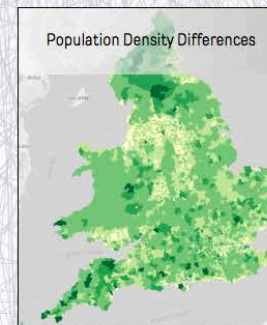
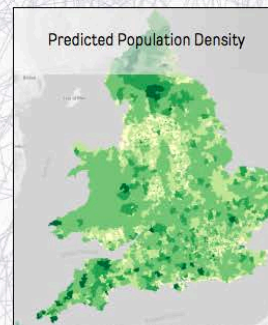
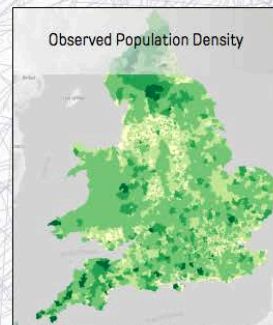
**Population**

This is total working population in each zone of the model. This is the employment which lives in residential zones.



**Population Density**

This is working population divided by the land area of the residential zone.





# QUANT

Alpha version



Explore Data

Run Model

Set Scenarios

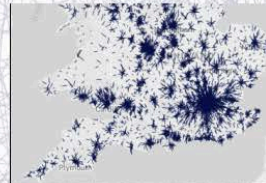
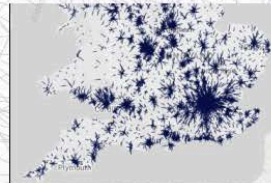
Home

Maps

Graphs

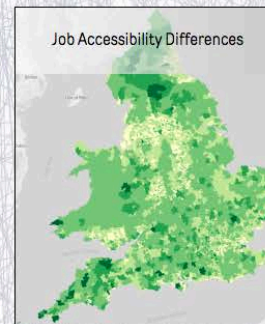
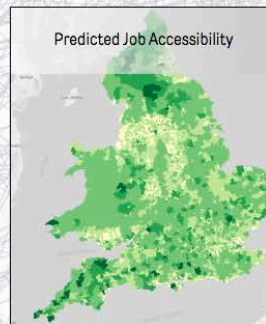
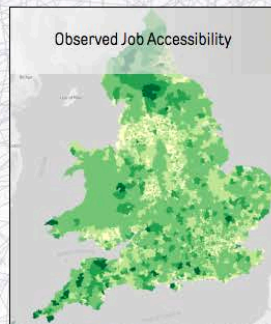
Charts

trips from the residential zone to the employment zone.



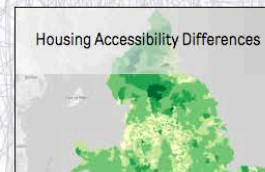
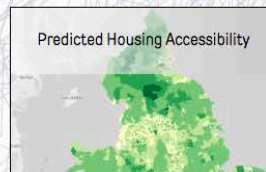
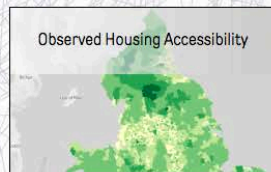
Accessibilities Origins

This is the accumulated weighted distance from each individual workplace zone to all residential zones.



Accessibilities Destinations

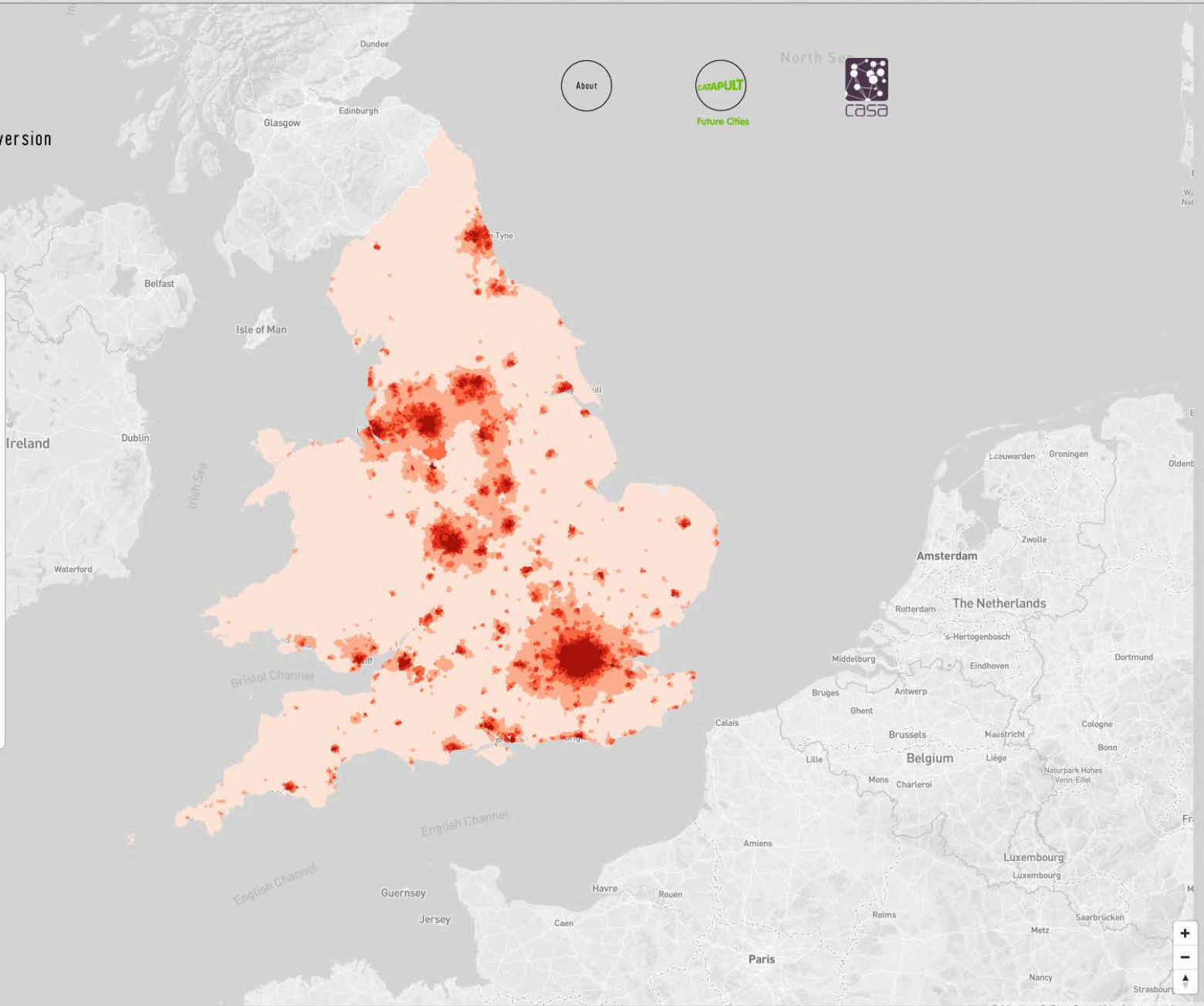
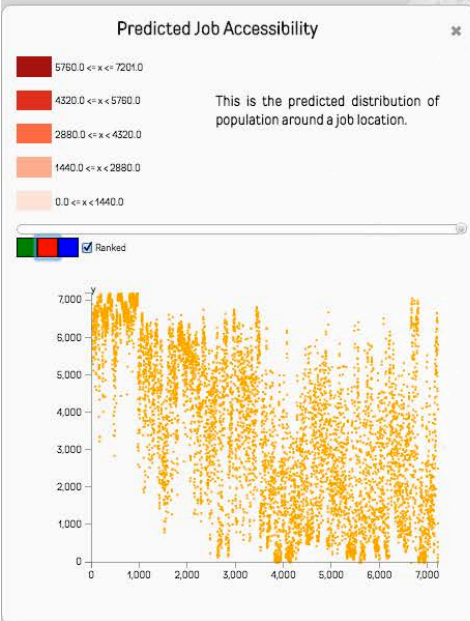
This is the accumulated weighted distance from each individual residential zone to all workplace zones.





# QUANT

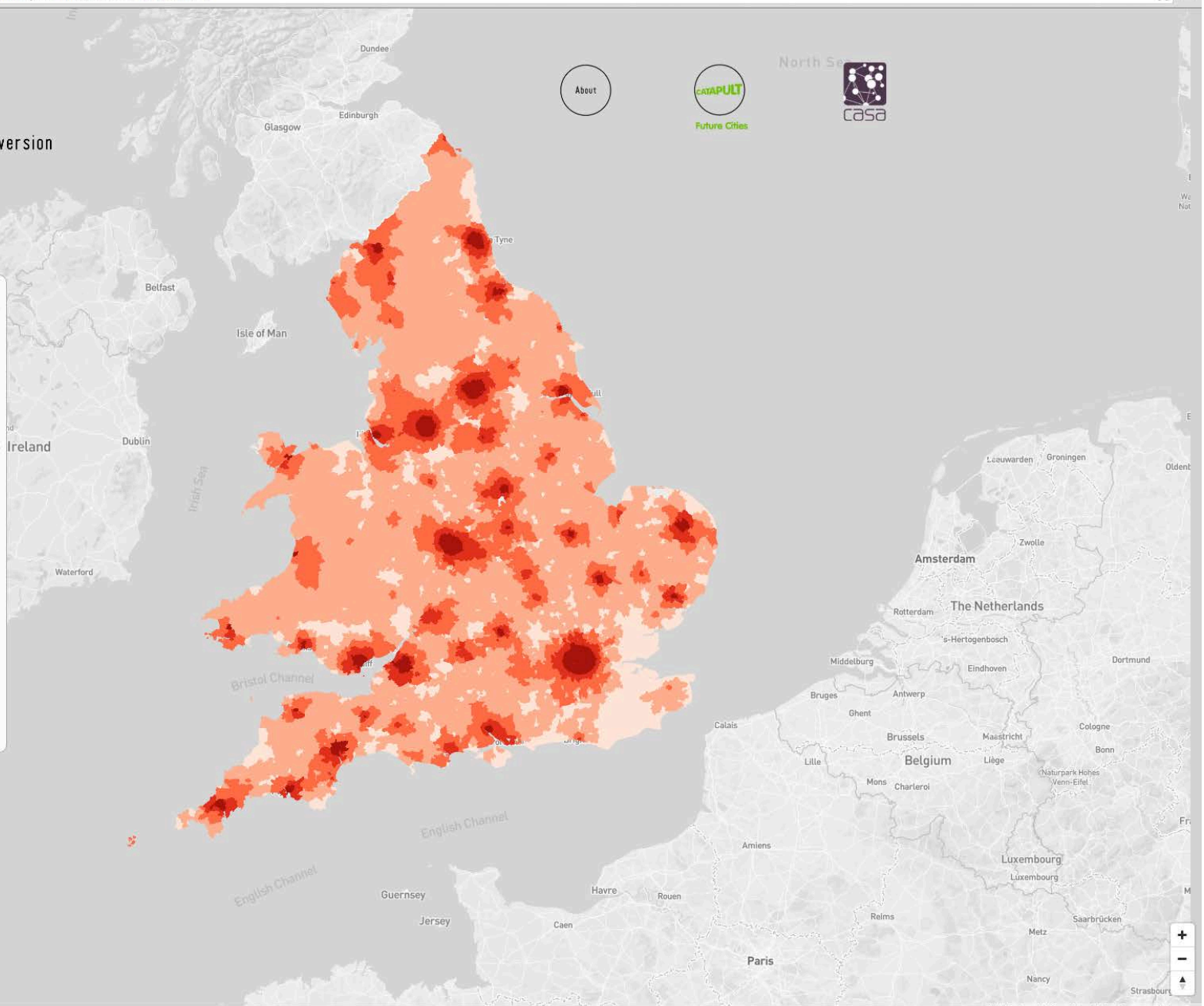
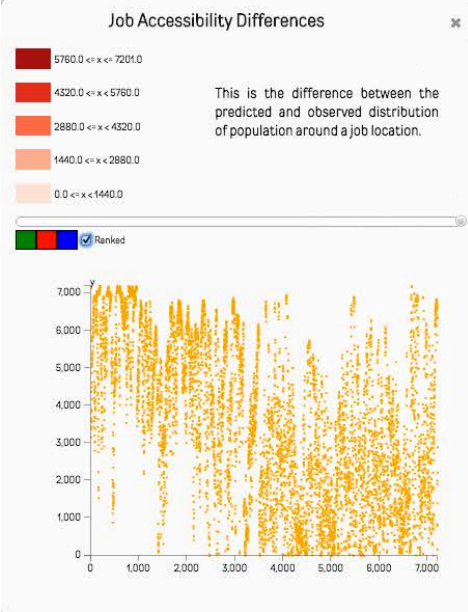
Alpha version

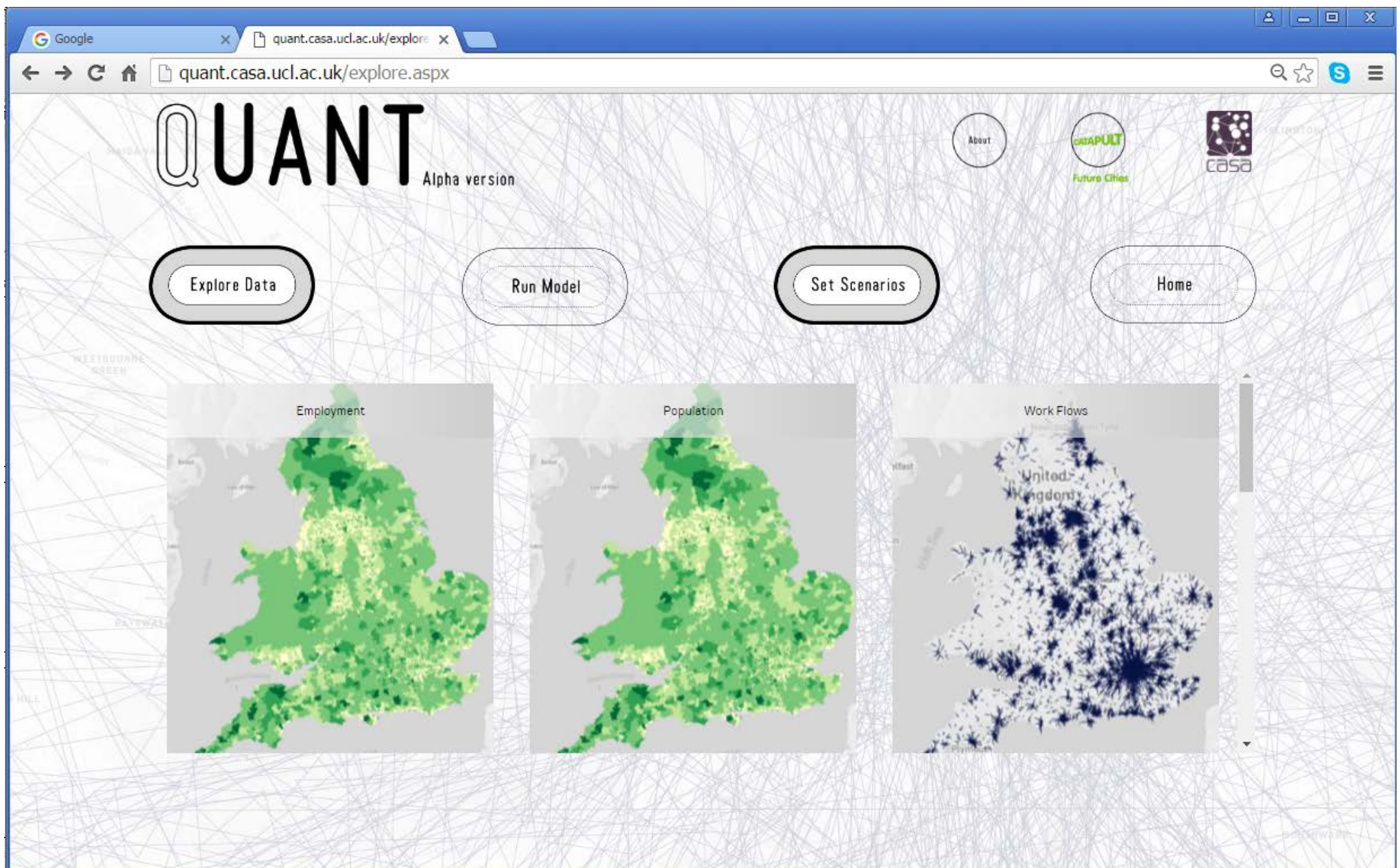




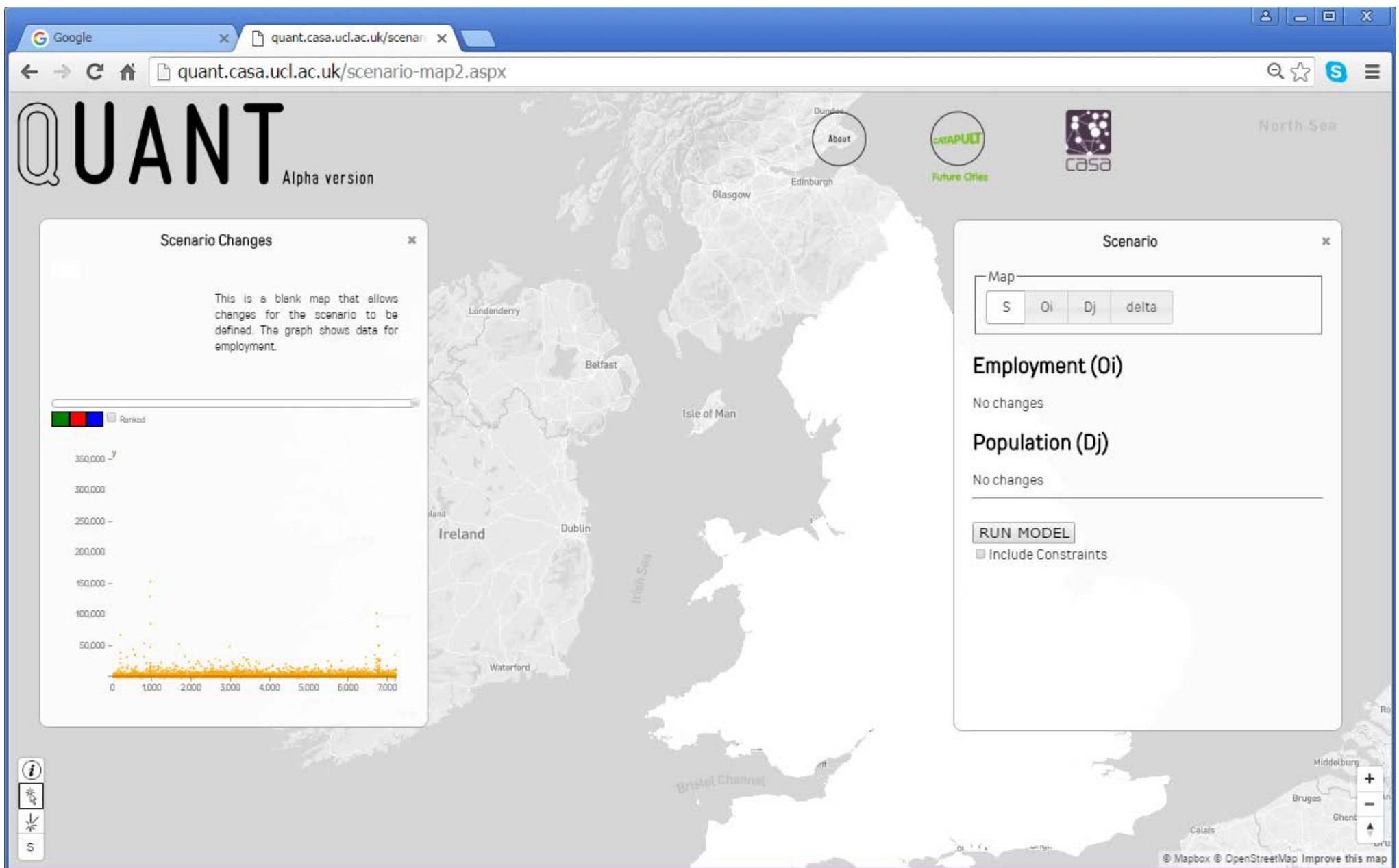
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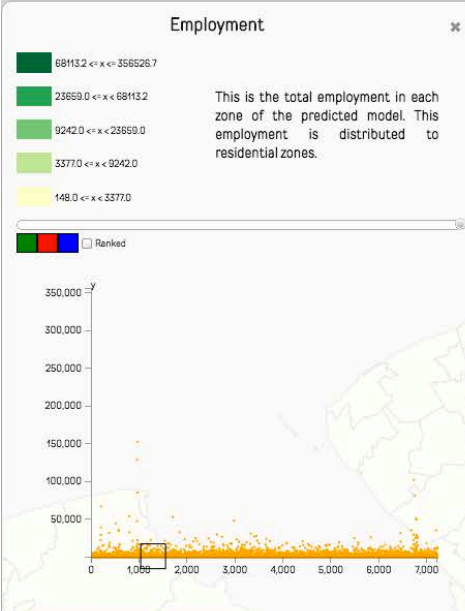






# QUANT

Alpha version



**Knowsley 008**

value: 736.9992  
rank [0..7200]: 992

	Obs	Pred	Scenario
Oi	737	737	737
Dj	1514	1765	1765

Oi: 737  
Dj: 1514

**Scenario**

Map

S Oi Dj delta

**Employment (Oi)**

No changes

**Population (Dj)**

No changes

**RUN MODEL**

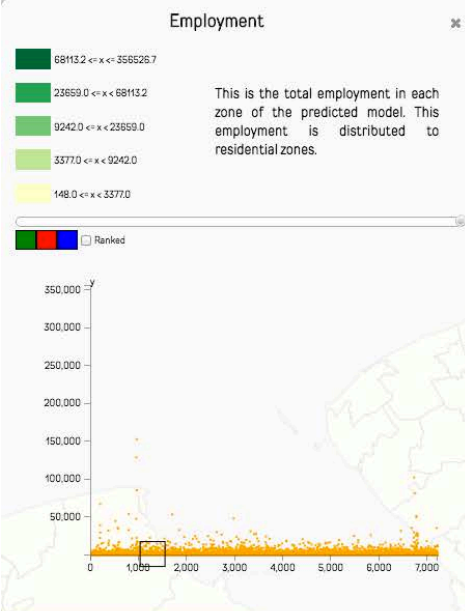
☐ Include Constraints





# QUANT

Alpha version



### Knowsley 008

value: 736.9992  
rank [0..7200]: 992

	Obs	Pred	Scenario
O:	737	737	737
D:	1514	1765	1765

O: 40000  
D: 1514

### Scenario

Map

S Oi Dj delta

#### Employment (Oi)

name	OiObs	OiPred	Oi	%	edit
Knowsley 008	737	737	40000	+5327%	<input checked="" type="checkbox"/>

#### Population (Dj)

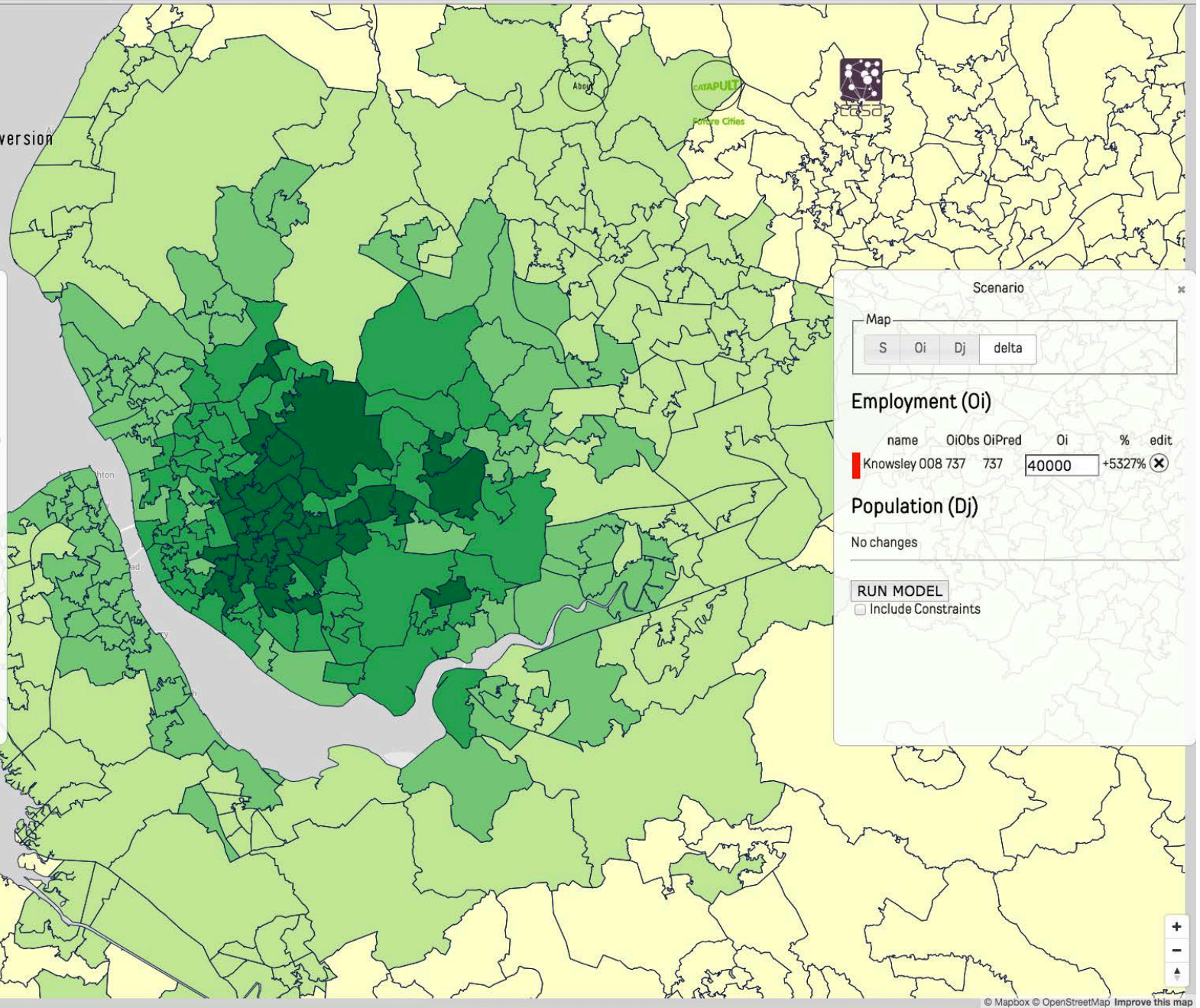
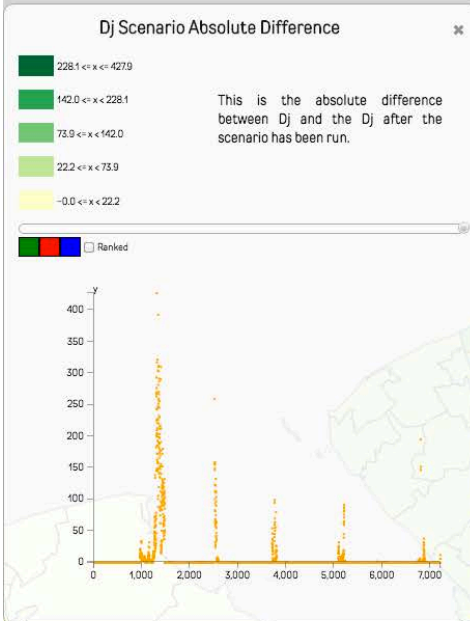
No changes

☐ Include Constraints



# QUANT

Alpha version



### Scenario

Map

S Oi Dj delta

#### Employment (Oi)

name	OiObs	OiPred	Oi	%	edit
Knowsley 008 737	737	40000	+5327%	<input checked="" type="checkbox"/>	

#### Population (Dj)

No changes

**RUN MODEL**

☐ Include Constraints



# QUANT

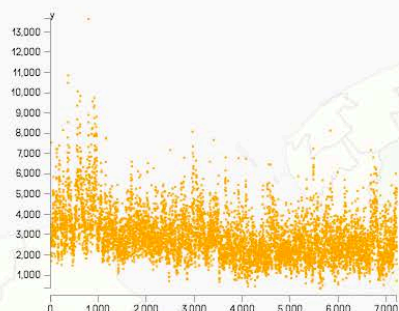
Alpha version

## Predicted Population



This is total working population in each zone of the model. This is the employment which lives in residential zones.

☐ Ranked



## Scenario

### Map

S Oi Dj delta

### Employment (Oi)

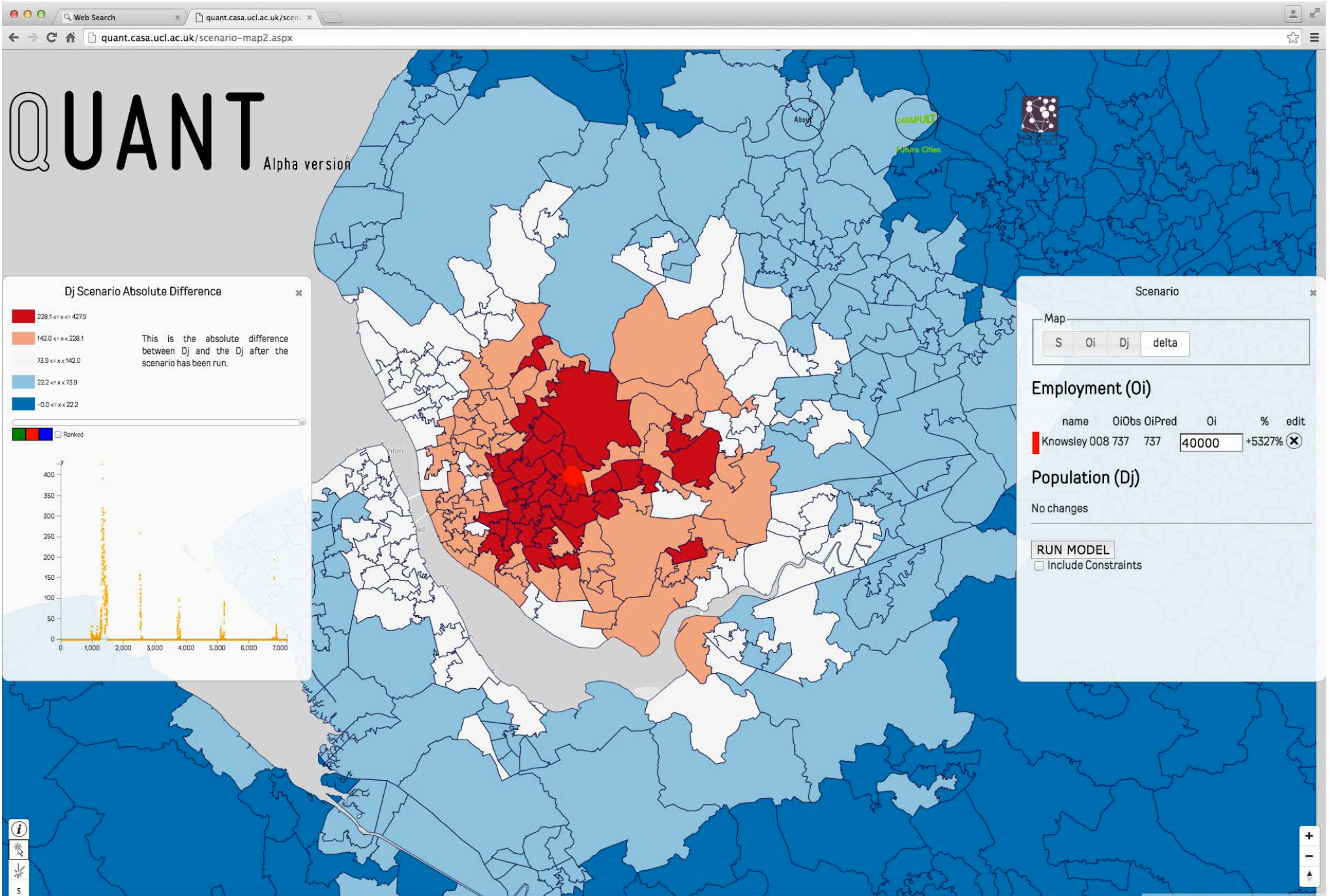
name	OiObs	OiPred	Oi	%	edit
Knowsley 008 737	737	40000		+5327%	<input type="button" value="X"/>

### Population (Dj)

No changes

☐ Include Constraints





# Conclusions and Next Steps

We need to do many things and the most important is working with decision-makers – people who will use these tools.

Our view is that the process of translating these models into tools is one which is lengthy and involved – more involved than building these models *per se* and we have only just begun to recognise these

The extent to which these models needs to be known by those who use them is still a crucial issue – in fact a more important issue than ever we realised before.



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<http://www.complexcity.info/>

<http://www.spatialcomplexity.info/>

<http://blogs.casa.ucl.ac.uk/>