# **Smart Cities**

Session 2: Lecture 2:

Material and Electronic Networks: Transport

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#### **Outline of the Lecture**

- 1. Context: New Data for and from the Smart City
- 2. Multimodal Flows: An Analytical Gap
- 3. Why London?
- 4. A Partial View of London's Network
- 5. Three Approaches, Three Problems –
- 6. Problem 1: Flows on the tube lines by trains
- 7. Problem 2: Representing Networks
- 8. Problem 3: The 'Shortest'-Path Problem





## Context: New Data for and from the Smart City

- Here we are going to move directly to networks in the literal sense, Remember Metcalfe's Law, so we will look also at flows on networks
- Our understanding of urban flows tends to be based almost exclusively on single-mode systems and each mode is handled separately.
- Reflects a widespread focus on auto-mobility, Also a function of analytical tractability. But the increasing availability of large behavioural data sets from Public Transport Networks (PTNs), combined with the increasing power and sophistication of computational approaches, creates new ways of exploring travel demand.
- Greater spatial resolution down to station or, soon, bus stop, and greater temporal resolution down to the minute, and greater coverage centralised collection of data means a single data store for an entire city
- PTN-derived data also avoids many of the privacy issues associated with mobile network data because the user becomes invisible as soon as they leave the system.





# Multimodal Flows: An Analytical Gap

	Pros	Cons	Representative Research/Resources
Graph Theoretical	Quick Full abstraction	Static Load-less Too simplistic	Von Ferber et al. (2007, 2010) Derrible (2010, 2012) Erath et al. (2009)
Agent-Based & Other Dynamic	Realistic Dynamic Capacity constrained	Slow Limited abstraction Can be too complex	Cats & Jenelius (2012) Manley et al. (2012) MATSim

Is there is an opportunity here to leverage digital data in order to drive more realistic models that combine the performance of graph theoretical approaches with the added realism of ABMs?





## Why London?

Transport for London's (TfL) RFID-based 'Oyster Card' is particularly attractive because users typically need to use their card at both ends of a trip, providing us with detailed origin and destination data for more than 3 million daily users.

The system is particularly large and complex:

Approximately 640 stations across all modes

340 stations with Oyster Card readers served by National Rail trains

80 stations served by Overground trains

270 stations served by Underground trains

45 stations served by Docklands Light Rail

39 stations served by Tram

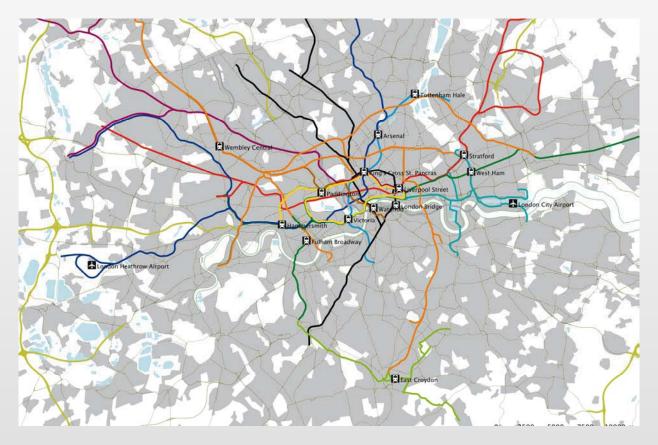
147 stations with some kind of interchange (between line or mode)

Aboveground coverage by Open Street Map (OSM) is also particularly good, allowing us to model walking behaviour using open source tools that respect pedestrian preferences for balancing directness with quieter streets.





#### A Partial View of London's Network



Although many users – especially visitors – are used to thinking about London in terms of the Beck schema, the combination of an online Journey Planner and regular travel on the network enable many to identify the quickest route *even if* it doesn't appear to be the most direct.





## Three Approaches, Three Problems

- 1. To explore what is happening to actual movements of trains to compute delays currently Transport for London TfL have an API for tubes and buses that enables the user to query the location and of all trains/buses on the network and to examine how they move from this we can calculate delay and compute a variety of measures. This is what Richard Milton, one of my colleagues has done and I will show some of this it is very preliminary
- 2. Use of classic graph measures to show how the network can be disrupted this is largely a topological graph/network approach that shows how betweenness centrality or accessibilities is disturbed I am responsible for this and will show some work
- 3. Fully fledged flow and graph measures in a multimodal context tube, overground, walk so far where we are computing changes in flow this work that <u>Jon Reades</u> is doing
- 4. All very preliminary I show all this in the spirit of work in progress





# Problem 1: Flows on the tube lines – by trains

As we will demonstrate, through the "Trackernet" system for London Underground and the "Countdown" system for buses, it is now possible to collect and visualise the positions of vehicles in real-time.

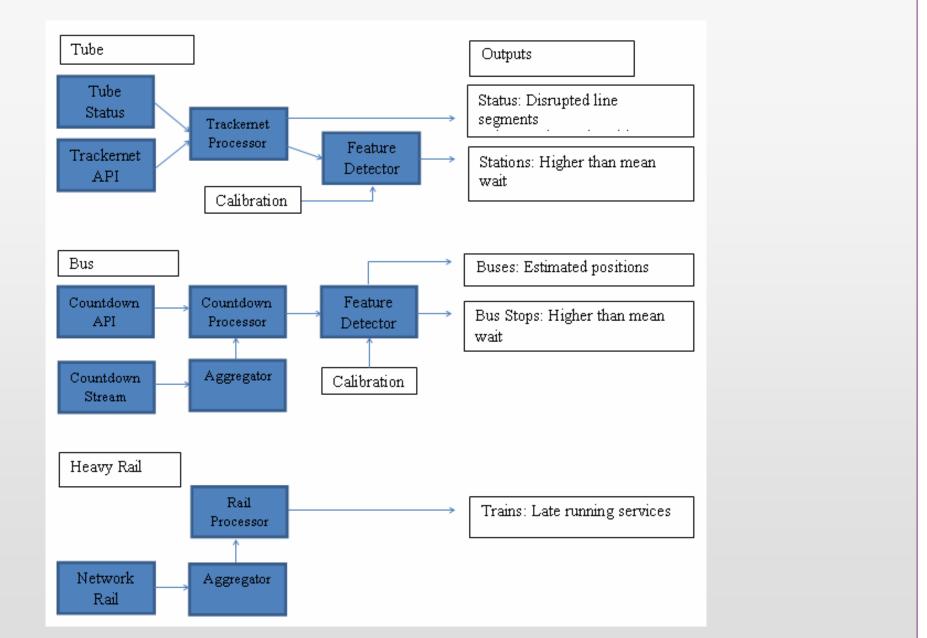
At peak periods there can be 7000 buses, 900 trains and 450 tubes running on the system

Delays for these transport systems were calculated by using an archive of historic data to find the mean wait time for every hour and every station or bus stop.

This can then be visualised in real time or after the event for further analysis. We show a mix of these visualisations in the figures that follow – as yet we have not developed an integrated analysis but all the ideas are there. We show the analysis first for the tube but here is the block diagram showing how we are assembling the data.

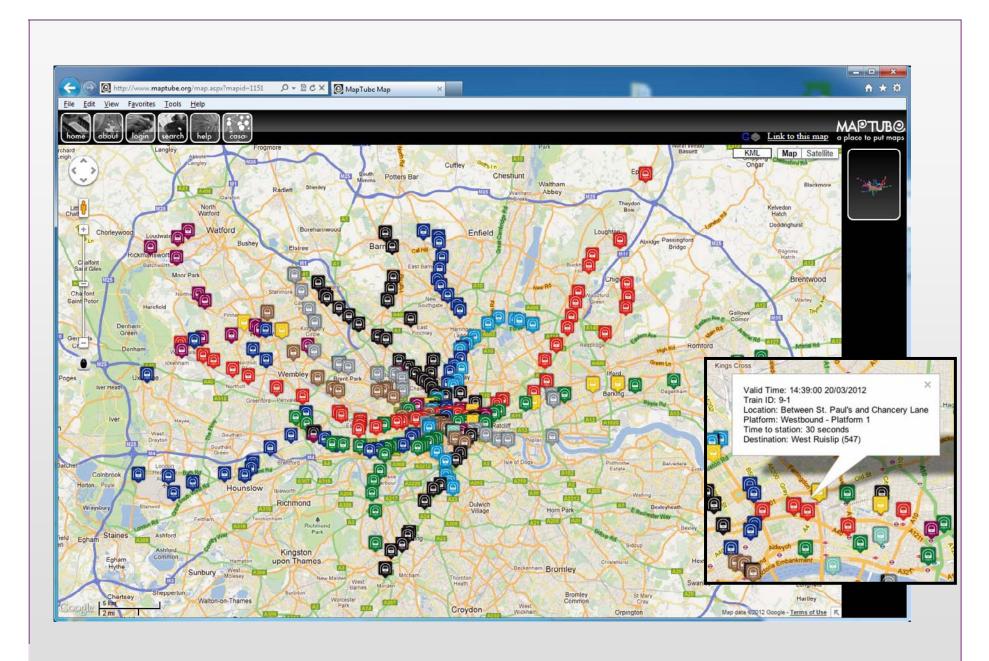






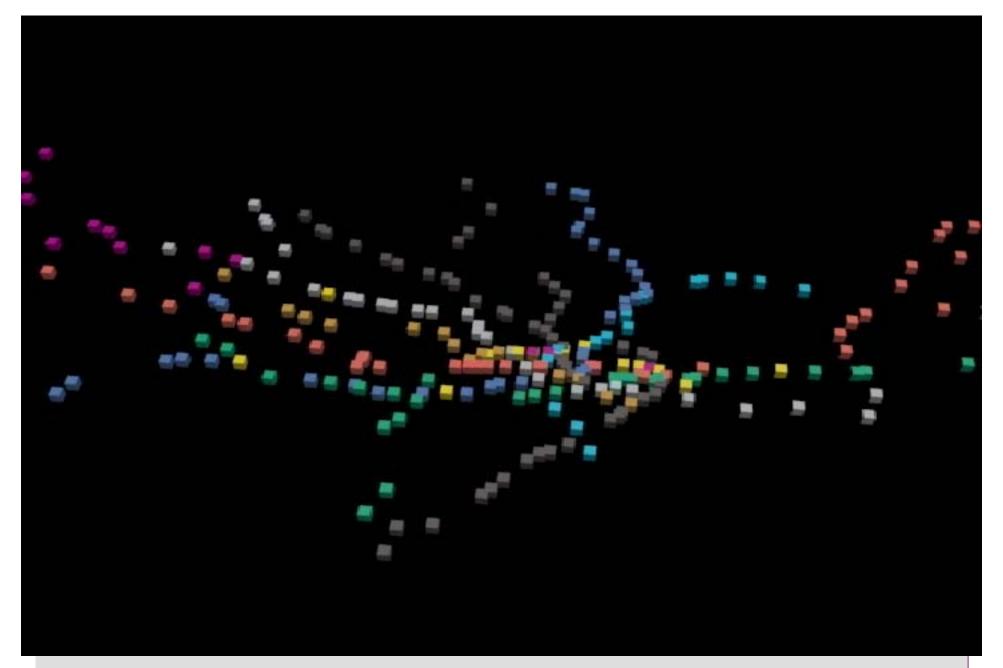








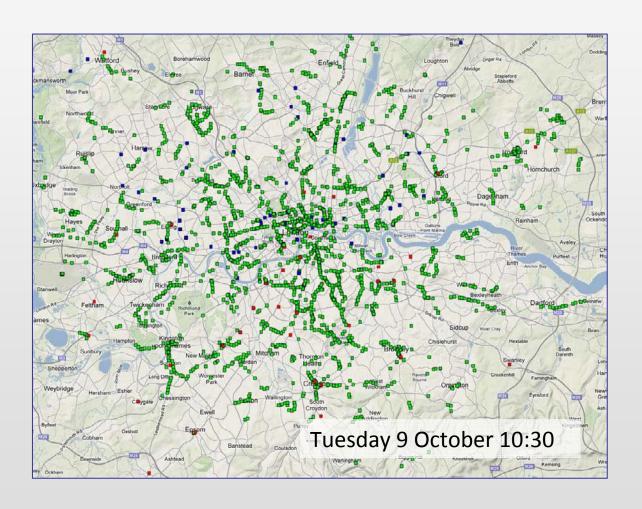




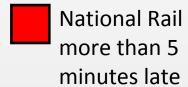


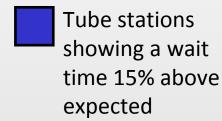


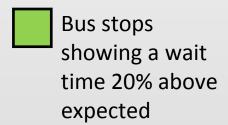
#### Delays from Tube, National Rail and Bus Fused



# Key





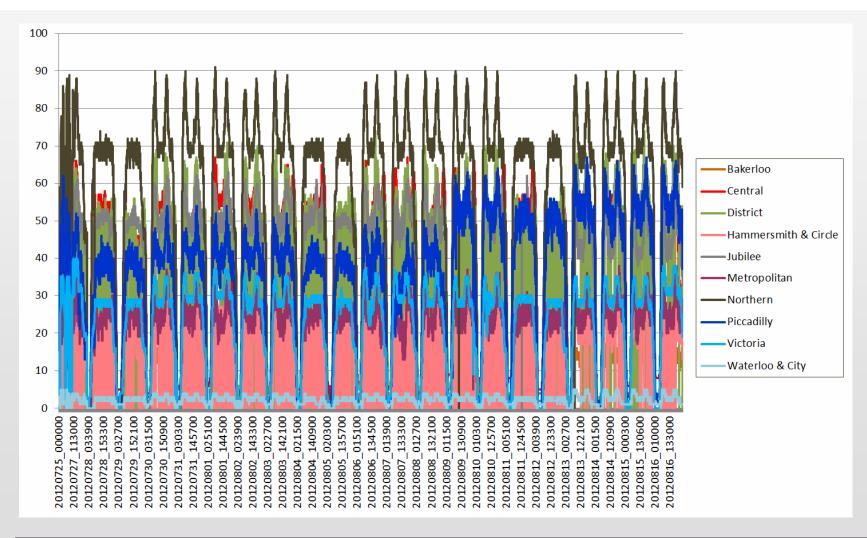


Tube delays from the TfL status feed are also plotted as lines



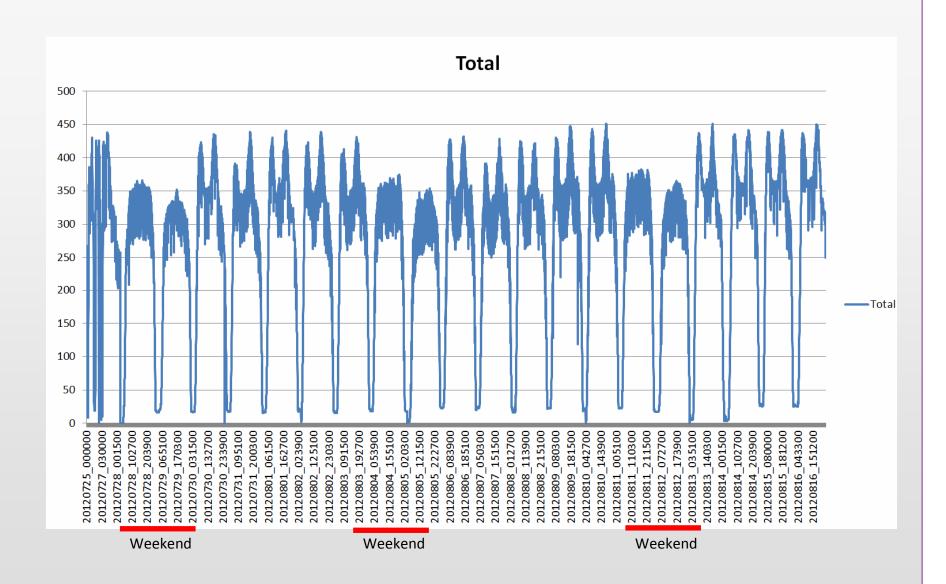


#### Flows During the Olympics – we are looking at this as a case study













#### The Effect of Bus Strike



Grand Union Canal
Slough
Astrodyn

Tuesday 22<sup>nd</sup> May 2012, 09:00

Wednesday 23<sup>rd</sup> May 2012, 09:00

The left image shows the effect of the bus strike on 22<sup>nd</sup> May 2012, while the image on the right shows a normal day.





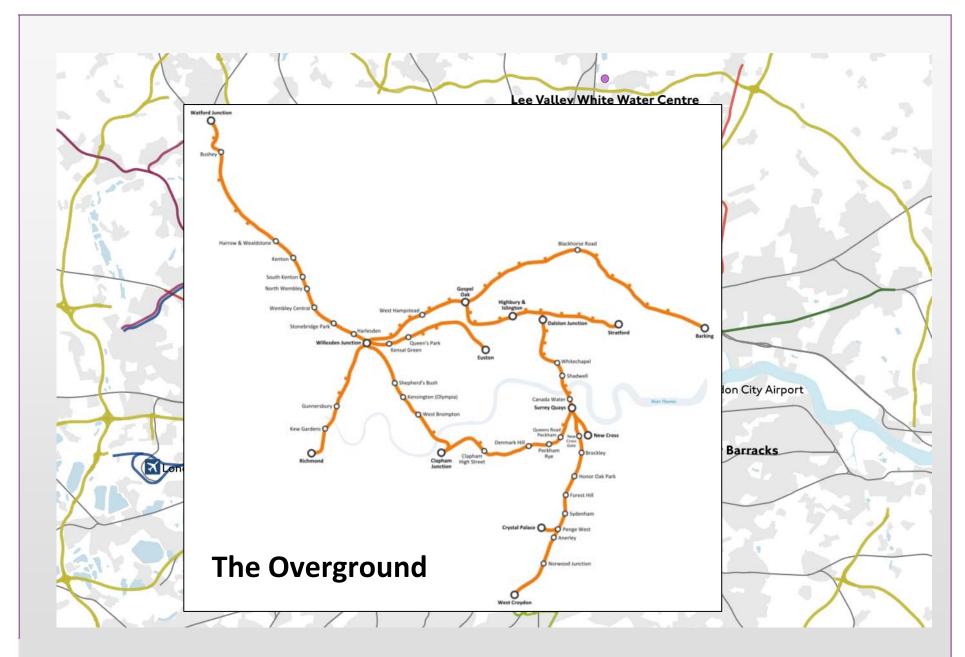
I want to introduce a little more background on the automated scheme in London that we are discussing – some history first and then a map of the network

It is quite confusing because so many networks overlap so we should be clear about the core network – the underground which has 270 stations; the Docklands Light railway has 45; the overground has 83 but this extends out side of Greater London: then national rail is more complicated because it dovetails with overground and underground stations also coincide but are not the same as national rail stations.

We are mainly going to deal here with the greater London networks but the touch card data is for a much bigger area and also for the bus system and it can also be related to national rail.

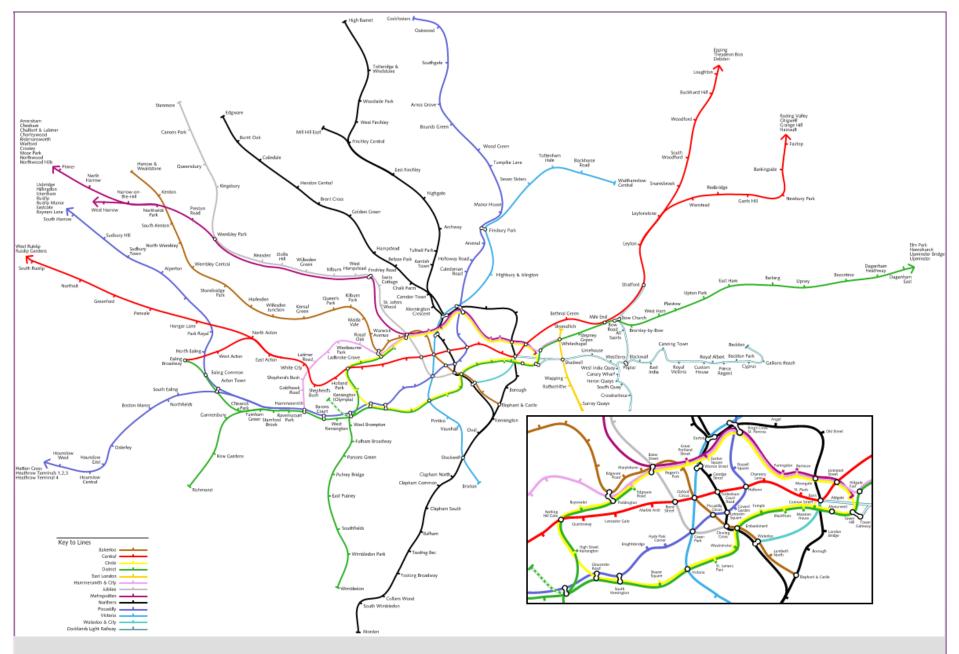






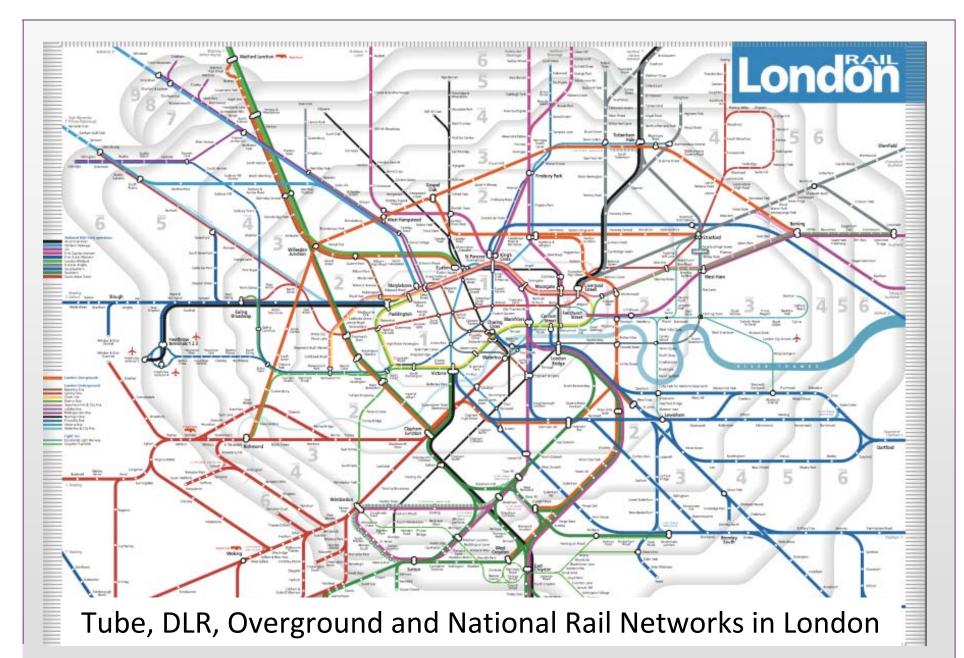








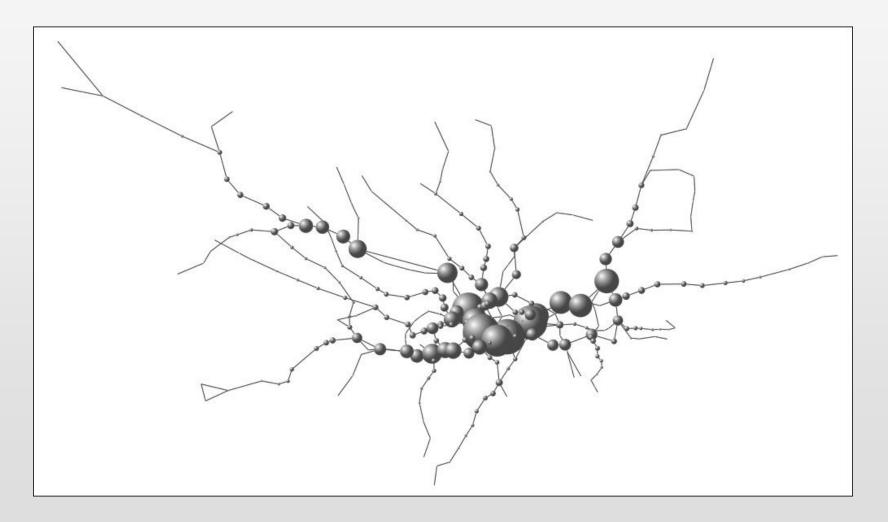








The simplest network for the first problem, based on an analysis of the network, not the flows: shown below are the degrees







# Problem 2: Representing Networks

We use standard graph algebra to represent the network where we define three indices of centrality

Betweenness Centrality 
$$C_k = \sum_i \sum_j \frac{\sigma_{ikj}}{\sigma_{ij}}$$

Closeness Centrality 
$$L_i = \mathit{KD}_i^{-1} = \mathit{K} \left( \sum_j d_{ij} \right)^{-1}$$





# Representing Flows

Trip Volume **Entries and Exits** 

$$\left\{egin{array}{l} T_i = \sum_j T_{ij} \ T_j = \sum_i T_{ij} \end{array}
ight\} \ T = \sum_i T_i = \sum_j T_j = \sum_i \sum_j T_{ij} \end{array}$$

Changes in **Trip Volumes** 

$$\left\{ egin{aligned} \Delta_i &= T_i - T_i' \ \Delta_j &= T_j - T_j' \end{aligned} 
ight. \left. \sum_i \Delta_i = \sum_i \Delta_j = 0 \end{aligned}$$

Centrality

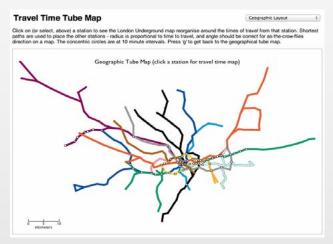
Weighted Betweenness 
$$p_{ijk} = \frac{\sigma_{ikj}}{\sigma_{ij}} = \frac{\sigma_{ikj}}{\sum_{\ell} \sigma_{i\ell j}}$$
 ,  $\sum_{k} p_{ikj} = 1$  Centrality

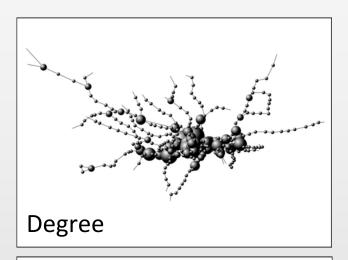
$$\widetilde{C}_{k} = \sum_{i} \sum_{j} T_{ij} p_{ikj} = \sum_{i} \sum_{j} T_{ij} \frac{\sigma_{ikj}}{\sigma_{ij}}$$

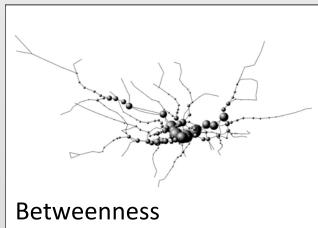


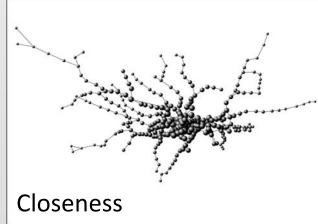


# A Preliminary Analysis (1) The Minimal Tube Network and the Three Centrality Indices













# A Preliminary Analysis (2)

- Top Stations
- By Centrality

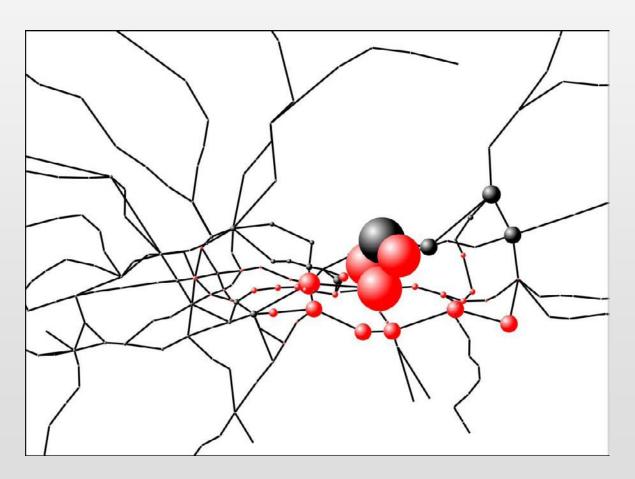
Station	$d_{i}$	Station	$\hat{\pmb{C}}_i$	Station	$\hat{L}_i$
Baker Street	7	Green Park	16399	Green Park	2.137
King's Cross	7	Waterloo	15644	Westminster	2.107
Bank	6	Bank	15008	Bond Street	2.101
Earl's Court	6	Baker Street	14441	Oxford Circus	2.089
Green Park	6	Westminster	14139	Waterloo	2.089
Oxford Circus	6	Bond Street	11429	Bank	2.074
Waterloo	6	Liverpool Street	11186	Baker Street	2.071
Canning Town	5	Stratford	10814	Victoria	2.065
Liverpool Street	5	Mile End	10302	Hyde Pk Corner	2.053
Paddington	5	Bethnal Green	10017	Embankment	2.041
Shadwell	5	Finchley Road	8905	Piccadilly Circus	2.041
Tumham Green	5	Earl's Court	8706	St. James's Park	2.035
Acton Town	4	King's Cross	8679	Regent's Park	2.032
Bond Street	4	Wembley Park	7968	King's Cross	2.029
Camden Town	4	South Ken	7182	Liverpool Street	2.026
Canada Water	4	Euston	7156	Marble Arch	2.026
Canary Wharf	4	Gloucester Rd	7042	Tottenham Ct Rd	2.026
Embankment	4	Paddington	7028	Moorgate	2.020
Euston	4	Victoria	6558	Charing Cross	2.017
Finchley Road	4	Harrow-o-t-Hill	6253	Great Portland St	2.017





# A Preliminary Analysis (3)

#### **Closing Liverpool Street**

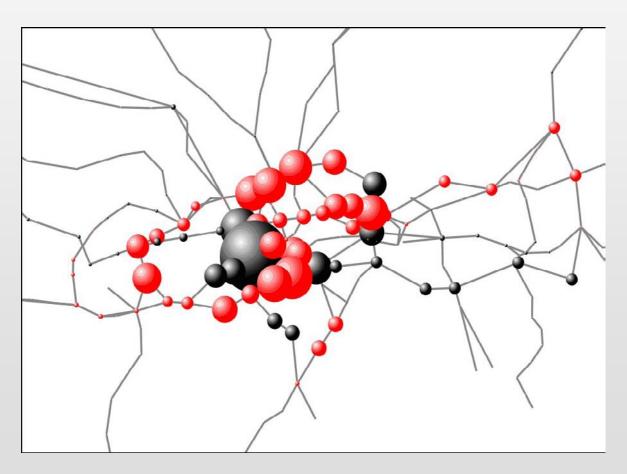






# A Preliminary Analysis (3)

## Closing Green Park

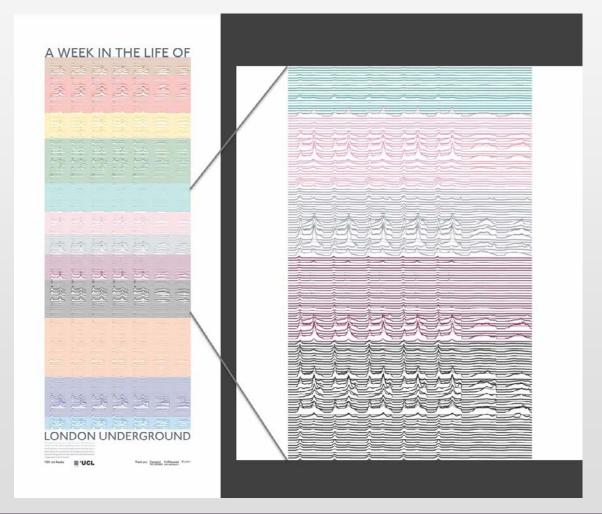






#### Problem 3: The Shortest Path

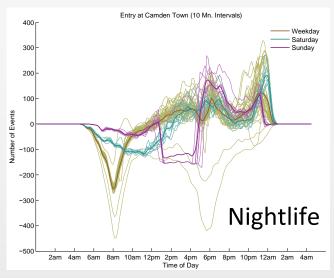
but to remind you, here we are dealing with flow data with the usual rush hour peaks and troughs

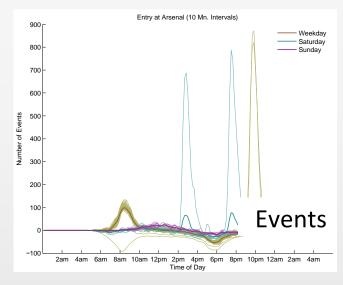


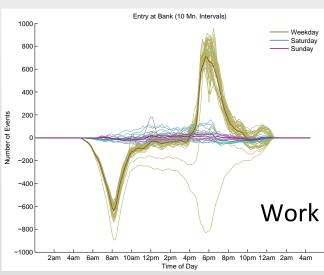


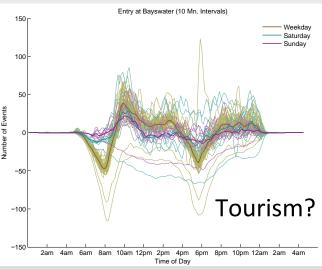


#### Station Time: Weekdays, Saturdays and Sundays





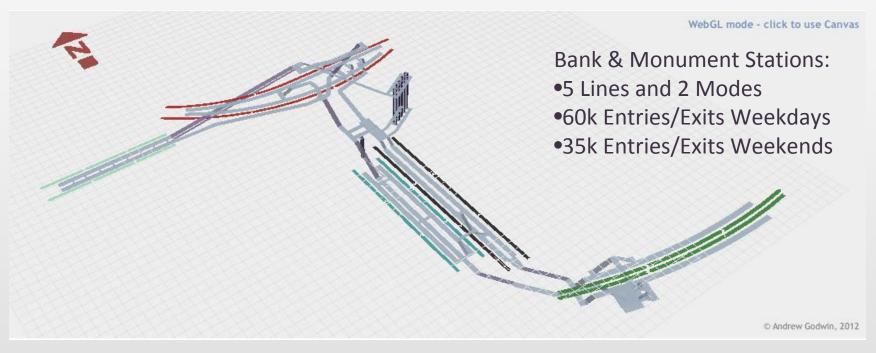








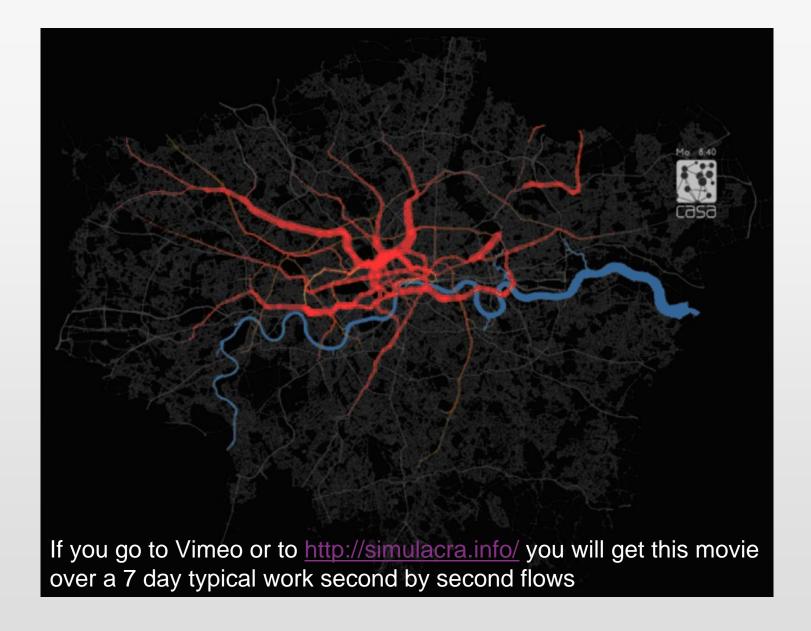
#### Problem 3: The 'Shortest'-Path Problem



Although a simple station/line network may be sufficient for small cities, for 'Mega-Cities' such as London, New York, or Tokyo a much more detailed network is needed with interchanges measured down to the platform level. The 'penalties' for changing lines (and permitted Out-of-Station Interchanges) can be severe and should be included in a schematic network representation.











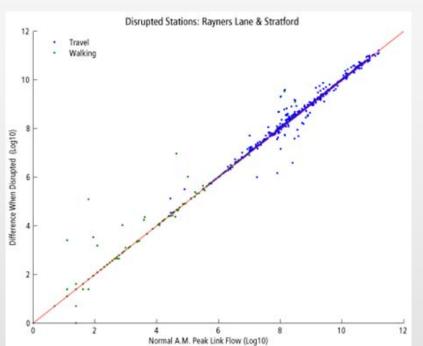
#### Methodology

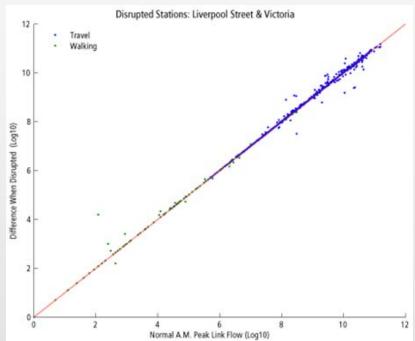
- 1. Build average O/D flows between all Under- and Over-ground stations
  - 33 days of activity with 100% coverage of pseudonymous Oyster cards
  - More than 300 million unique trip segments (of which 120 million by some form of rail)
- 2. Build walking network between all stations within 5km of each other
  - Routing on OSM network provided by routino using realistic preferences for walk speed and intensity of road usage
- 3. Build integrated travel-time network representation of both modalities
  - End-to-end travel time extracted from routing
  - Physical layout of stations inferred from real-time platform data
- 4. Simulate simultaneous disruption for 1 or 2 stations
  - Use real O/D matrix and remap disrupted trip segments
  - Realistic disruption on basis of entry/exit/interchange breakdown
  - Measure changes in volumes and 'lost' travel times across segments





#### Link-Level Disruption



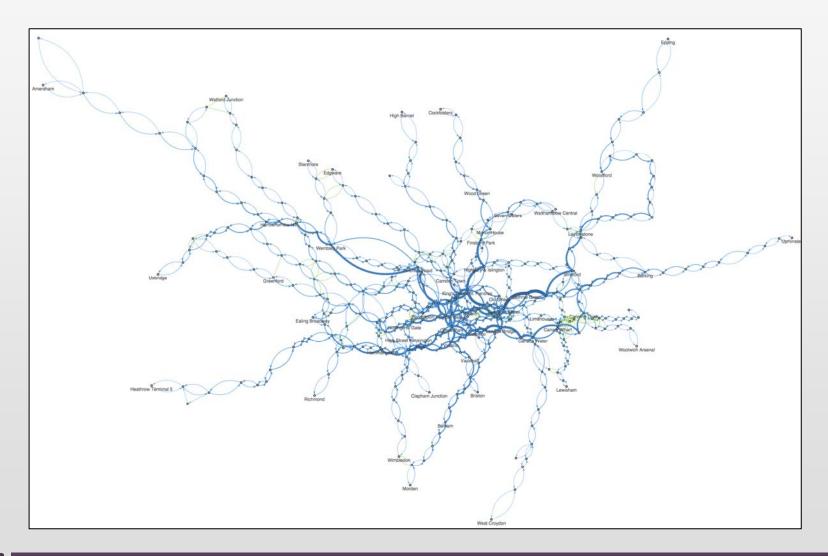


Single- and dual-station disruptions produce unexpected link-level interactions: changes in shortest-path typically cause some links to lose passengers, and gains are often less than expected. Moreover, it is not the biggest and most central stations that cause the largest shifts!





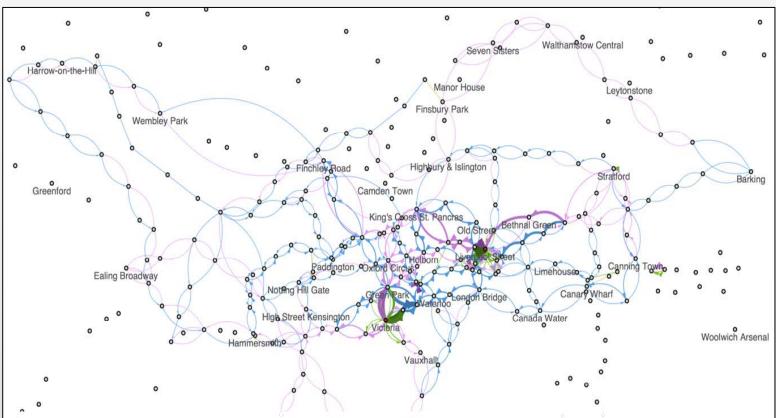
#### The Undisturbed Network







#### Liverpool Street & Victoria

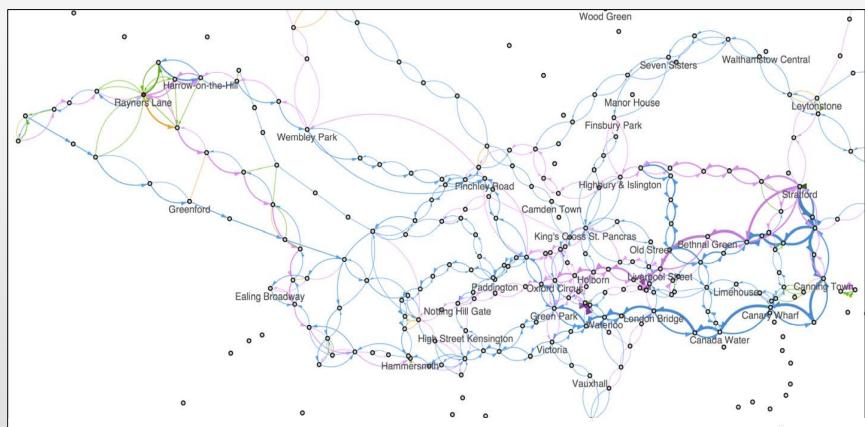


Two of London's busiest stations – because of connections to mainline rail – but if disruptions are localised to the Tube *alone* then there are many more local substitutes.





#### Rayners Lane & Stratford

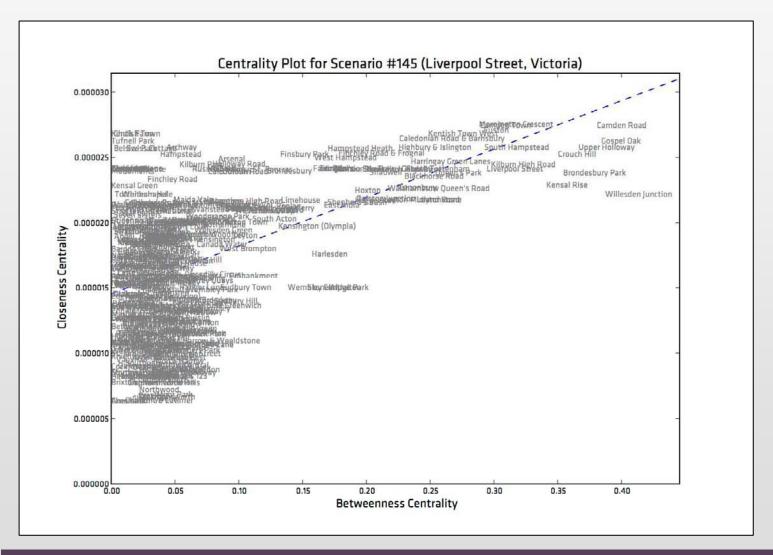


Secondary interchanges outside the core seem to cause greater disruption. Major re-routing required to complete journey, and time lost to walking long distances or travelling via more circuitous routes is much greater.





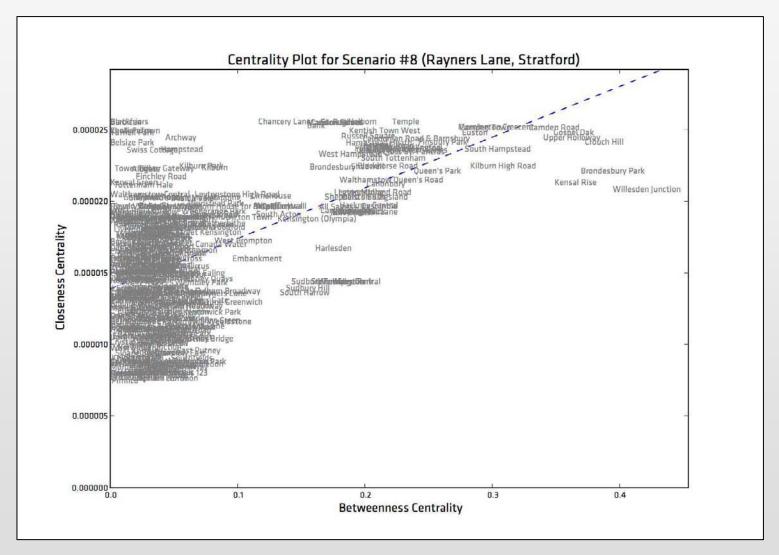
#### Betweenness & Closeness: Liverpool Street & Victoria







#### Betweenness & Closeness: Rayners Lane & Stratford







# The Impact of the London Olympics

#### Strategy #1: Demand Distribution

The distribution of Olympic venues helped to create spectator contra-flows to much of the normal rush hour traffic; however, commuters also adjusted their behaviour in small, but important ways: increasing their use of the routes least affected by Games (e.g. the Victoria line, some NR) and moderating their use of heavily trafficked routes (e.g. the Central line).

#### Strategy #1: Demand Distribution

Londoners with greater discretion over their travel clearly responded more actively to suggestions that they avoid intensive travel during the Games. However, the single largest group of users – 'Commuters' who make use of bus, rail and Tube network on a daily basis – made proportionally fewer adjustments to their average weekly activity.





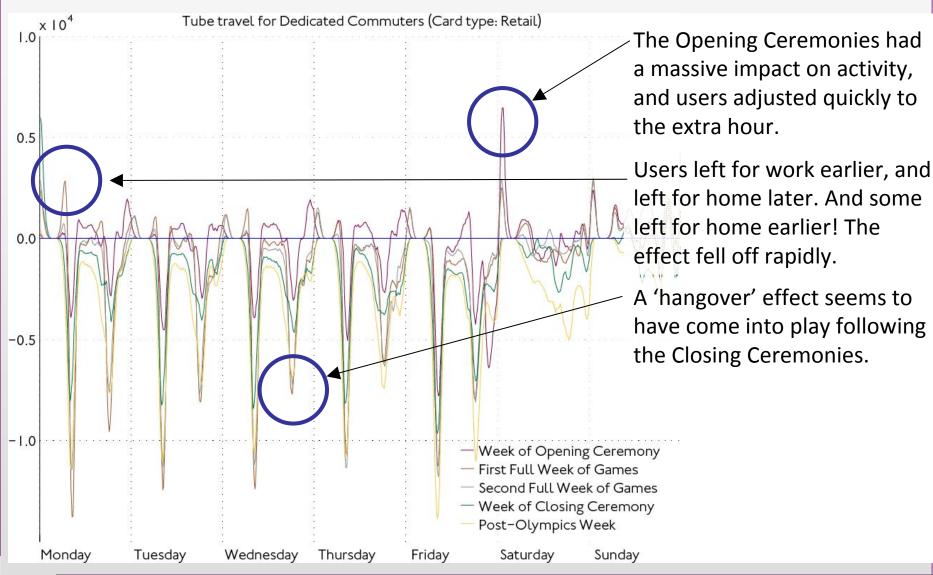
Before the Games	Games Week #1	Games Week #2	Count	Pct.
Commuter	Commuter	Commuter	2,250,000	87%
Commuter	Commuter	Infrequent	122,000	5%
Commuter	Infrequent	Commuter	146,000	5%
1-3 Times/Week	1-3 Times/Week	1-3 Times/Week	782,000	48%
1-3 Times/Week	Infrequent	Infrequent	274,000	17%
1-3 Times/Week	Commuter	Commuter	75,000	5%

For Commuters, n ≥2,580,000; For 1-3 Times/Week, n ≥1,640,000





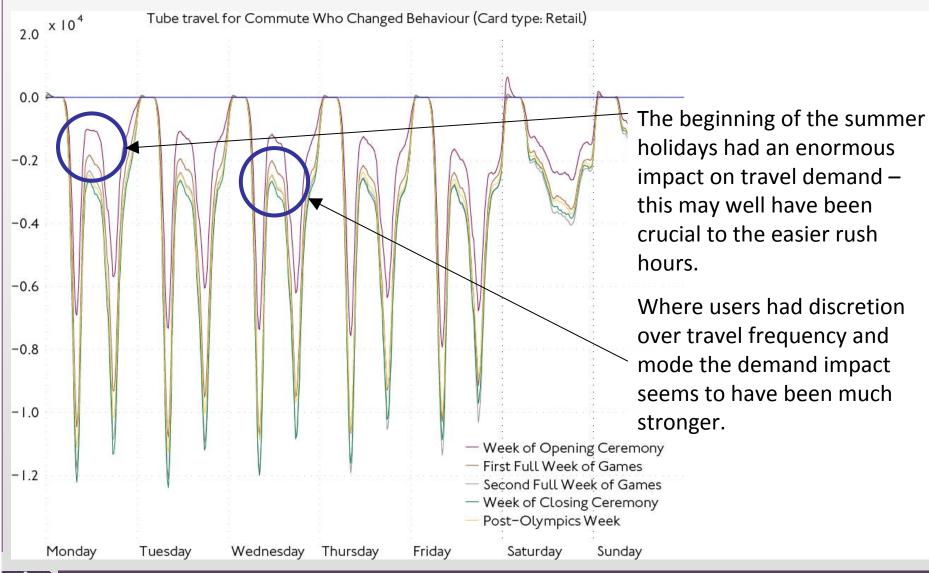
# Strategy #3: Demand Shifting







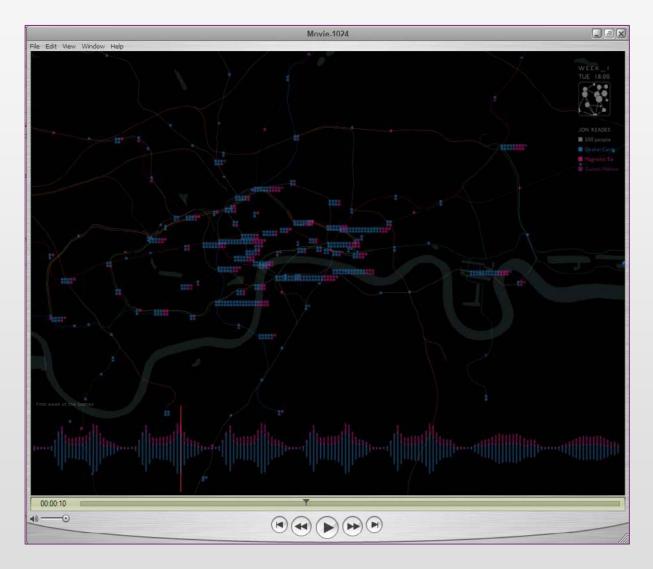
# Strategy #3: Demand Shifting







#### To end, an interesting movie of the flows during the Olympics







#### Next Steps

- 1. Identify meaningful measures enabling comparison *between* scenarios:
- 2. Need to capture both individual station and cumulative network impacts
- 3. As well, the network model could be improved in several ways:
- 4. Better-respect known route-choice preferences using RODS survey data and, potentially, Space Syntax-like 'cognitive complexity' (e.g. compare difference between route time and map complexity)
- 5. Improve modelling of interchange penalties by taking re-entry and ticket cost into account for different passenger groups
- 6. Improve modelling of interchange times at particularly large and complex stations (e.g. King's Cross St. Pancras)
- 7. Incorporate National Rail system flows since this will change many network measures substantially (although non-Oyster users would make analysis of potential hotspots much, much more complicated).



