

# Smart Cities

## SESSION III: Lecture 1: Transport and Transit: Smart Systems and Big Data

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

# This Lecture

Note that Lecture 2 in Session 2 has been incorporated into Lecture 1 in Session 3. If you follow the PDFs on the blog, then you find this material repeated for some of the slides in this lecture

## Outline of the Lecture

1. The Context Once Again: Smart Cities, Real Time Streaming, Big Data and Movement Data
2. The Example of Transit in London
3. The Supply of Vehicles: Trains and Buses
4. The Demand for Travel: Smart Cards, Oyster
5. Resilience and Disruption
  - Examining Network Disruption: Station and Line Closures*
  - Examining Traveller Disruption: Stalled Trains*
6. Other Measures: Daily Polycentricity: Information

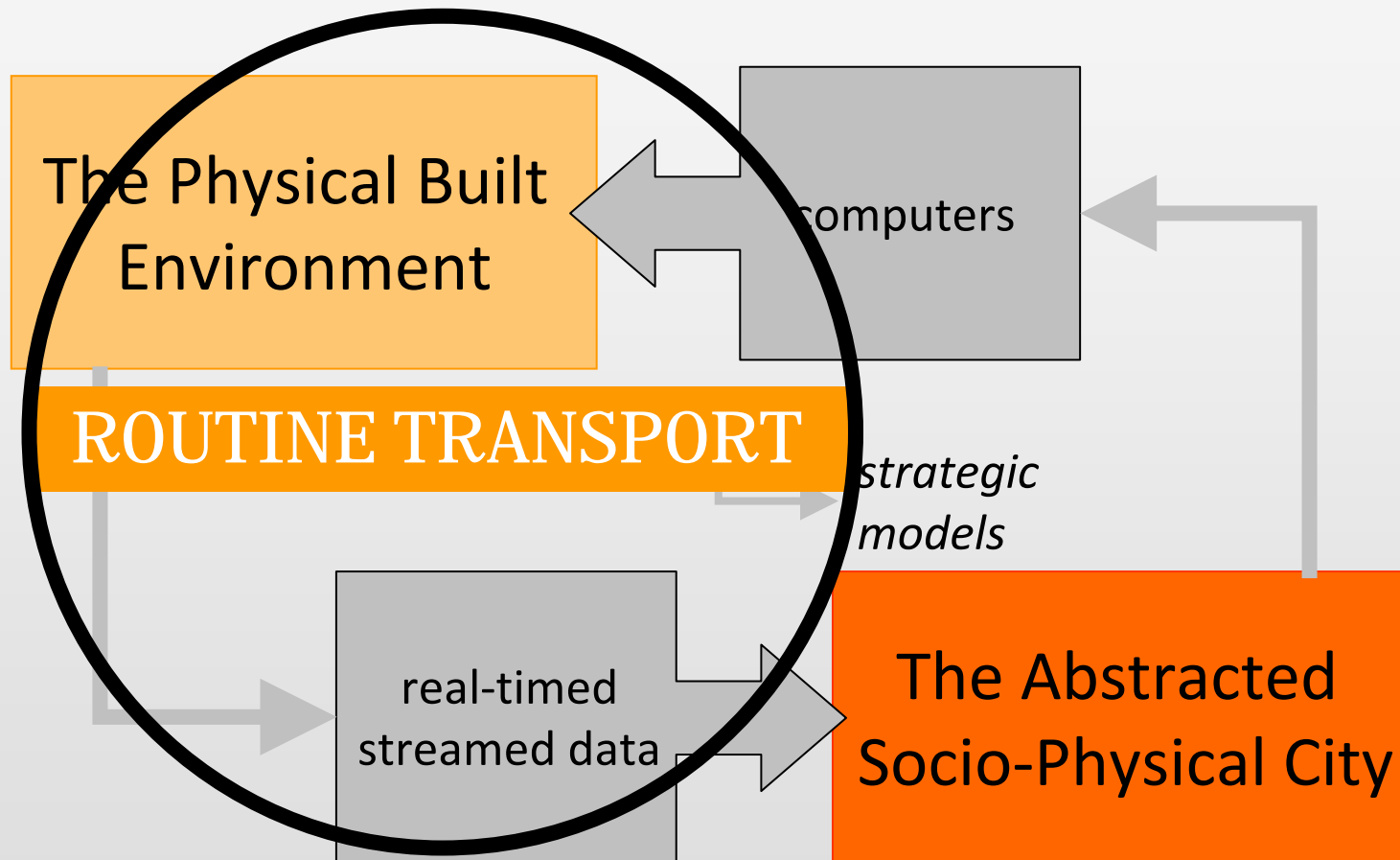
## The Context Once Again: Smart Cities, Real Time Streaming, Big Data and Movement Data

Now last time we looked at a simple model of smart cities where we said that data flow and the analysis and prediction of routine functions where computers of various sorts were embedded into the built environment were the main subject of our smart cities analytics.

The very best example of this is transit. For in enabling users and vehicles to be computerised, we can collect large volumes of data automatically – and in real time as users and vehicles operate the system – and begin to examine problems of management and control as well as prediction.

Urban analytics in this sense is about analysing the data, visualising it, and building models for short term prediction

Our focus on transport and communications mainly affect the routine aspects of smart cities over short time periods



# The Example of Transit in London

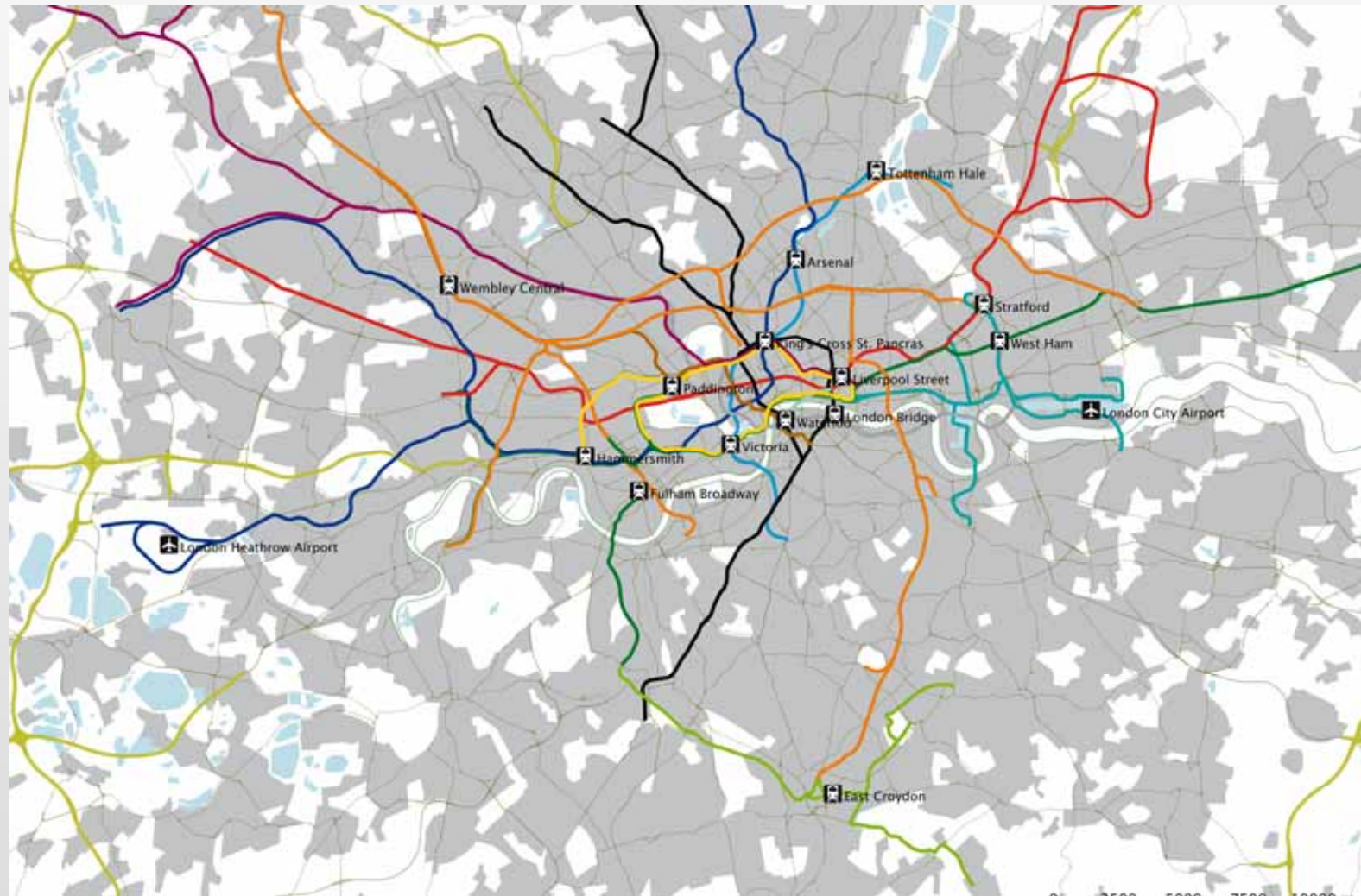
London is a good place because out of 8 million people in the inner metro area, some 50% or more are riding public transport – bus, tube (subway and light rail), heavy rail

Here are some parameters of the system

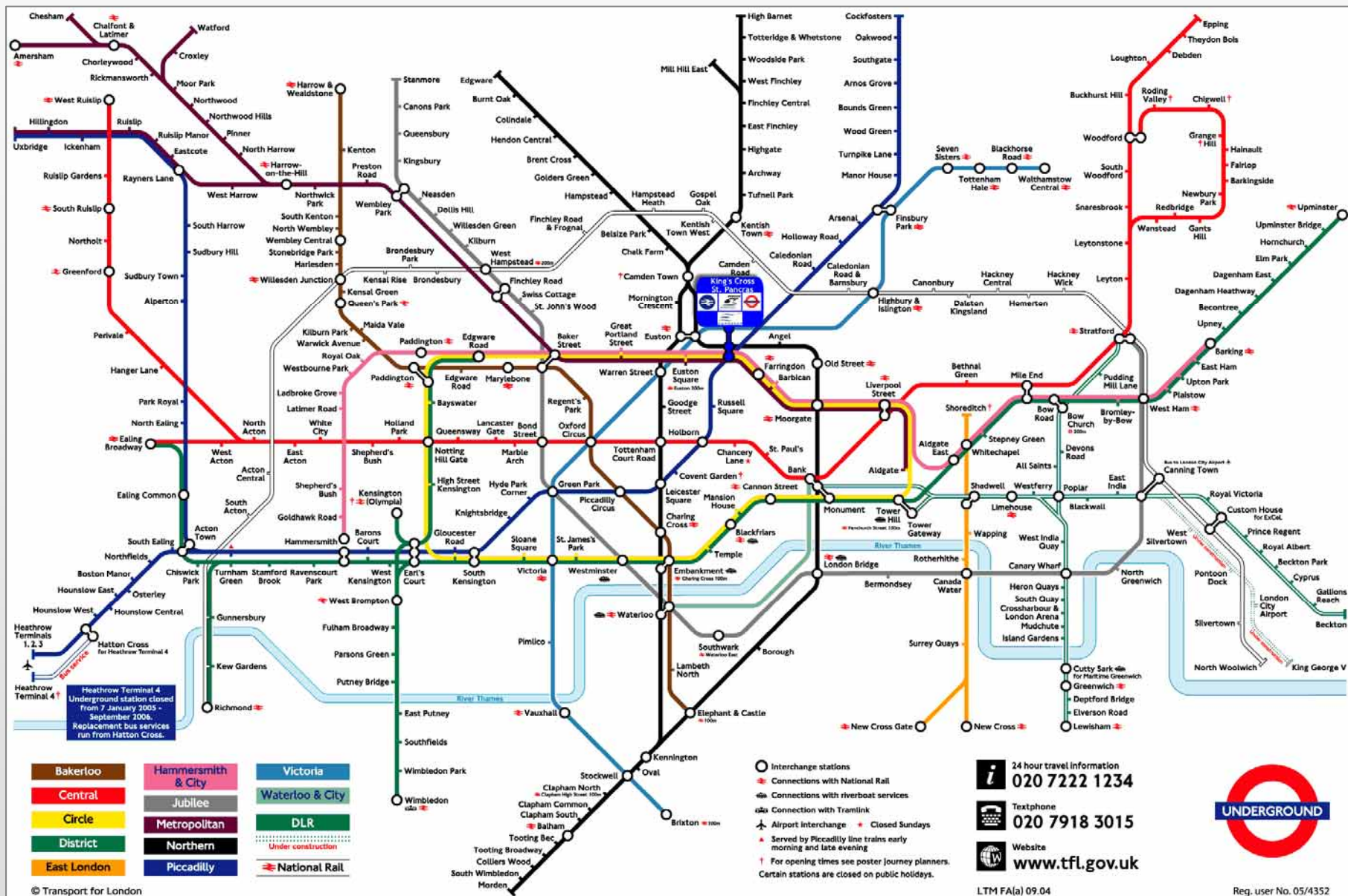
Transport for London's (TfL) RFID-based 'Oyster Card' is particularly attractive because users typically need to use their card at both ends of a trip, providing us with detailed origin and destination data for more than 3 million daily users.

- The system is particularly large and complex – our focus on tube, DLR and overground has about 400 stations, tube about 270
- 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)

## Partial Views of London's Network

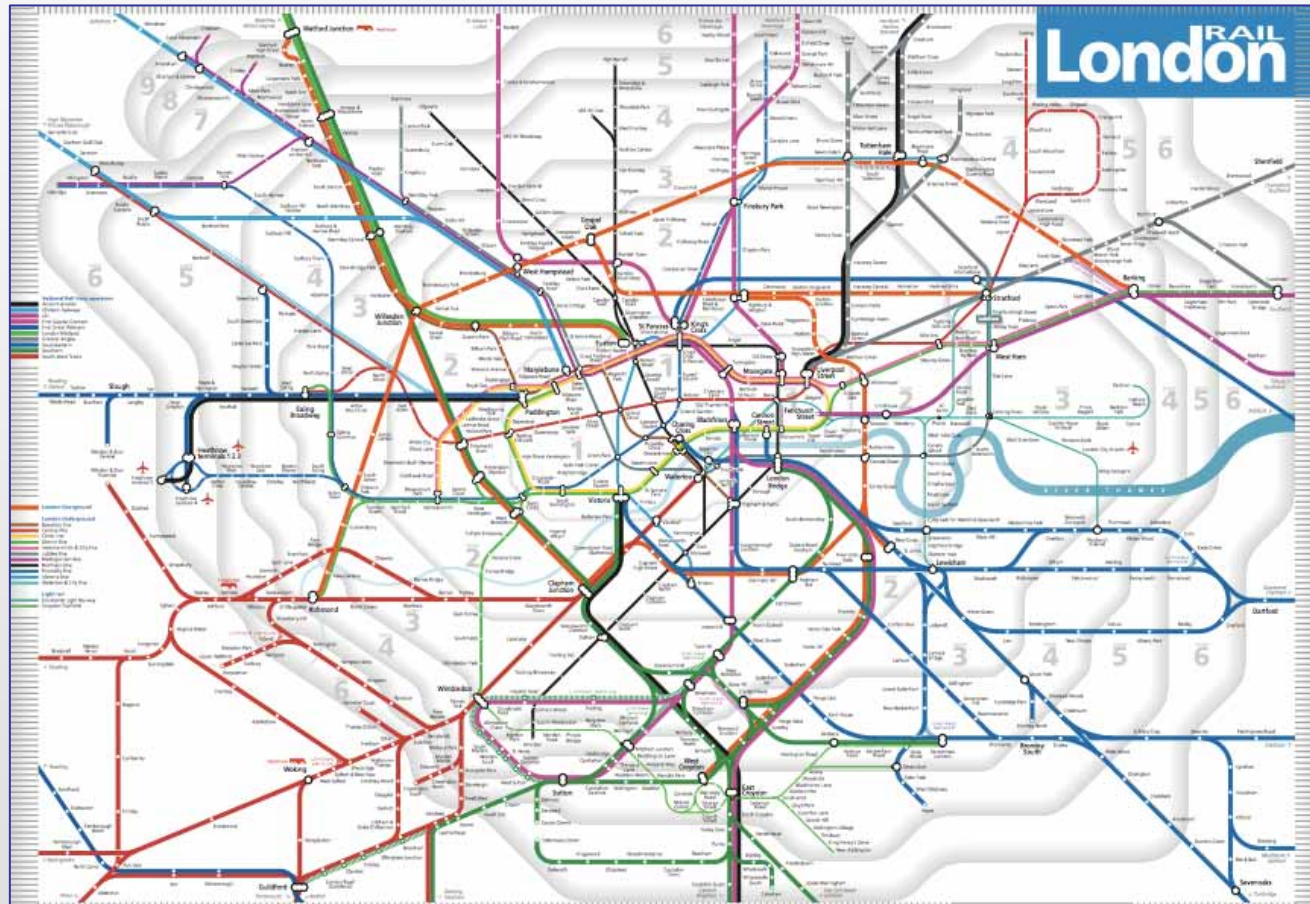








## *The Total Rail Transport Environments in London*



Tube, Overground and National Rail Networks in London  
where Oyster cards can be used

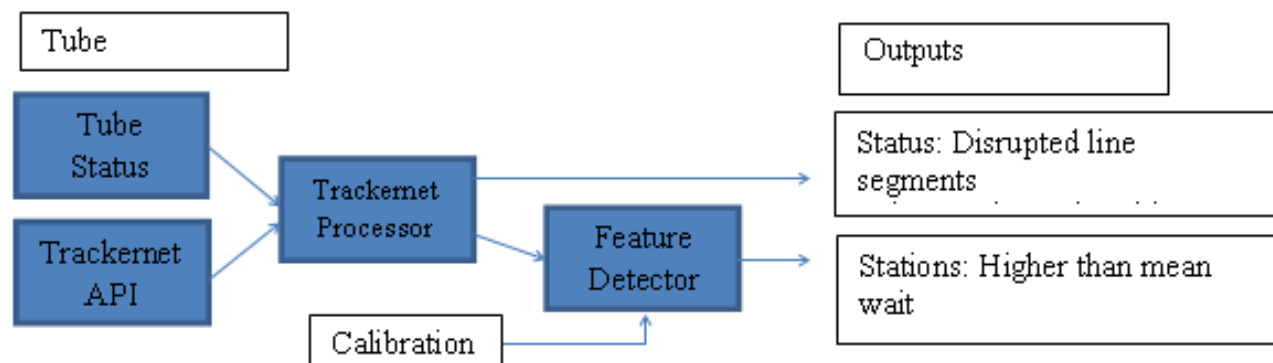
## The Supply of Vehicles: Trains and Buses

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.

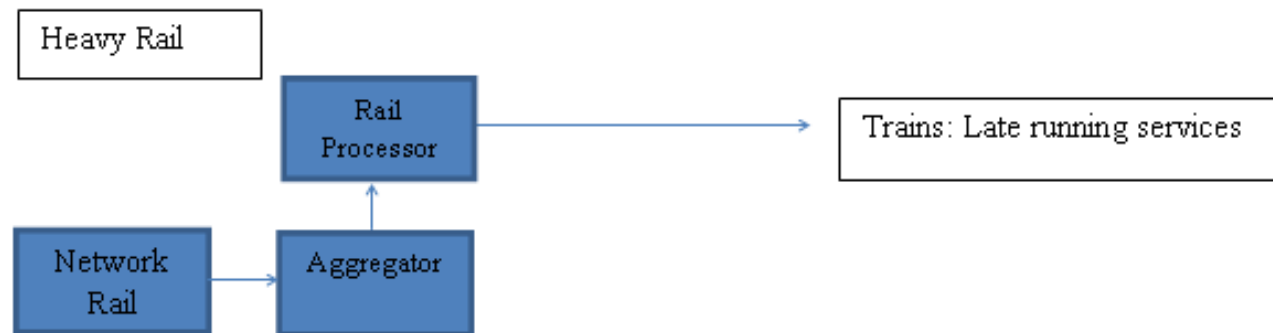
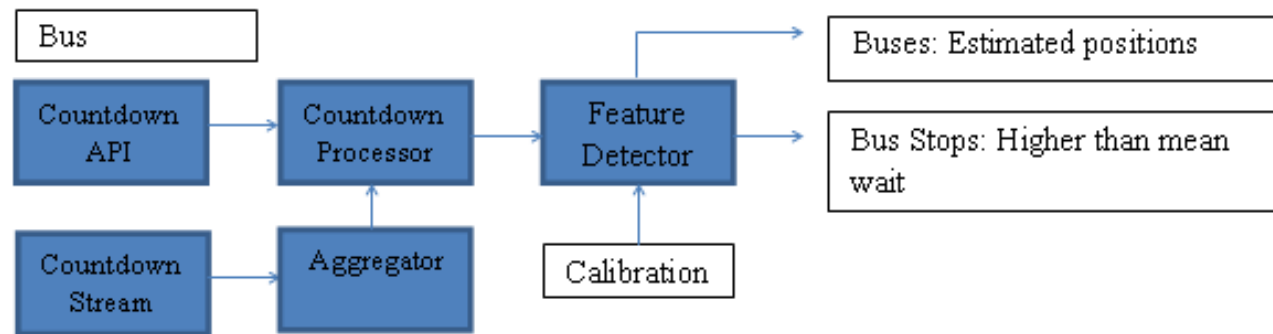
At peak periods there can be 7000 buses, 900 trains and 450 tubes running on the system. 270 underground stations.

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 – here is first is the block diagram showing how we are assembling the data.

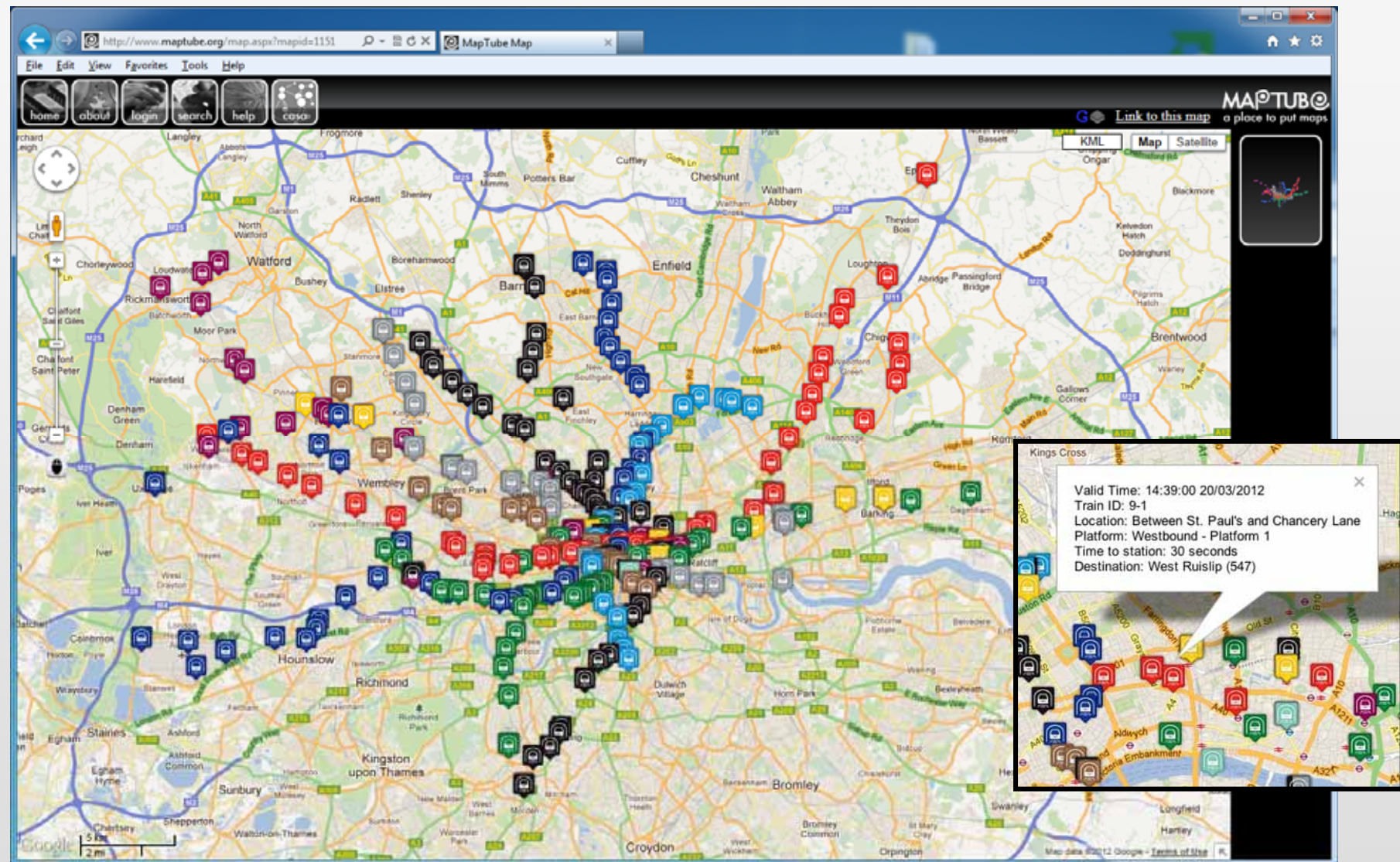


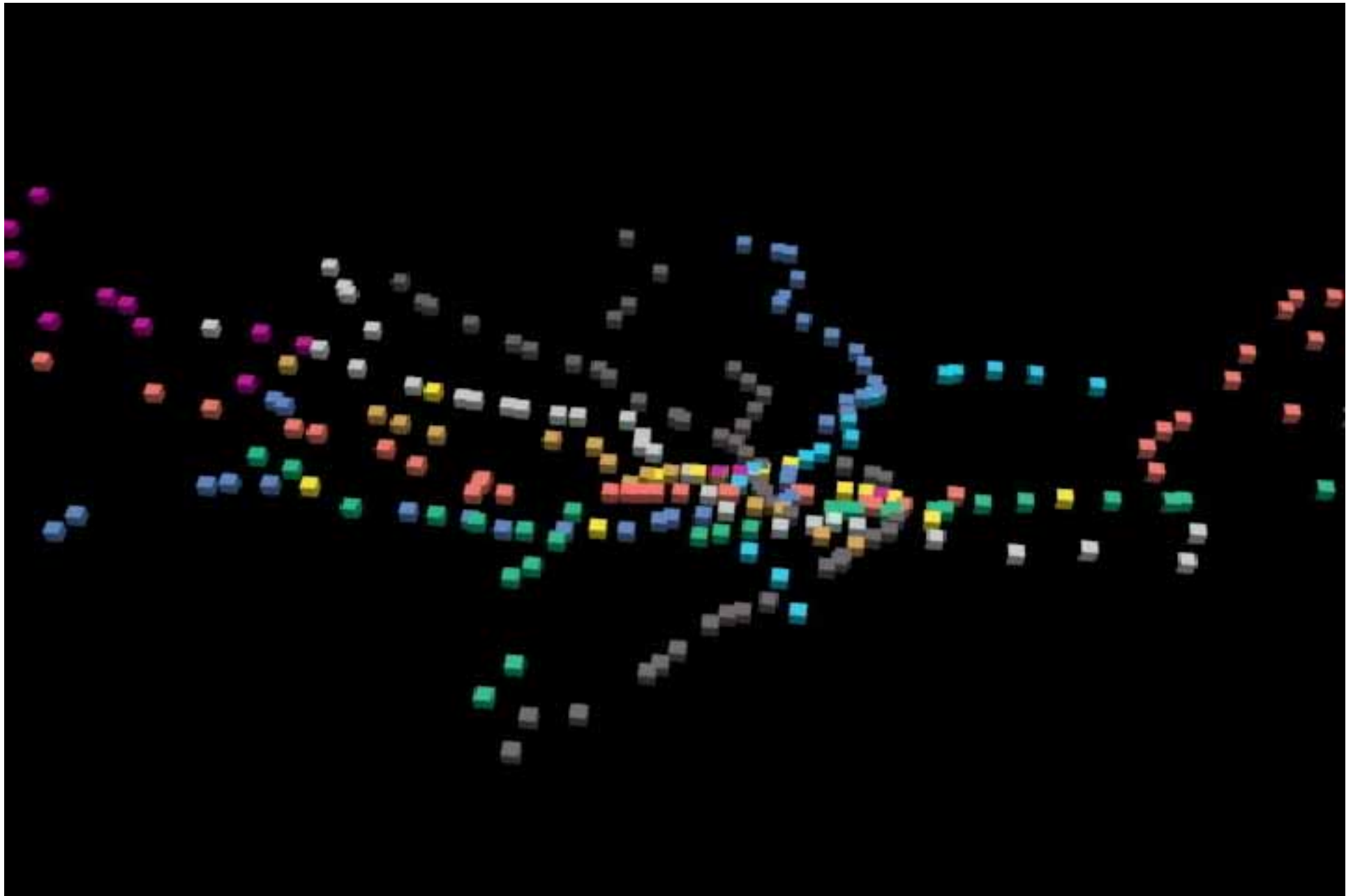
A live map: <http://traintimes.org.uk/map/tube/>

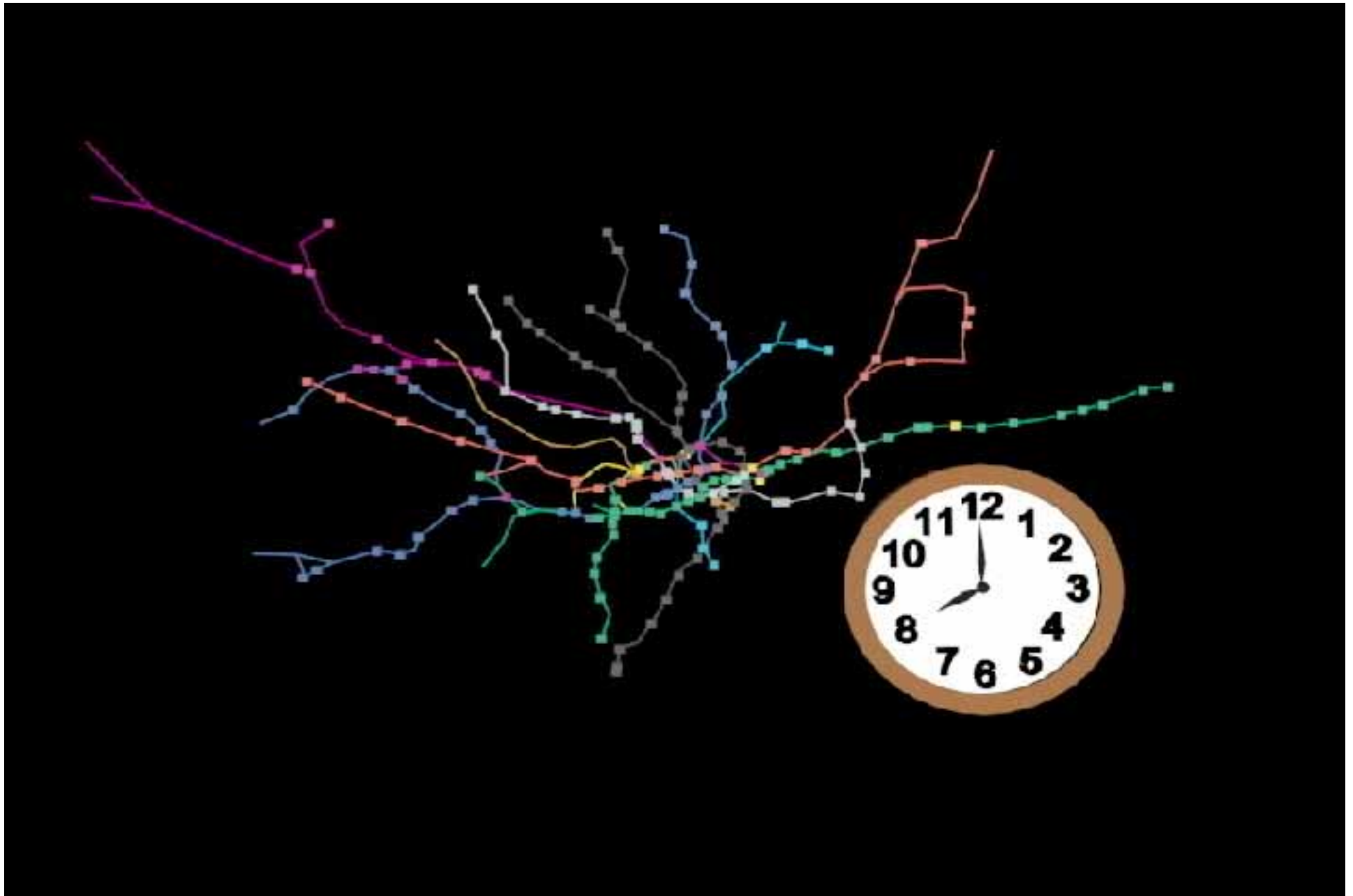




[https://www.youtube.com/watch?v=3CnQ\\_ilx9RU&feature=youtu.be](https://www.youtube.com/watch?v=3CnQ_ilx9RU&feature=youtu.be)









Transport For London Unified API

BIKEPOINT

JOURNEY

LINE

MODE

NOTIFICATION

PLACE

ROAD

SEARCH

STOPPOINT

Line

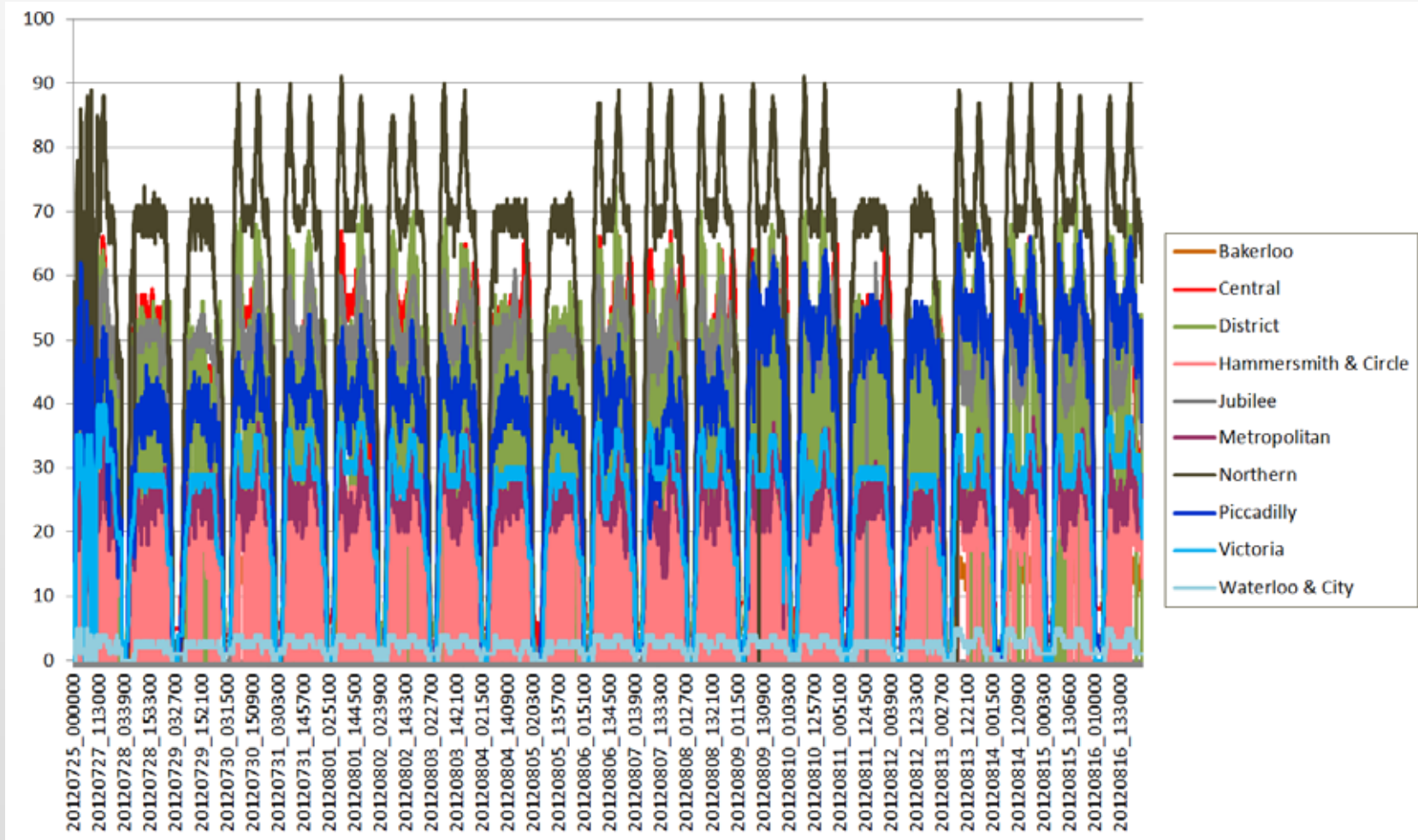
- GET** 1 minute `/Line/{ids}`  
Gets line specified by the line id.
- GET** 50 seconds `/Line/{ids}/Arrivals`  
Get the list of arrival predictions for given line ids
- GET** 50 seconds `/Line/{ids}/Arrivals?stopPointId={stopPointId}`  
Get the list of arrival predictions for given line ids based at the given stop
- GET** 50 seconds `/Line/{ids}/Arrivals/{stopPointId}?direction={direction}`  
Get the list of arrival predictions for given line ids based at the given stop going in the provided direction
- GET** 50 seconds `/Line/{ids}/Arrivals/{stopPointId}?direction={direction}&destinationStationId={destinationStationId}`  
Get the list of arrival predictions for given line ids based at the given stop going in the provided direction and terminating at given destination
- GET** 1 minute `/Line/{ids}/Disruption`  
Get disruptions for the given line ids
- GET** 1 minute `/Line/{ids}/Disruption?startDate={startDate}&endDate={endDate}`  
Get disruptions for the given line ids during provided date range

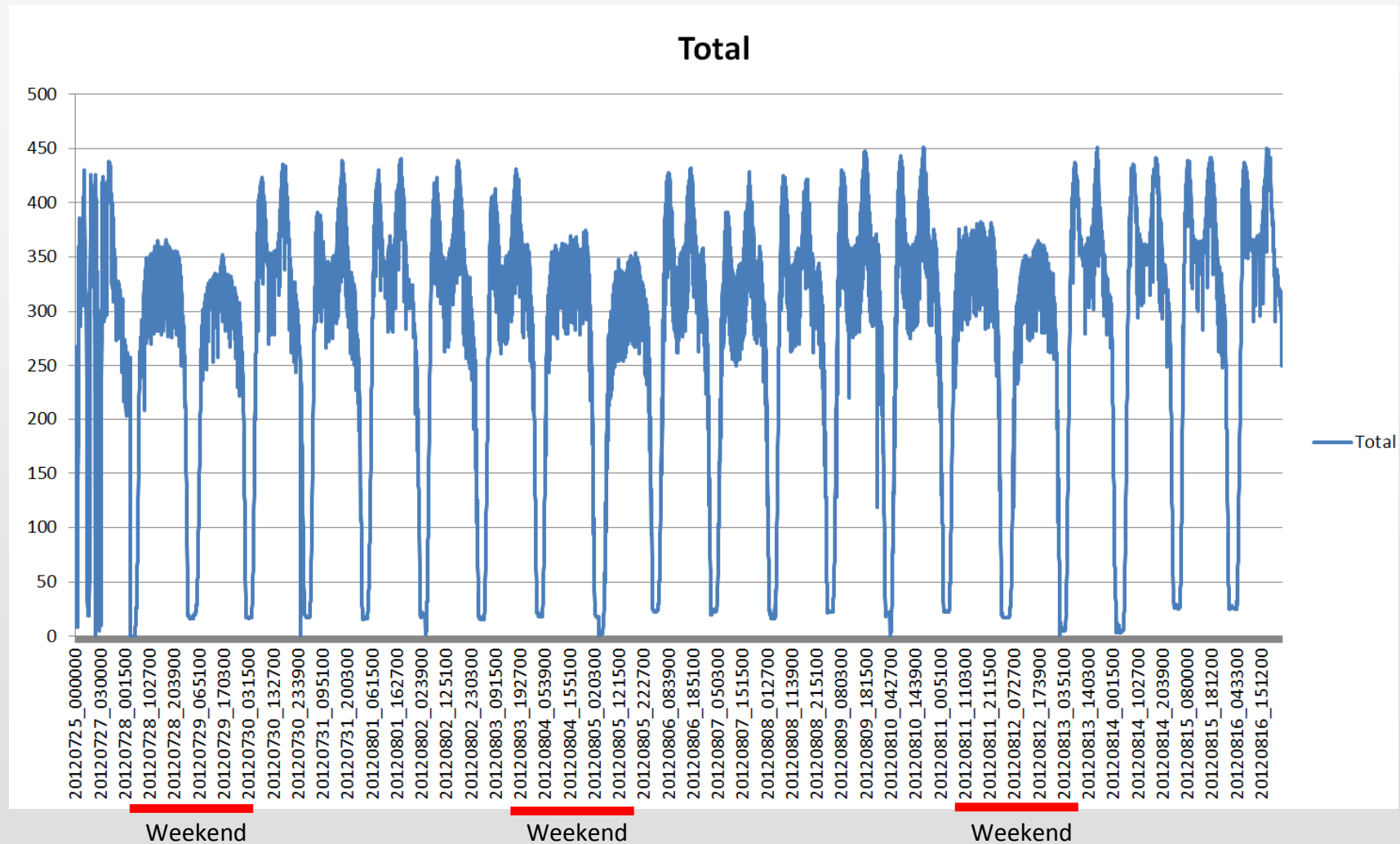
Our unified API brings together data across all modes of transport into a single RESTful API. This API provides access to the most highly requested realtime and status information across all the modes of transport, in a single and consistent way. Access to the developer documentation is available at <https://api.tfl.gov.uk>

API area	Example calls	Description
Journey planning - Westminster to Bank (Disambiguation)	<a href="https://api.tfl.gov.uk/journey/journeyresults/westminster/to/bank">https://api.tfl.gov.uk/journey/journeyresults/westminster/to/bank</a>	Journey planner supports many parameters to help find the right journey

[illegible]

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







## The Effect of Bus Strike



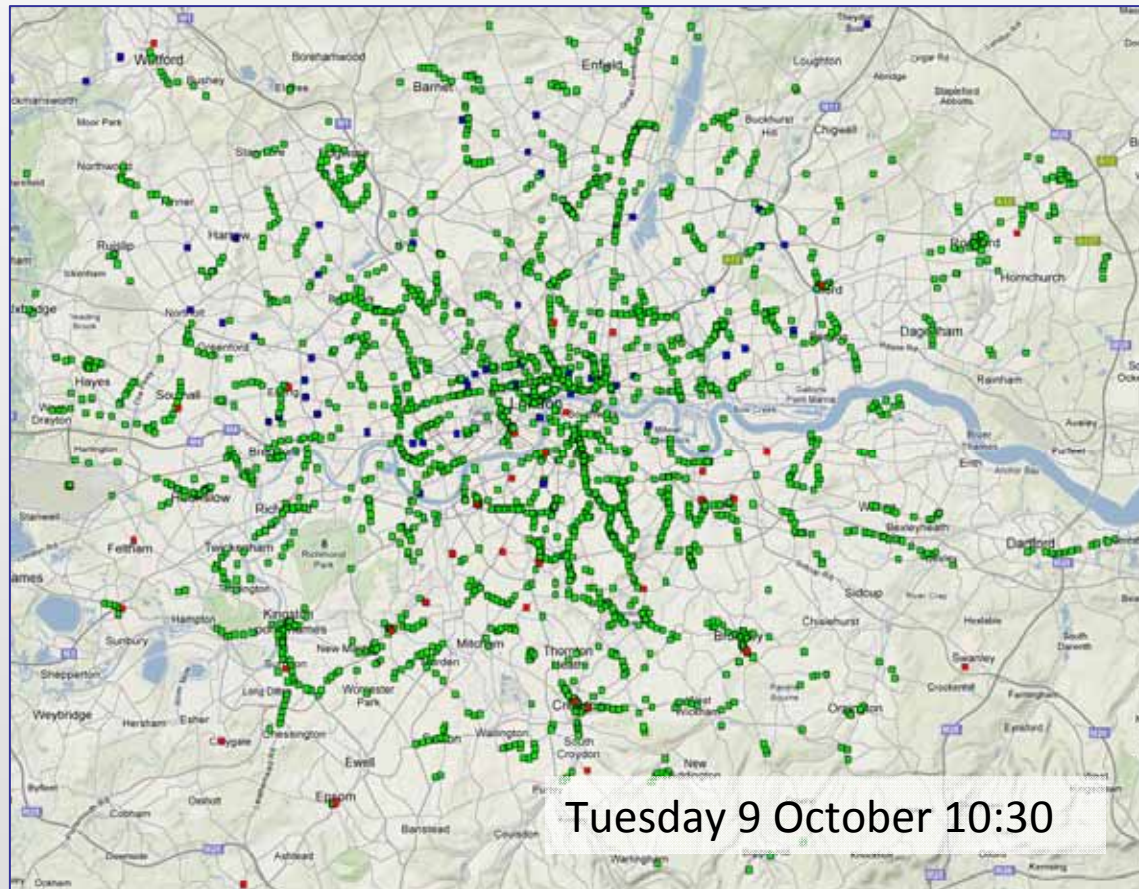
Tuesday 22<sup>nd</sup> May 2012, 09:00



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.

# Delays from Tube, National Rail and Bus Fused



## Key

- National Rail more than 5 minutes late
- Tube stations showing a wait time 15% above expected
- Bus stops showing a wait time 20% above expected

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



# The Demand for Travel: Smart Cards, Oyster

**UCL ENGINEERING**  
Change the world



Menu +

Home - Projects - Oyster gives up pearls

## OYSTER GIVES UP PEARLS

How studying millions of Oyster Card journeys reveals London's 'polycentres'



Researchers from UCL have analysed millions of Oyster Card journeys in a bid to understand how, why and where we travel in London.

Professor Michael Batty (UCL Centre for Advanced Spatial Analysis) and Dr Soong Kang (UCL Management Science and Innovation) applied the techniques of statistical physics to their mountain of raw data.

The pair joined forces with a computational social scientist and a physicist, both based in Paris, to explore patterns of commuting by tube into central London.

UCL Engineering - Oyster Gives up Pearls



0:00 / 3:32

They used Transport for London's database of 11 million records taken over one week from the Oyster Card electronic ticketing system.

**Latest news from UCL Engineering**

 New web privacy system could revolutionise the safety of surfing

UCL host Google Girls Coding Programme with Generating Genius and University of West Indies

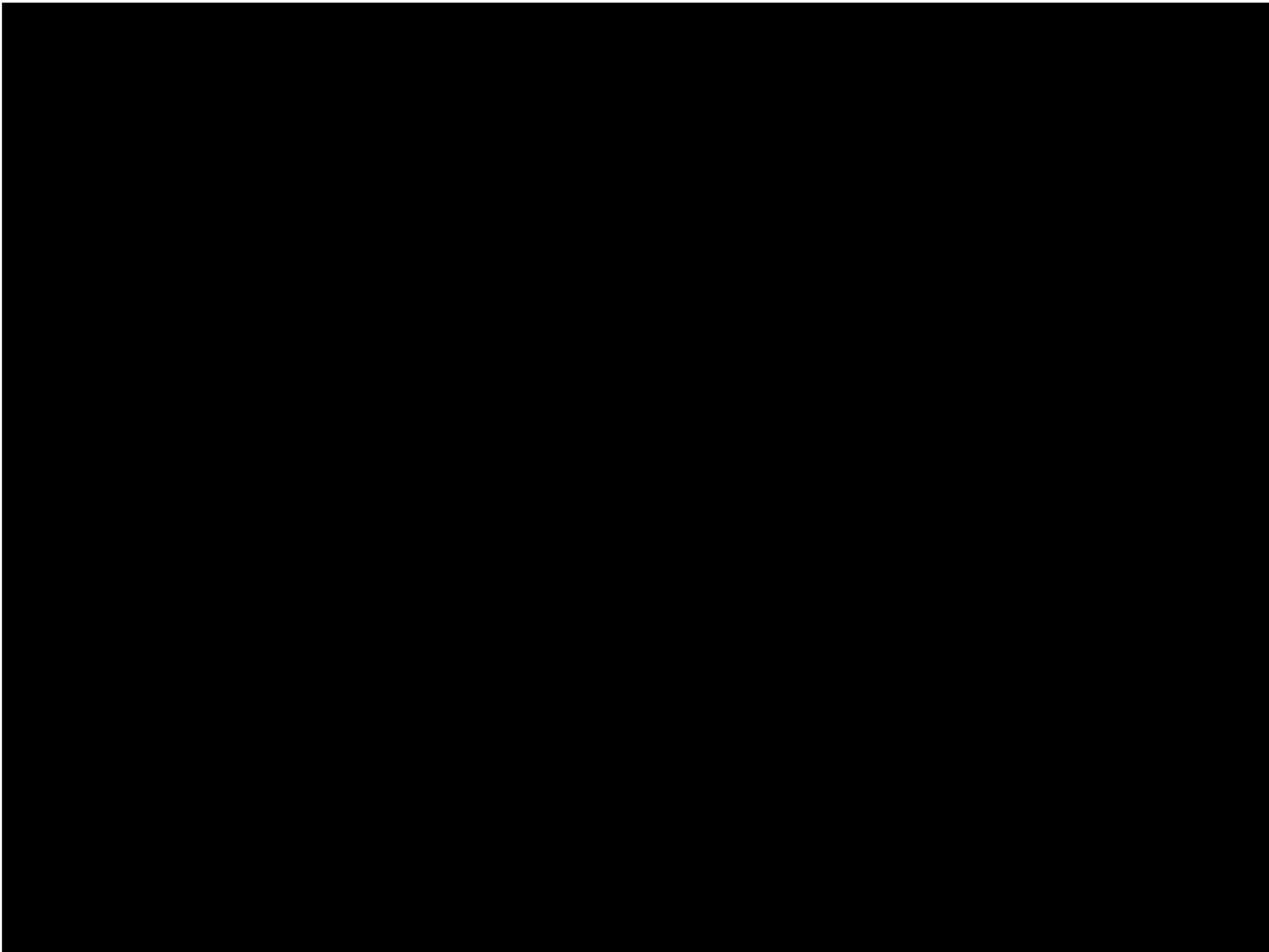
Professor Polina Bayvel to Give Royal Society Lecture

**Twitter feed**

 RT @markmiodownik: Am giving a ENGins seminar today for @UCLEngineering @UCLENGins all UCL engineers welcome - Roberts G06, 6:30pm. <http://...>  
8:58am Thu 9th October 2014

RT @Centre4EngEdu: We're hiring! Multi-talented Centre Administrator required to help us launch and expand! [bit.ly/z6ERSM](http://bit.ly/z6ERSM)  
10:54am Wed 8th October 2014

**Join our mailing list**

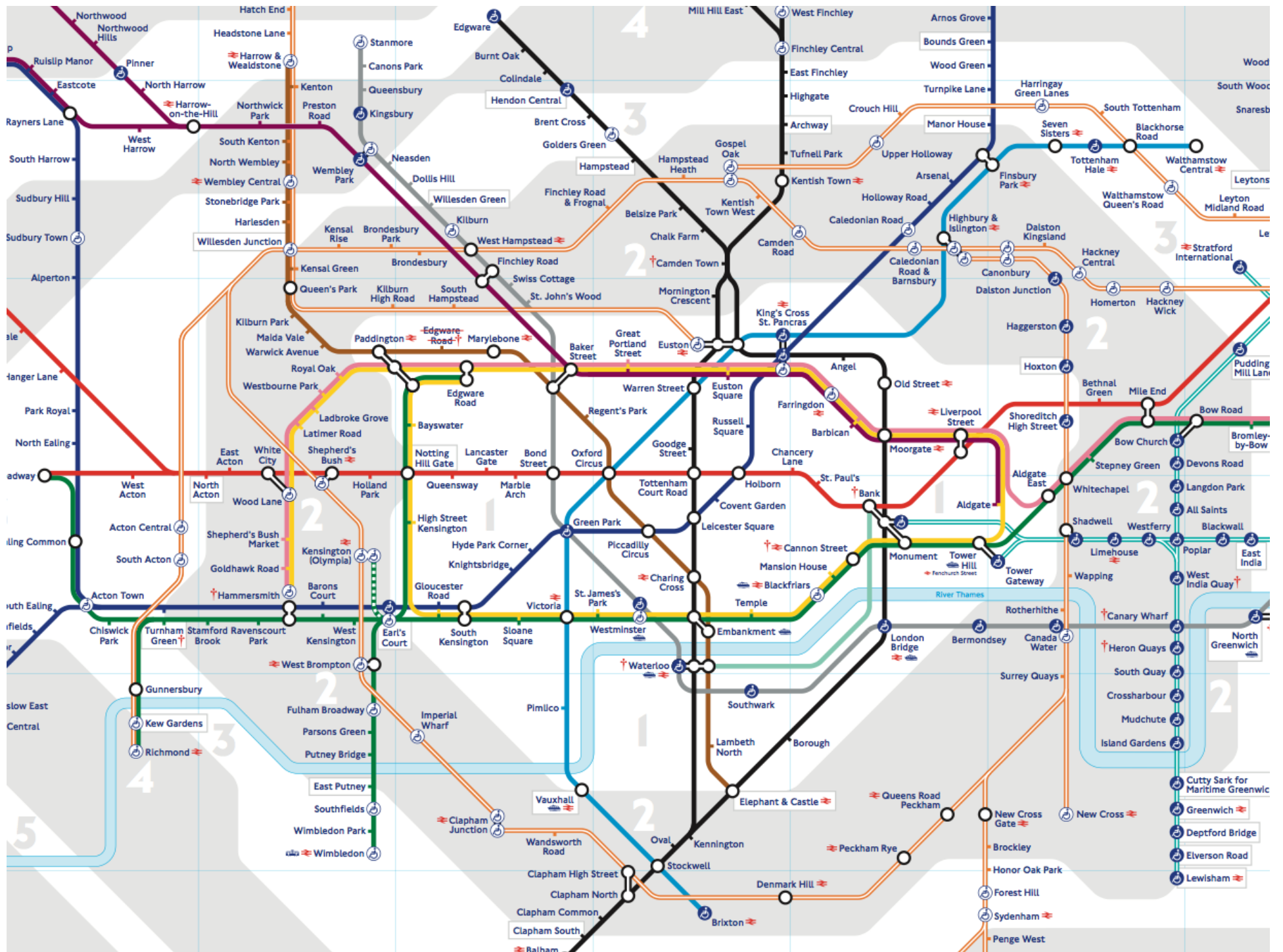


# Smart Card Data

## *Oyster Card Taps*

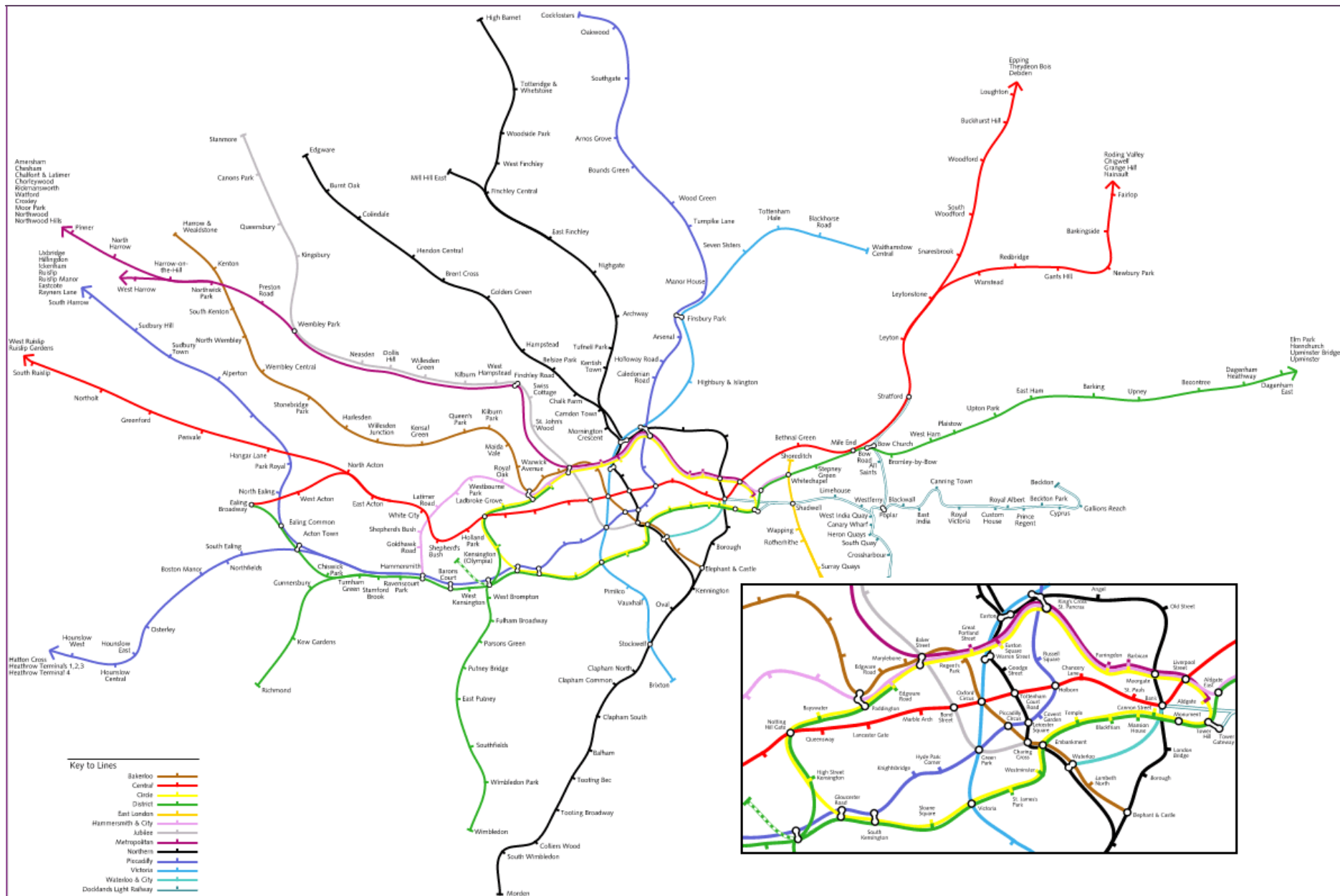
- Tap at **start** and **end** of train journeys
- Tap at **start only** on buses
- Accepted at 695 Underground and rail stations, and on thousands of buses. **1.053 billion** Oyster Card taps over July to September 201
- 762 million OD trips, 291 million UG and rail trips, 11.5m Oyster Cards. No routing data









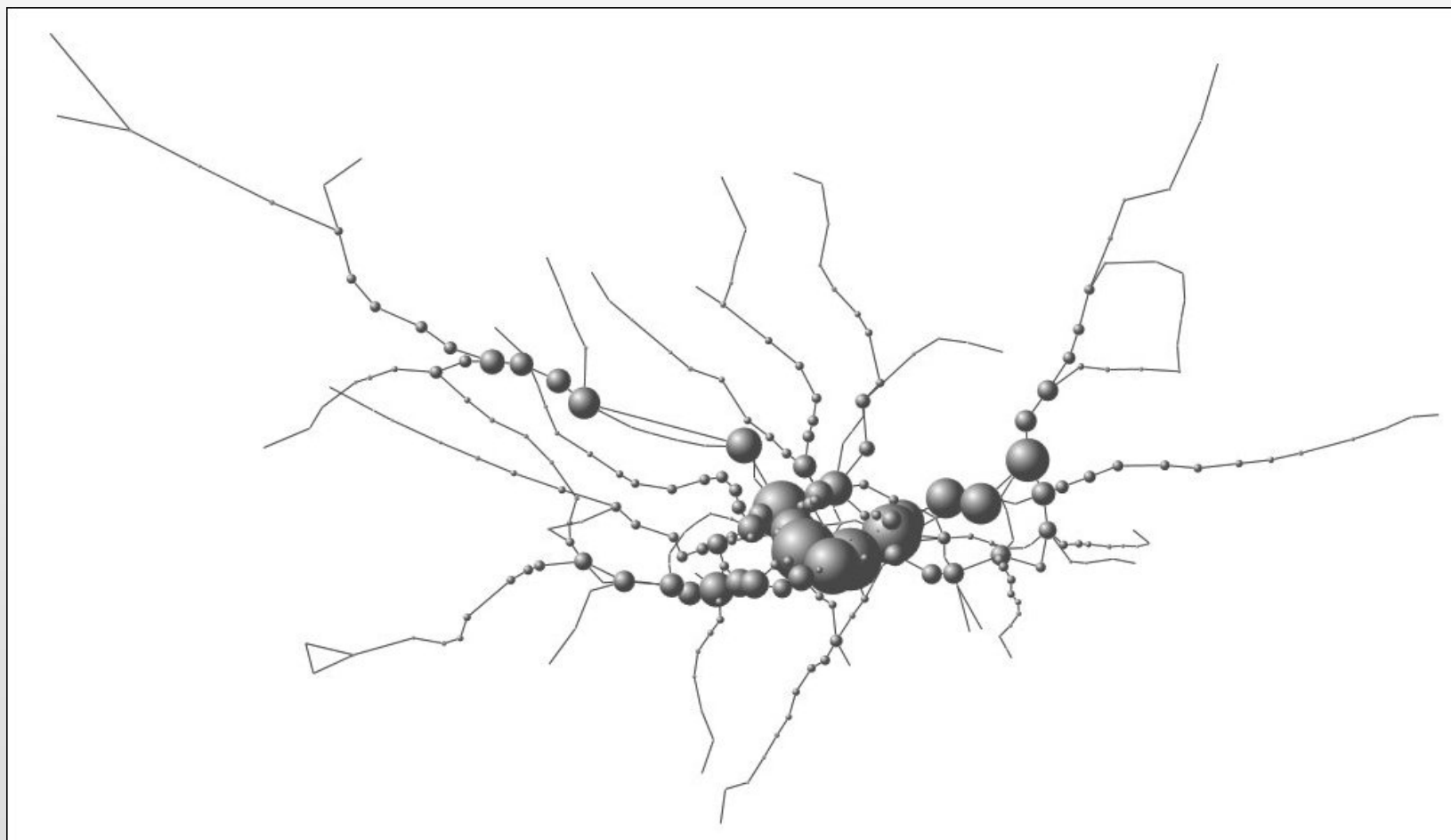






# Resilience and Disruption

## *Examining Network Disruption: Station and Line Closures*



# Representing the Tube Network

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

Degrees of the graph  $\left. \begin{array}{l} \sigma_i = \sum_j a_{ij} \\ \sigma_j = \sum_i a_{ij} \end{array} \right\} \quad \sigma = \sum_i \sigma_i = \sum_j \sigma_j = \sum_i \sum_j a_{ij}$

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

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

# Representing Flows

Trip Volume  
Entries and Exits

$$\left. \begin{aligned} T_i &= \sum_j T_{ij} \\ T_j &= \sum_i T_{ij} \end{aligned} \right\} T = \sum_i T_i = \sum_j T_j = \sum_i \sum_j T_{ij}$$

Changes in  
Trip Volumes

$$\left. \begin{aligned} \Delta_i &= T_i - T'_i \\ \Delta_j &= T_j - T'_j \end{aligned} \right\} \sum_i \Delta_i = \sum_j \Delta_j = 0$$

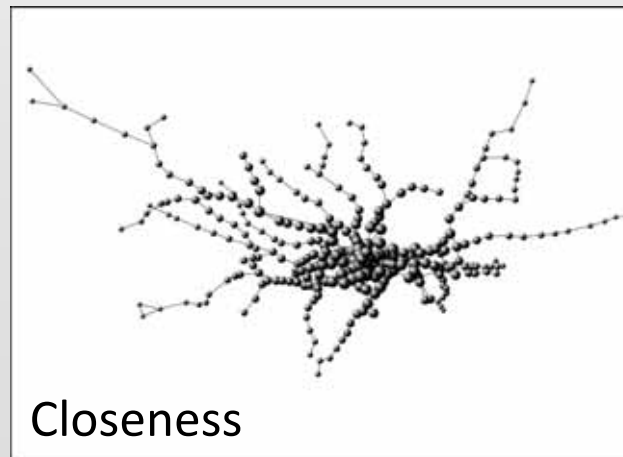
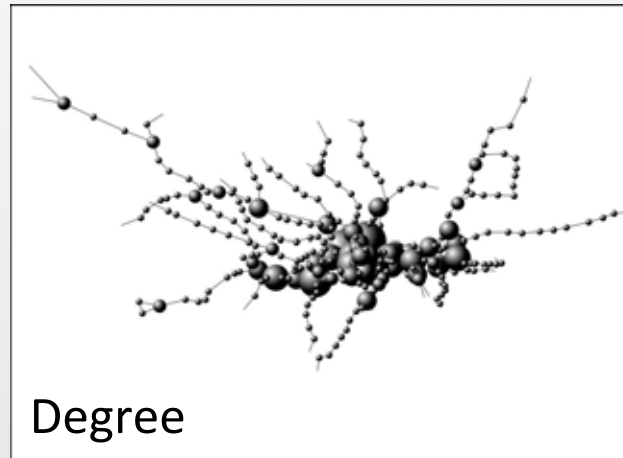
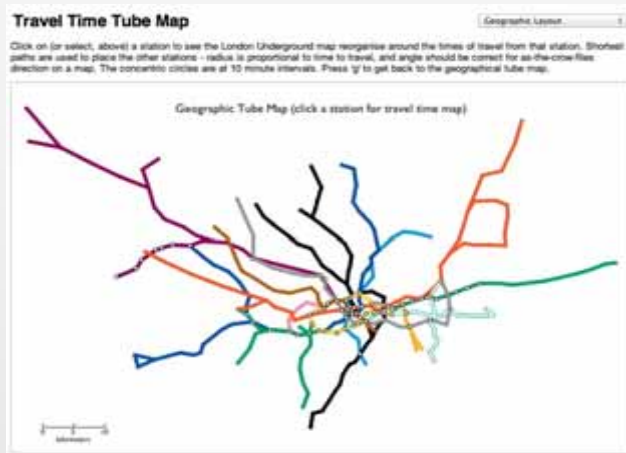
Weighted Betweenness  
Centrality

$$p_{ijk} = \frac{\sigma_{ikj}}{\sigma_{ij}} = \frac{\sigma_{ikj}}{\sum_{\ell} \sigma_{i\ell j}}, \quad \sum_k p_{ikj} = 1$$

$$\tilde{C}_k = \sum_i \sum_j T_{ij} p_{ikj} = \sum_i \sum_j T_{ij} \frac{\sigma_{ikj}}{\sigma_{ij}}$$

# A Preliminary Analysis (1)

## The Minimal Tube Network and the Three Centrality Indices



## A Preliminary Analysis (2)

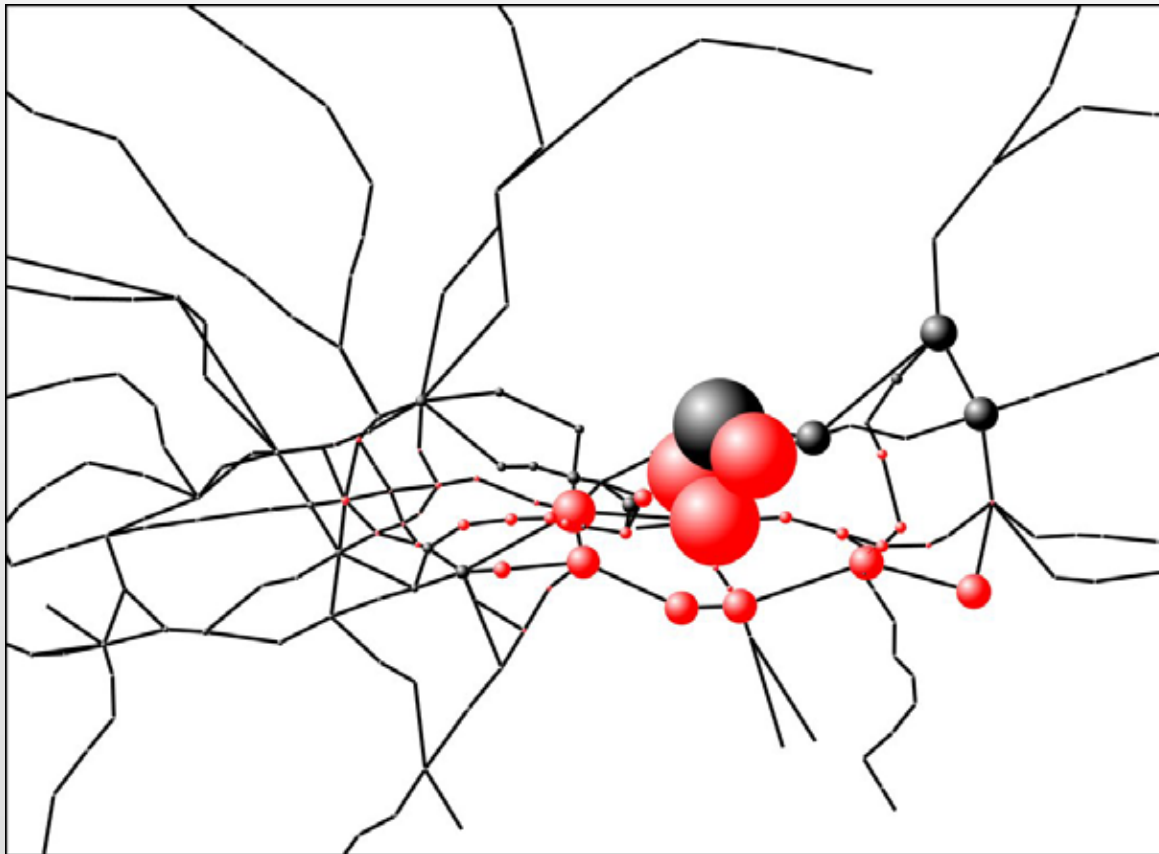
- Top Stations
- By Centrality

Station	$d_i$	Station	$\hat{C}_i$	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	<b>Liverpool Street</b>	11186	Baker Street	2.071
Canning Town	5	Stratford	10814	Victoria	2.065
<b>Liverpool Street</b>	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
Turnham 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	<b>Liverpool Street</b>	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



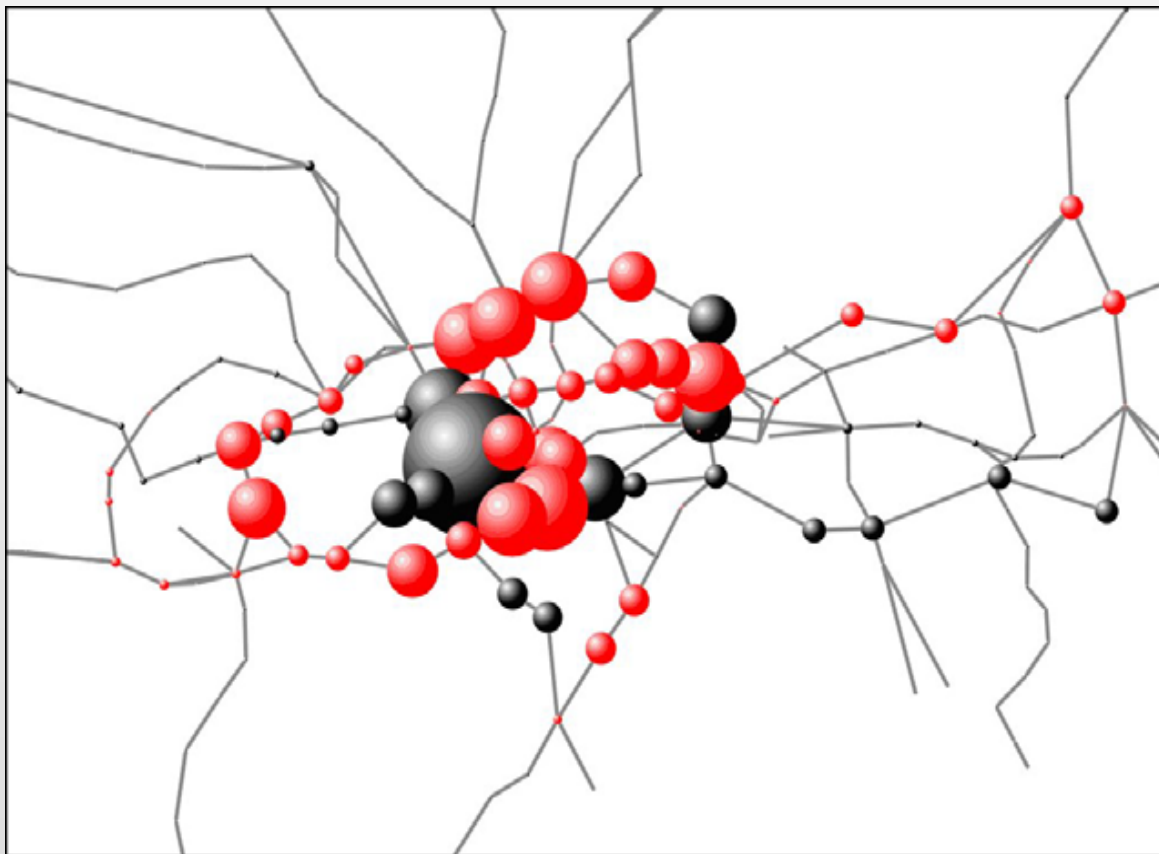
## A Preliminary Analysis (3)

### Closing Liverpool Street

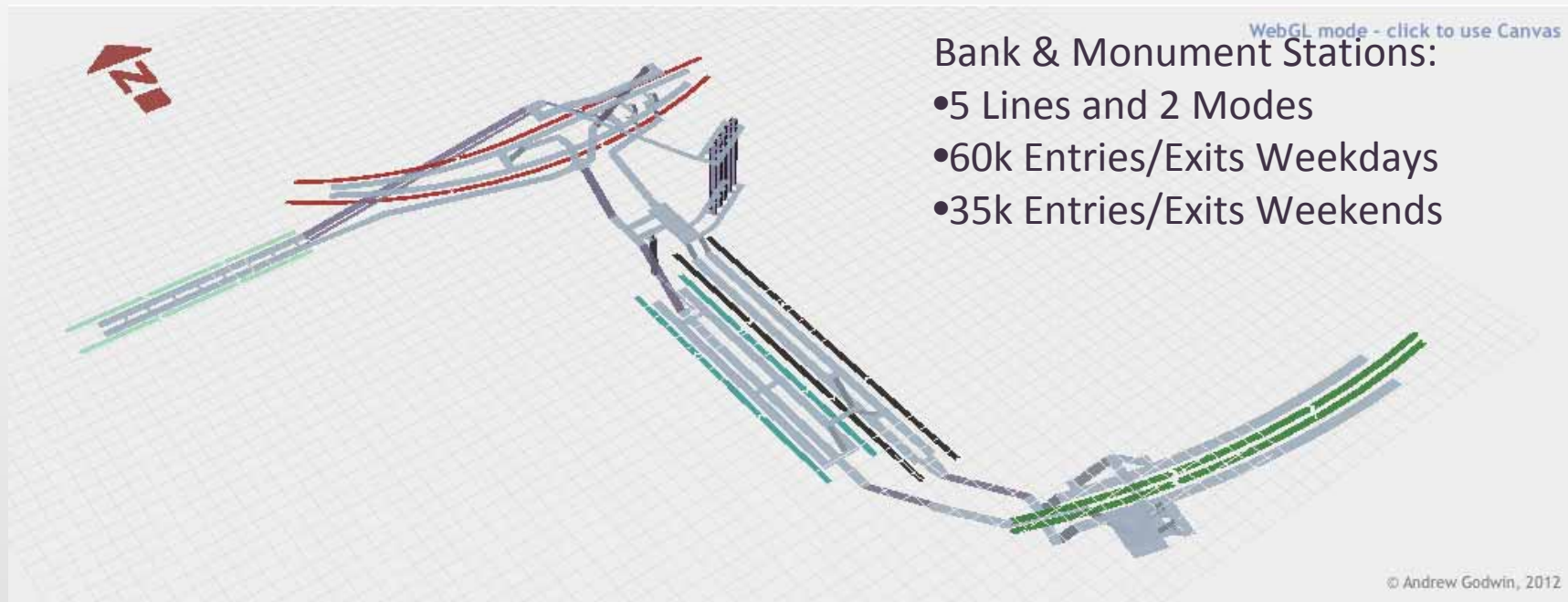


## A Preliminary Analysis (3)

### Closing Green Park



## Looking at Station Closures and Shifts of Travellers Using The 'Shortest'-Paths



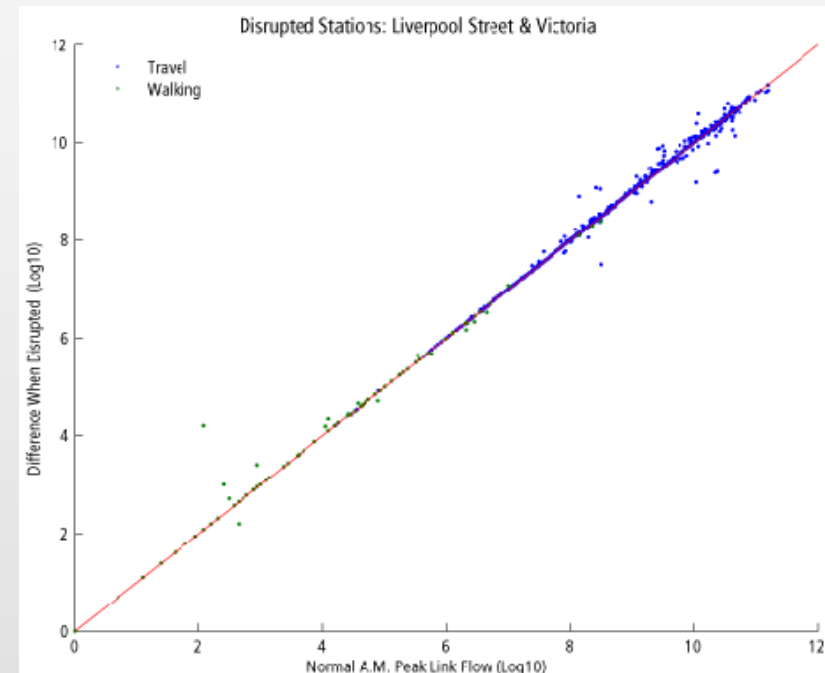
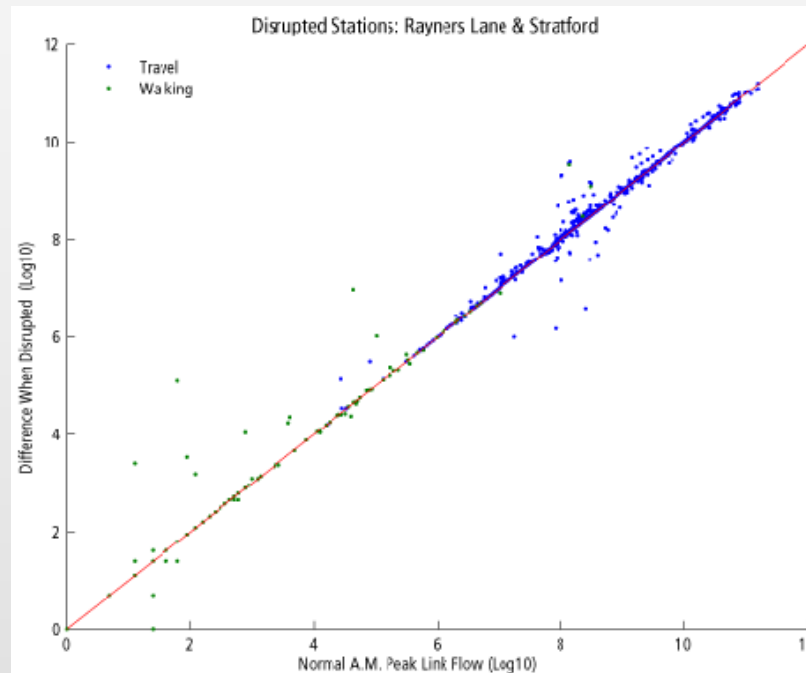
Although a simple station/line network may be sufficient for small cities, for 'Mega-Cities' such as London, New York, or Tokyo a much more detailed network is needed with interchanges measured down to the platform level. The 'penalties' for changing lines (and permitted Out-of-Station Interchanges)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

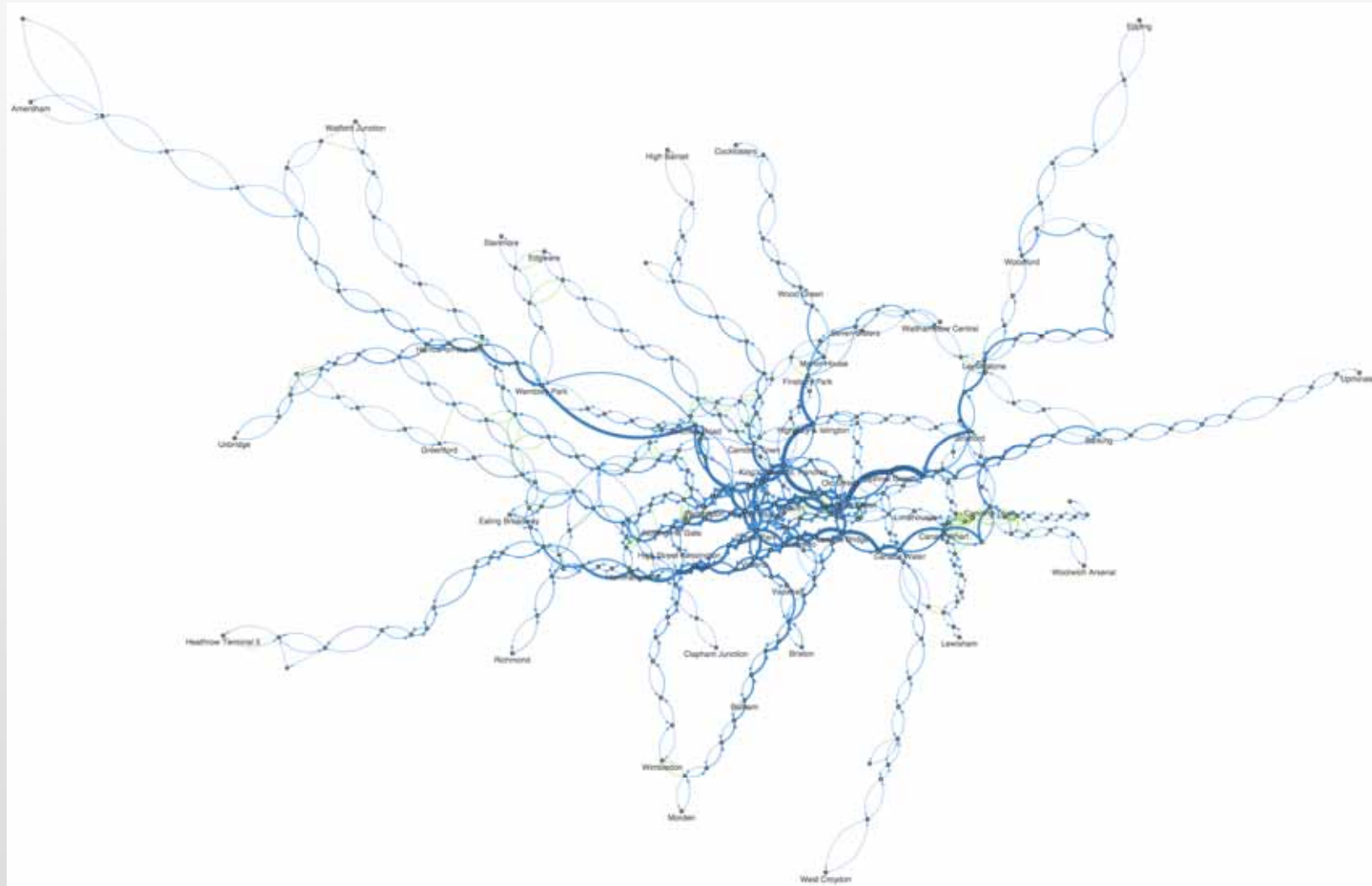


# Link-Level Disruption

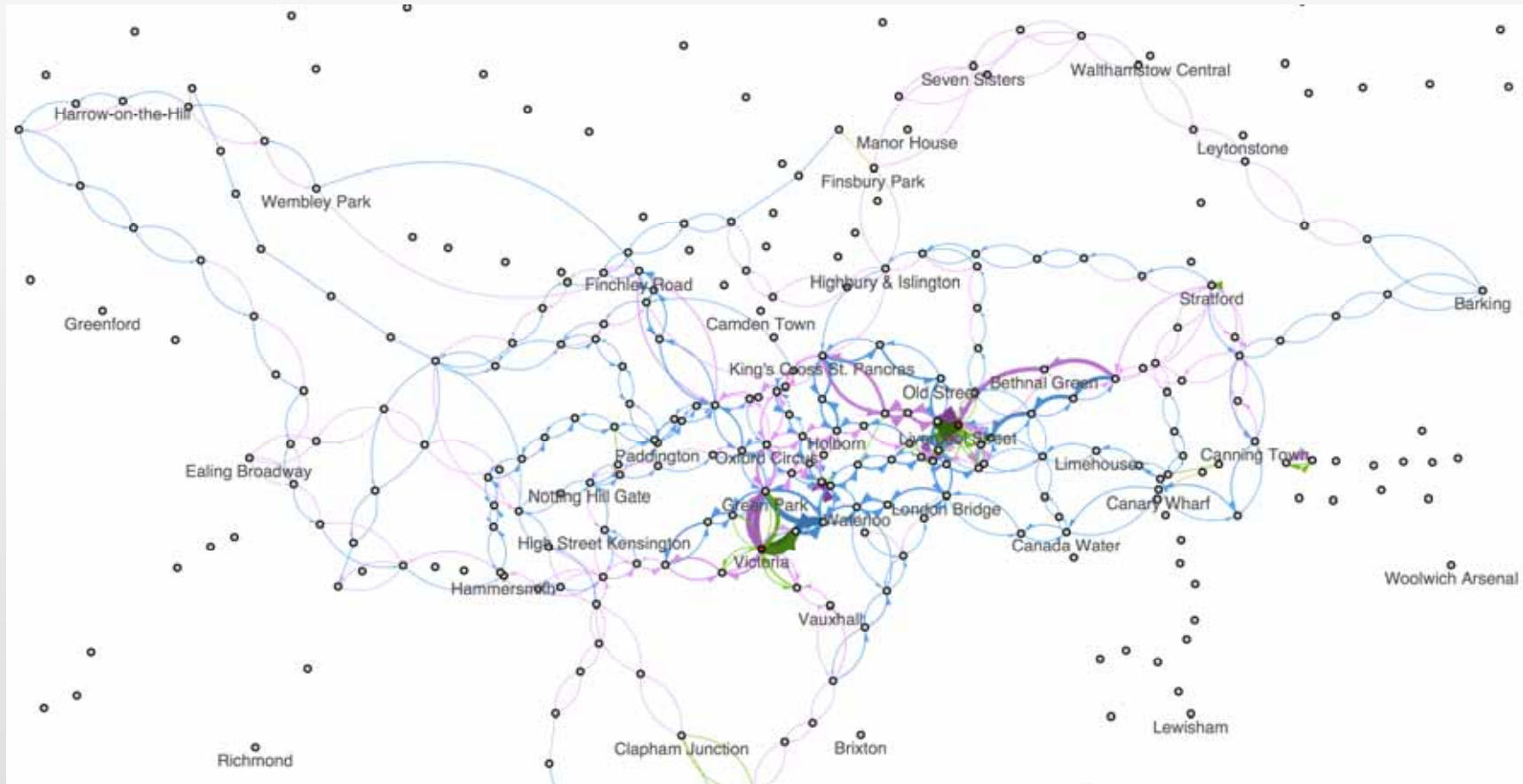


Single- and dual-station disruptions produce unexpected link-level interactions: changes in shortest-path typically cause some links to lose passengers, and gains are often less than expected. Moreover, it is not the biggest and most central stations that cause the largest shifts!

# The Undisturbed Network

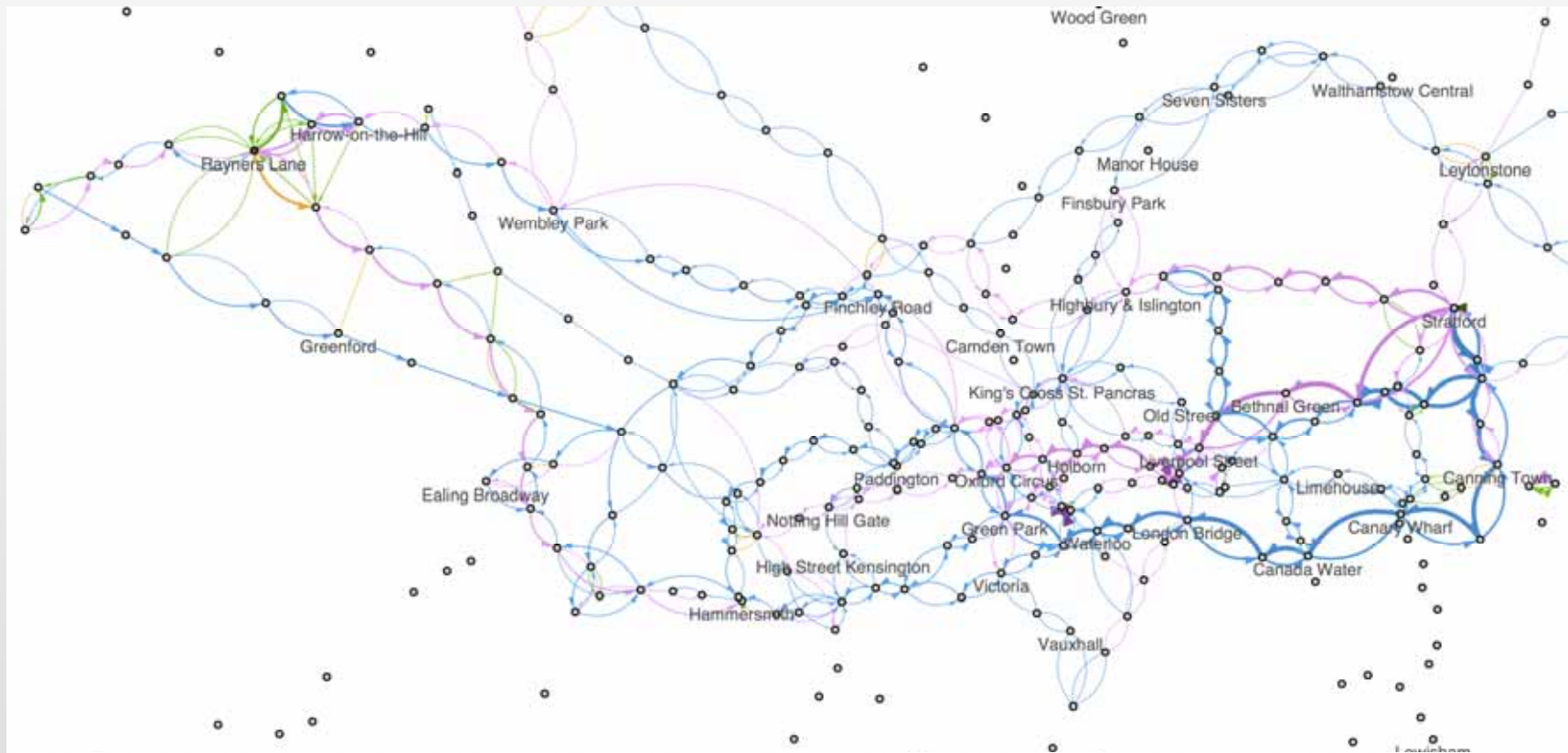


## Liverpool Street & Victoria



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.

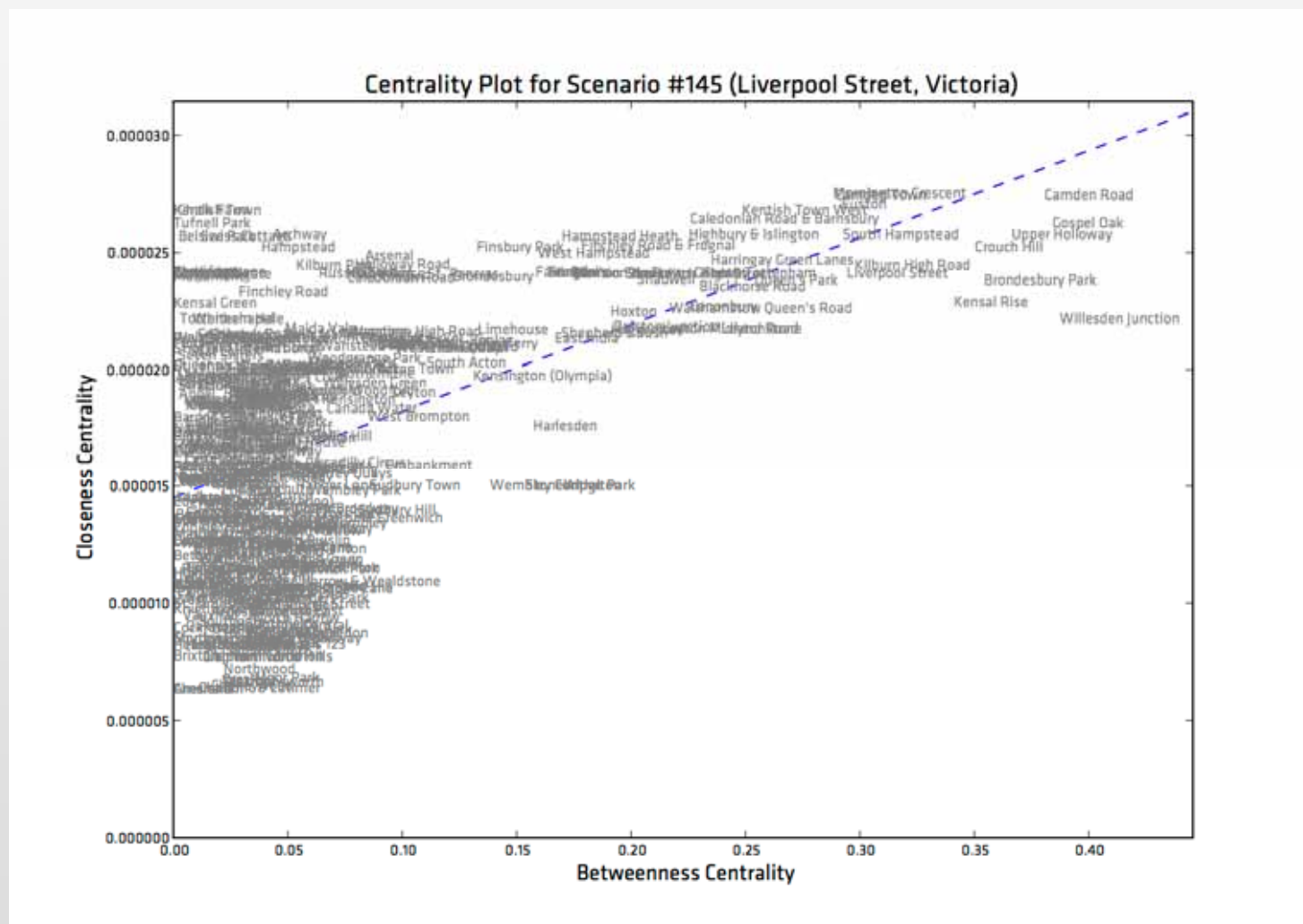
## Rayners Lane & Stratford



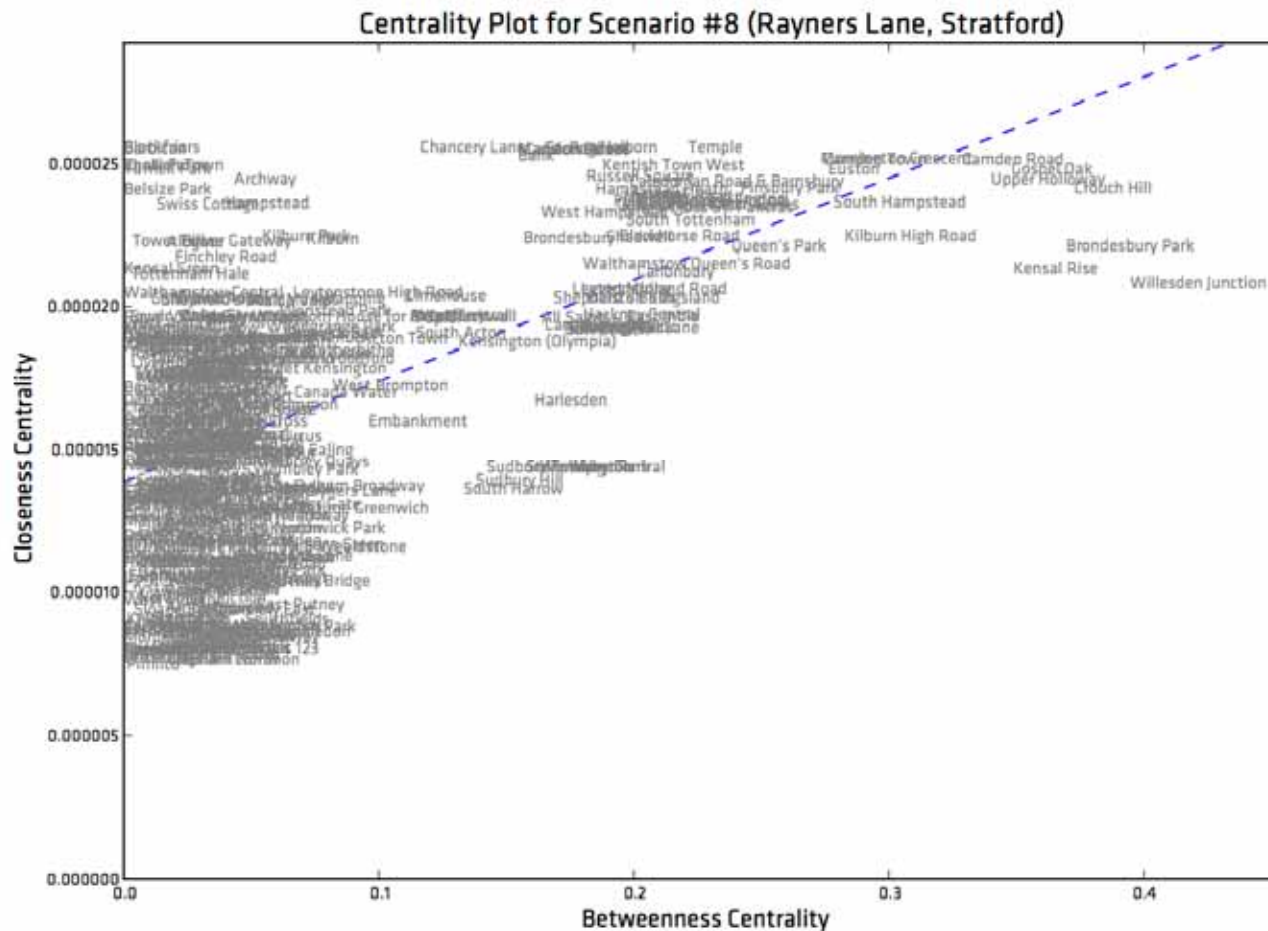
Secondary interchanges outside the core seem to cause greater disruption. Major re-routing required to complete journey, and time lost to walking long distances or travelling via more circuitous routes is much greater.



## Betweenness & Closeness: Liverpool Street & Victoria

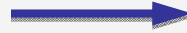


## Betweenness & Closeness: Rayners Lane & Stratford



## Disruption: Examining Traveller Disruption: Stalled Trains

*No  
Change*



*Change  
Origin*



*Change  
Destination*



*Change  
Mode*

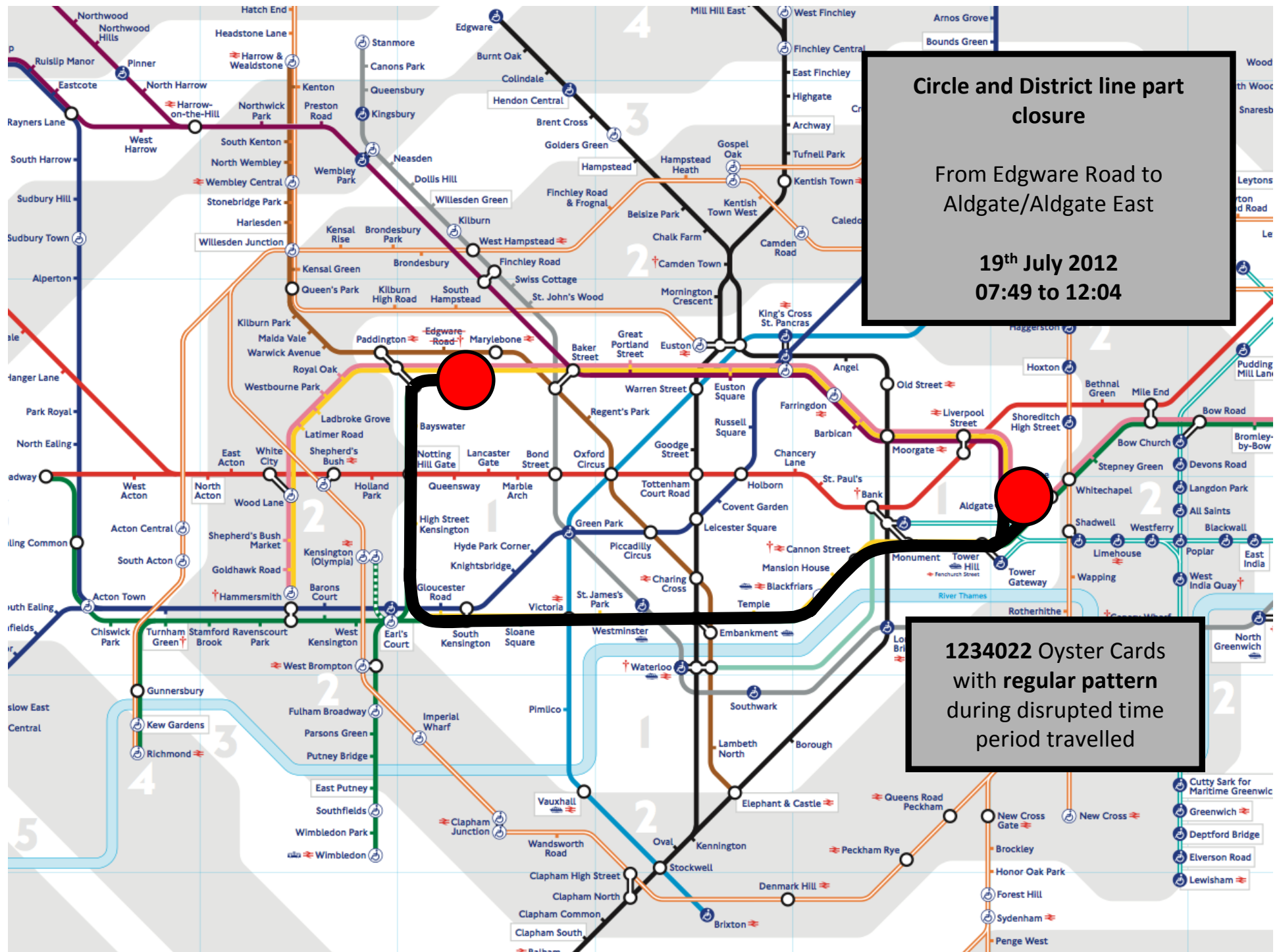


**Circle and District line part  
closure**

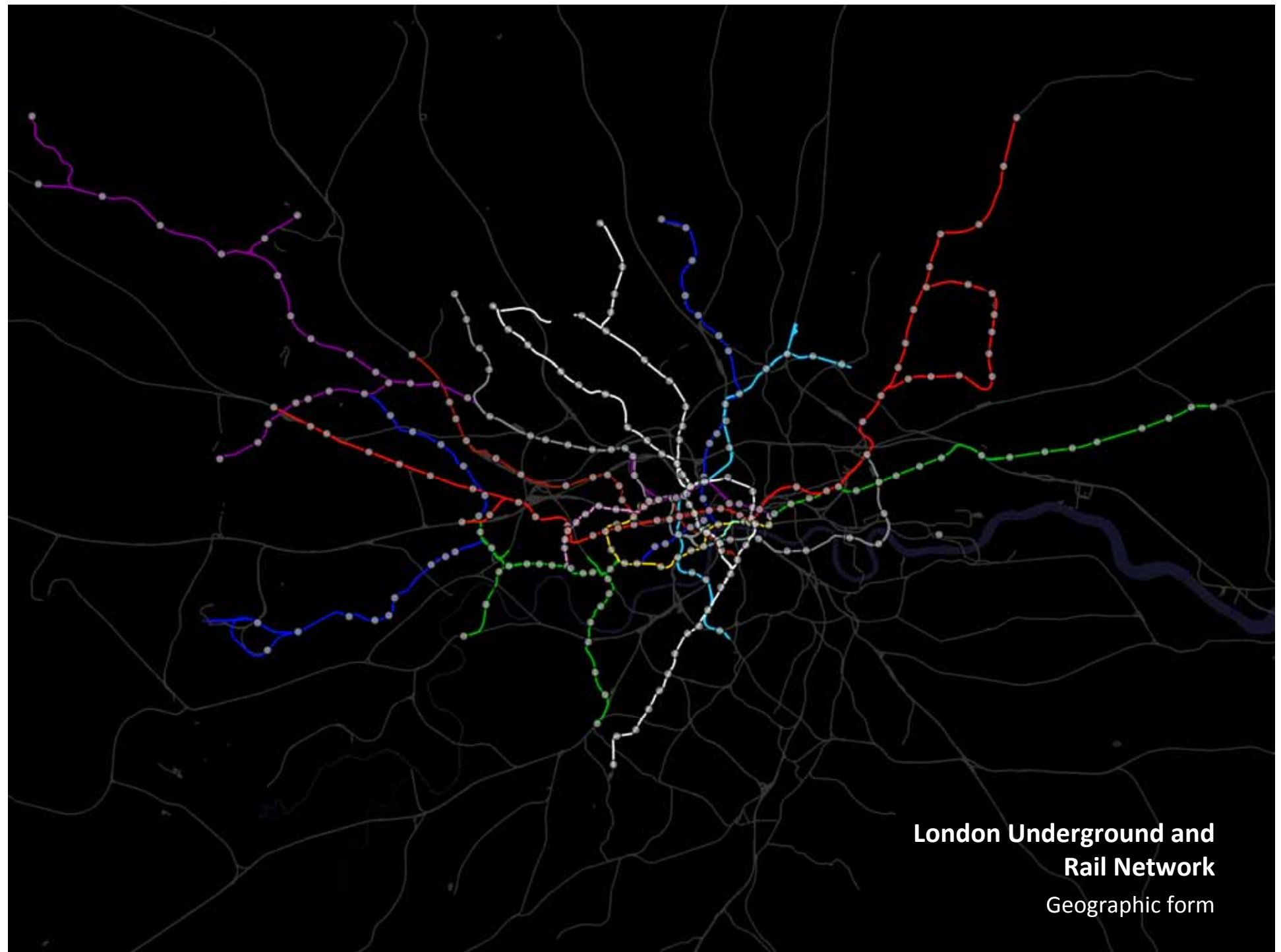
From Edgware Road to  
Aldgate/Aldgate East

**19<sup>th</sup> July 2012  
07:49 to 12:04**

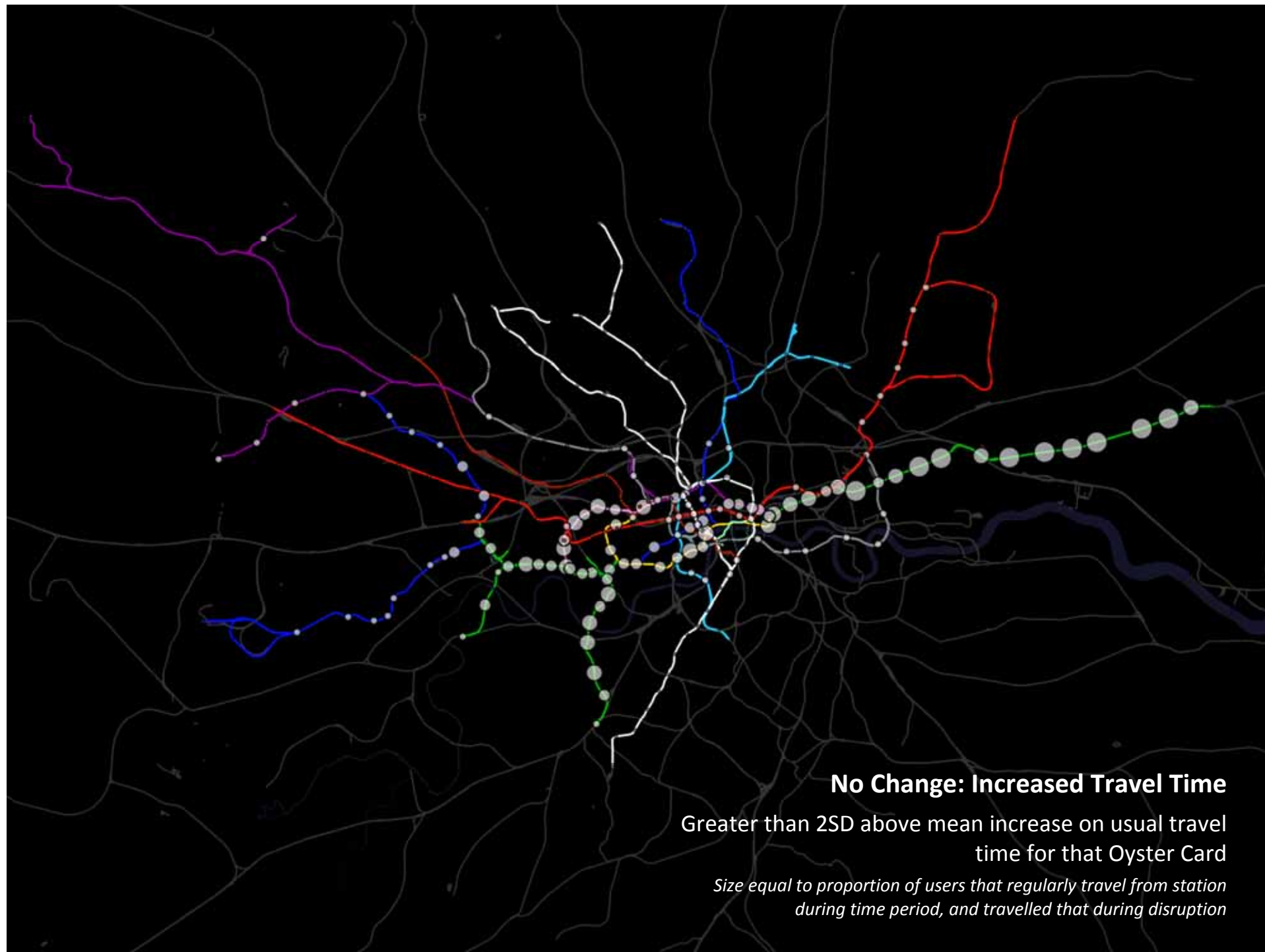
**1234022 Oyster Cards  
with regular pattern  
during disrupted time  
period travelled**

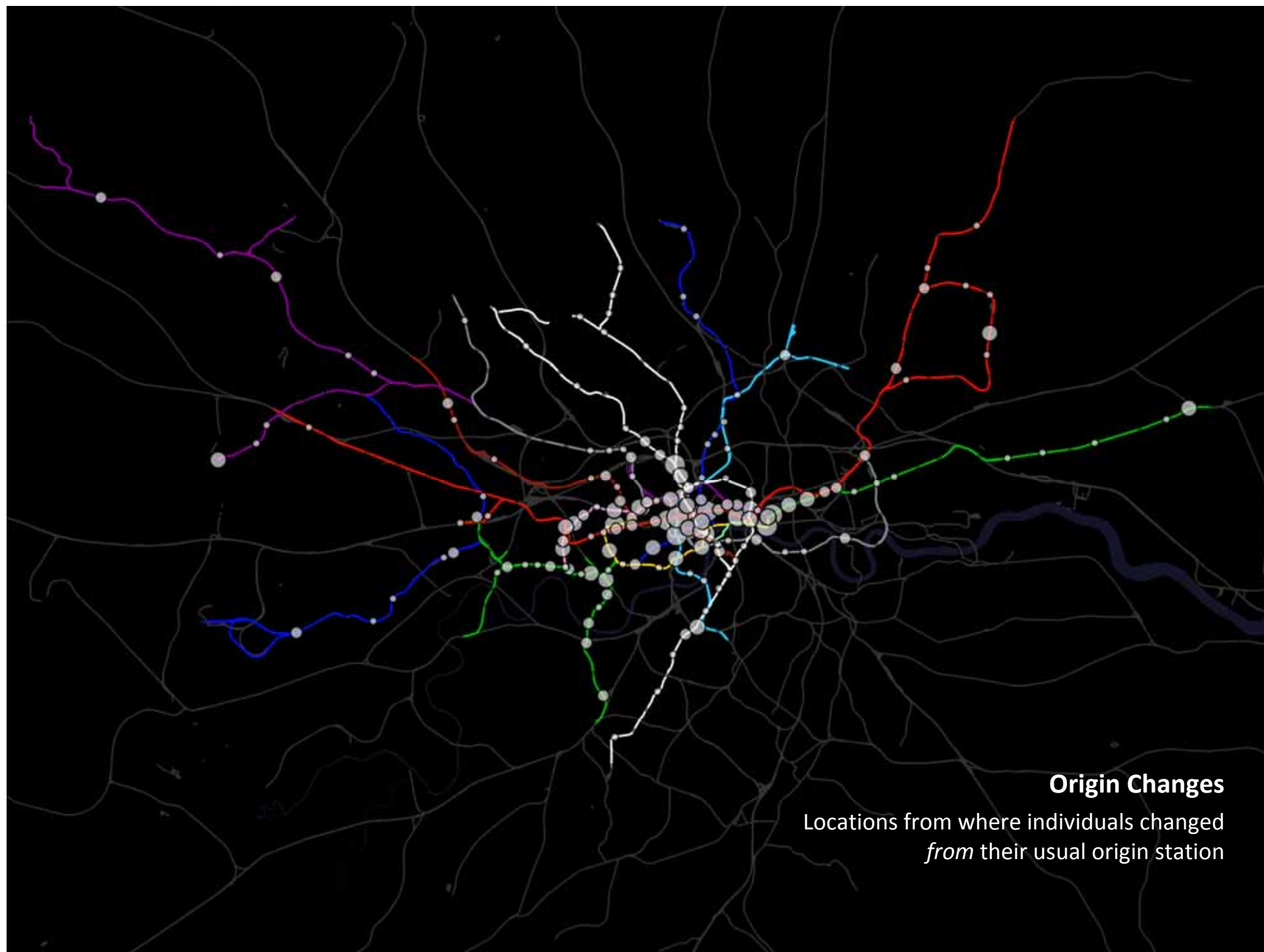


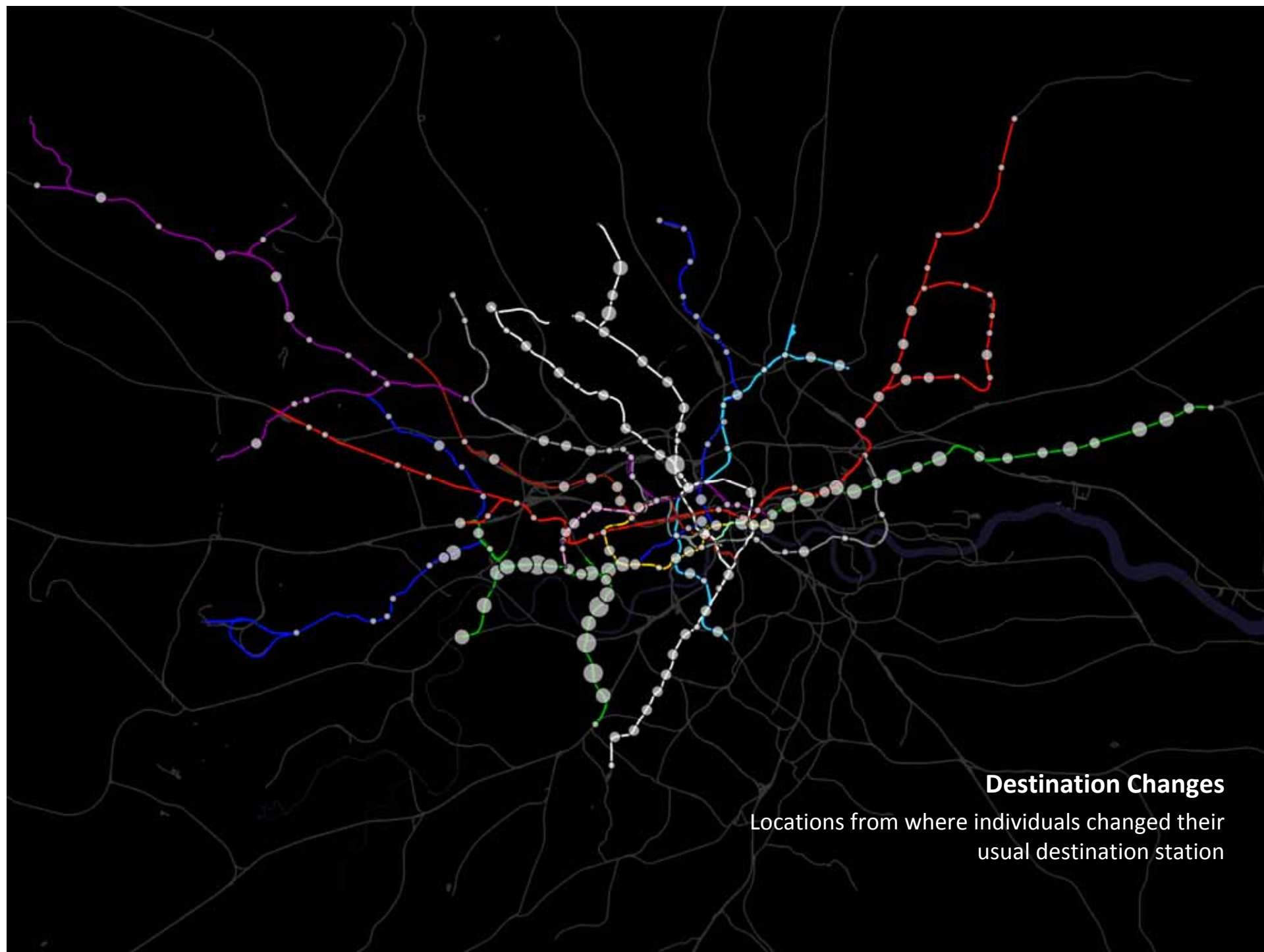




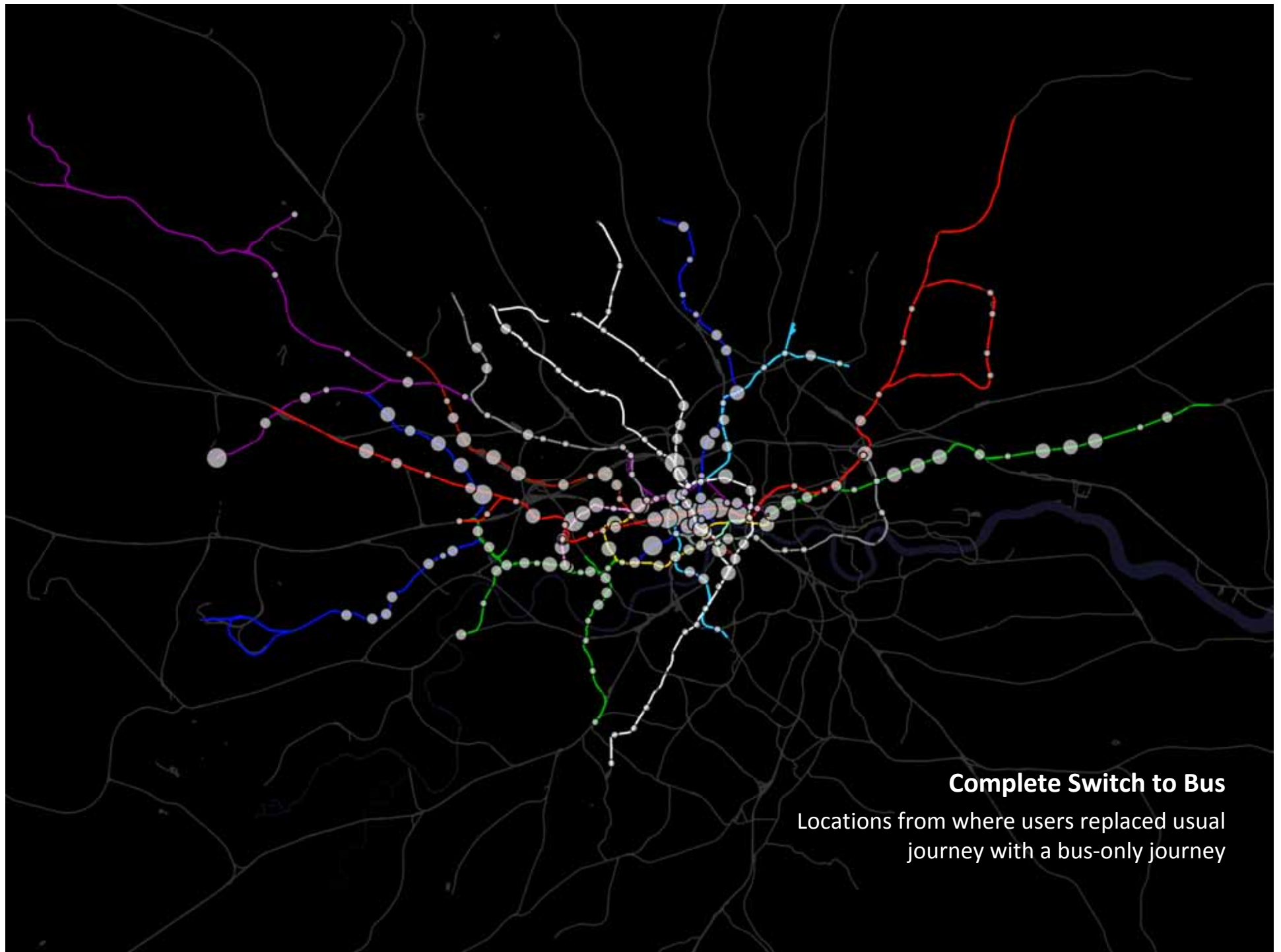
**London Underground and  
Rail Network**  
Geographic form

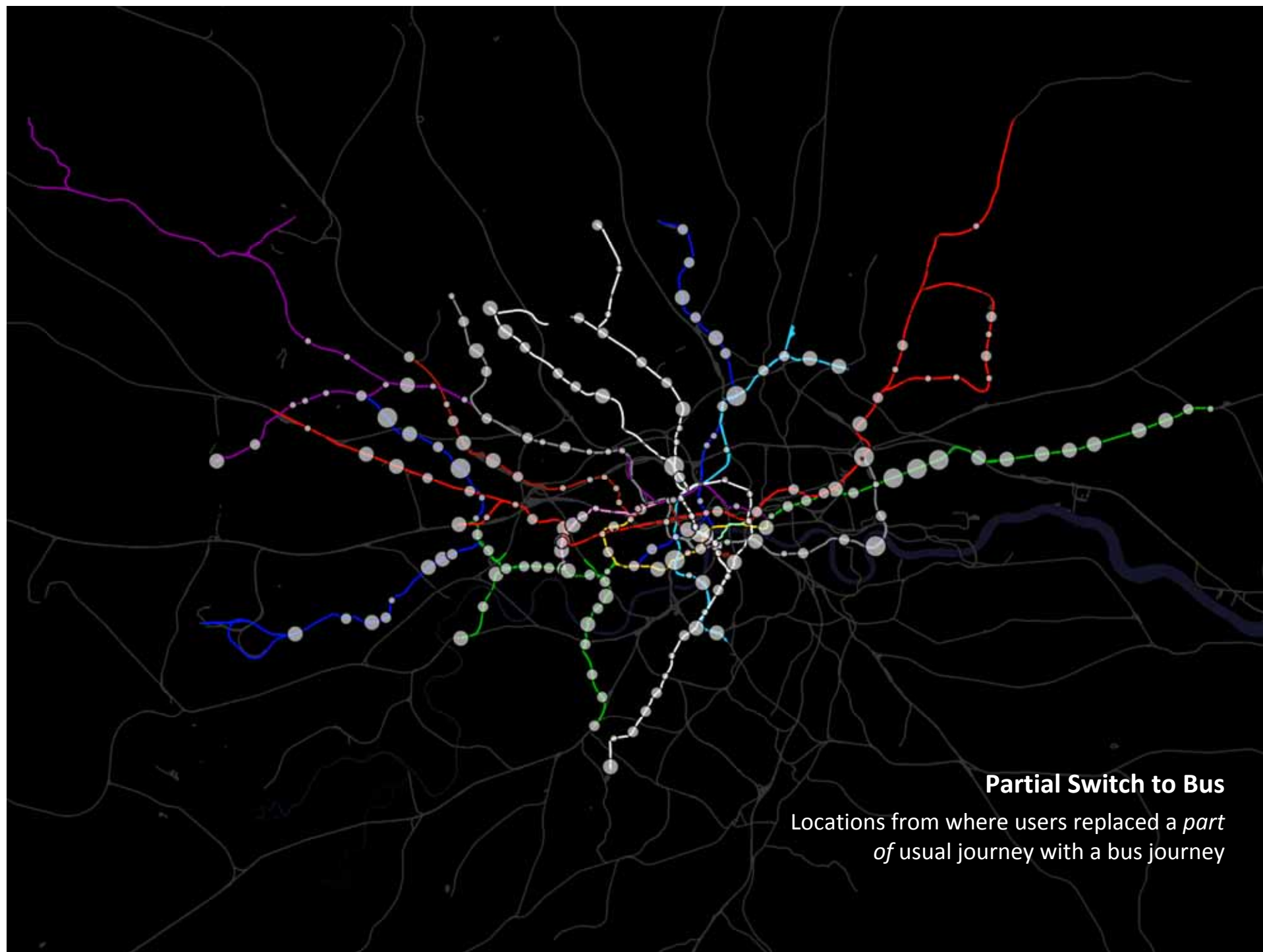


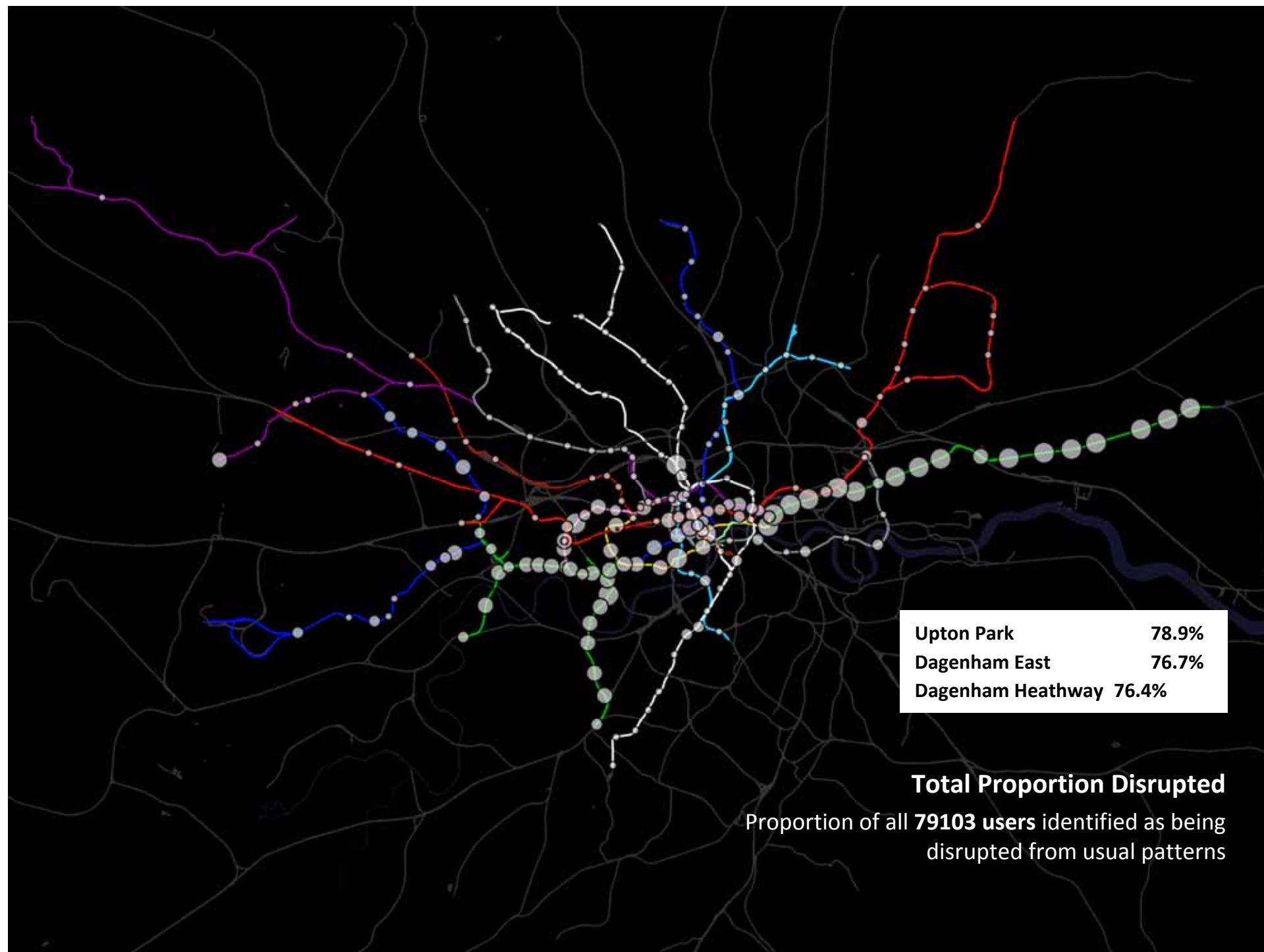












## Disruption Analysis Some Key Points

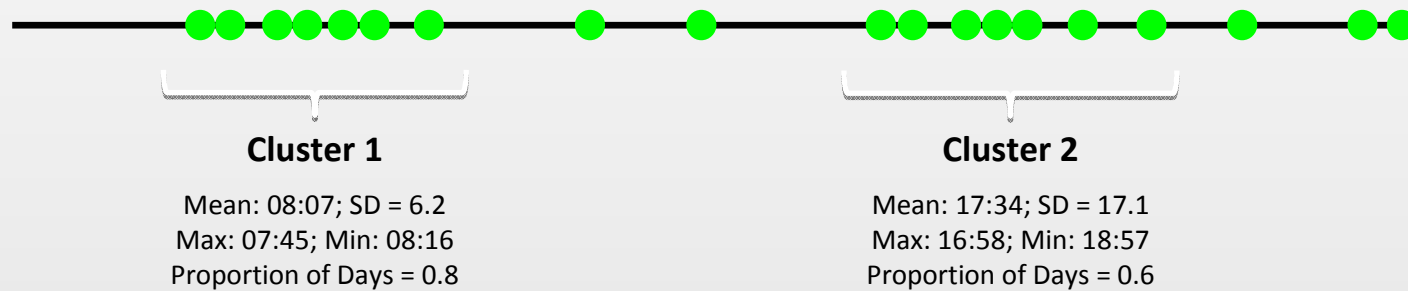
- Behaviours vary across network
- Different areas of network more resilient to disruption, due to available infrastructure and individual ability to change
- But areas of network are naturally closely tied through established usage patterns
- Individual-based analyses provide insight into behaviours underlying macroscopic flows



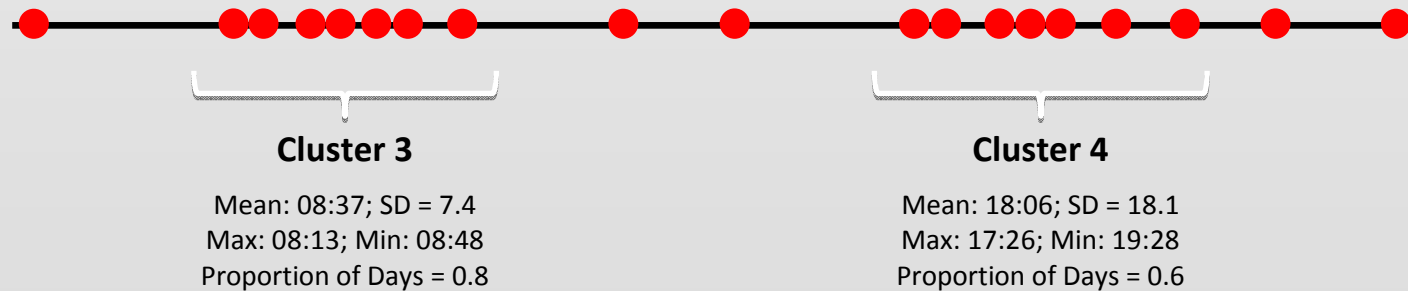
# Measuring Regularity

## Version 2: DBSCAN Method

Oyster Card A – Origin 747

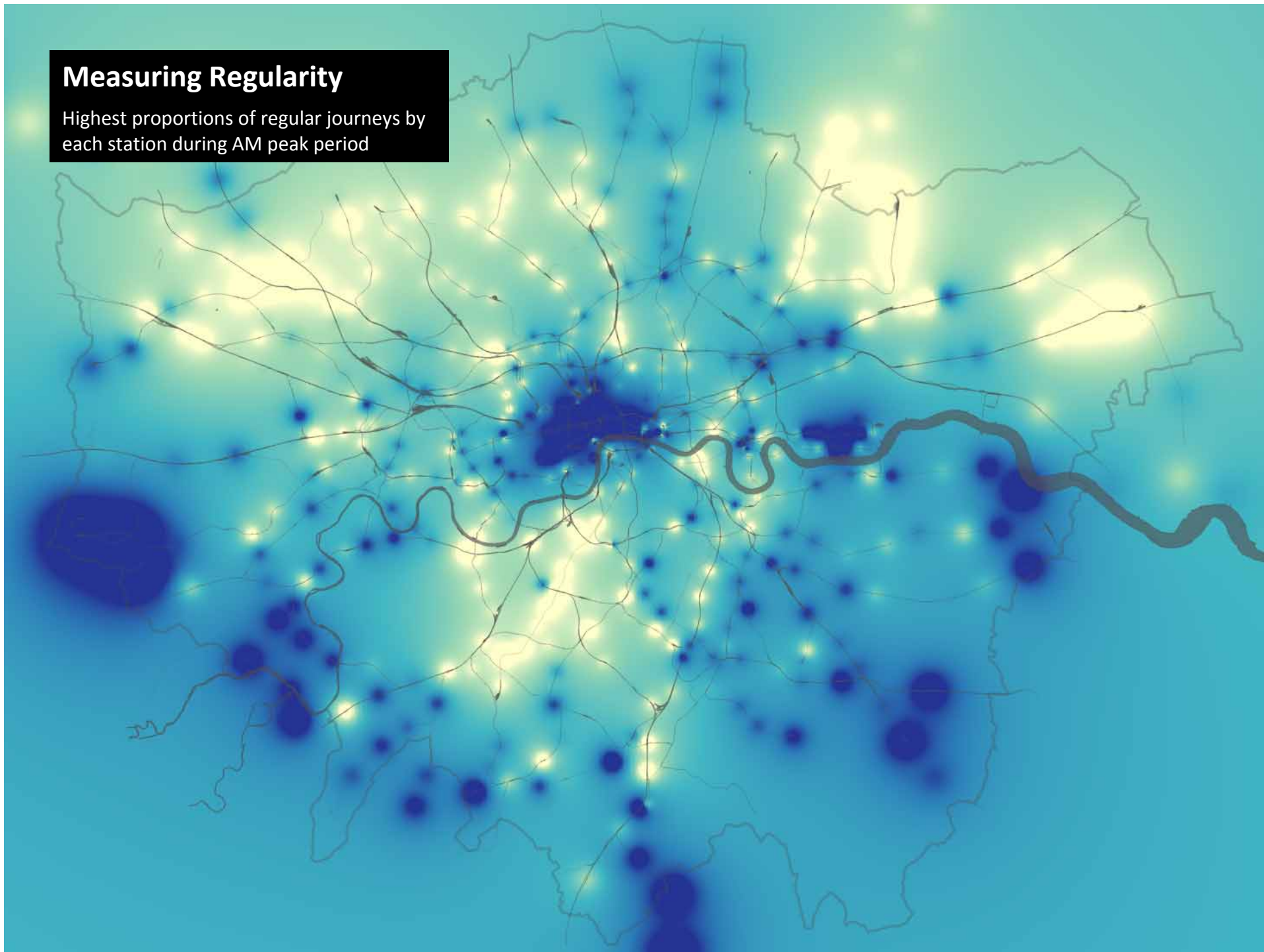


Oyster Card A – Destination 647



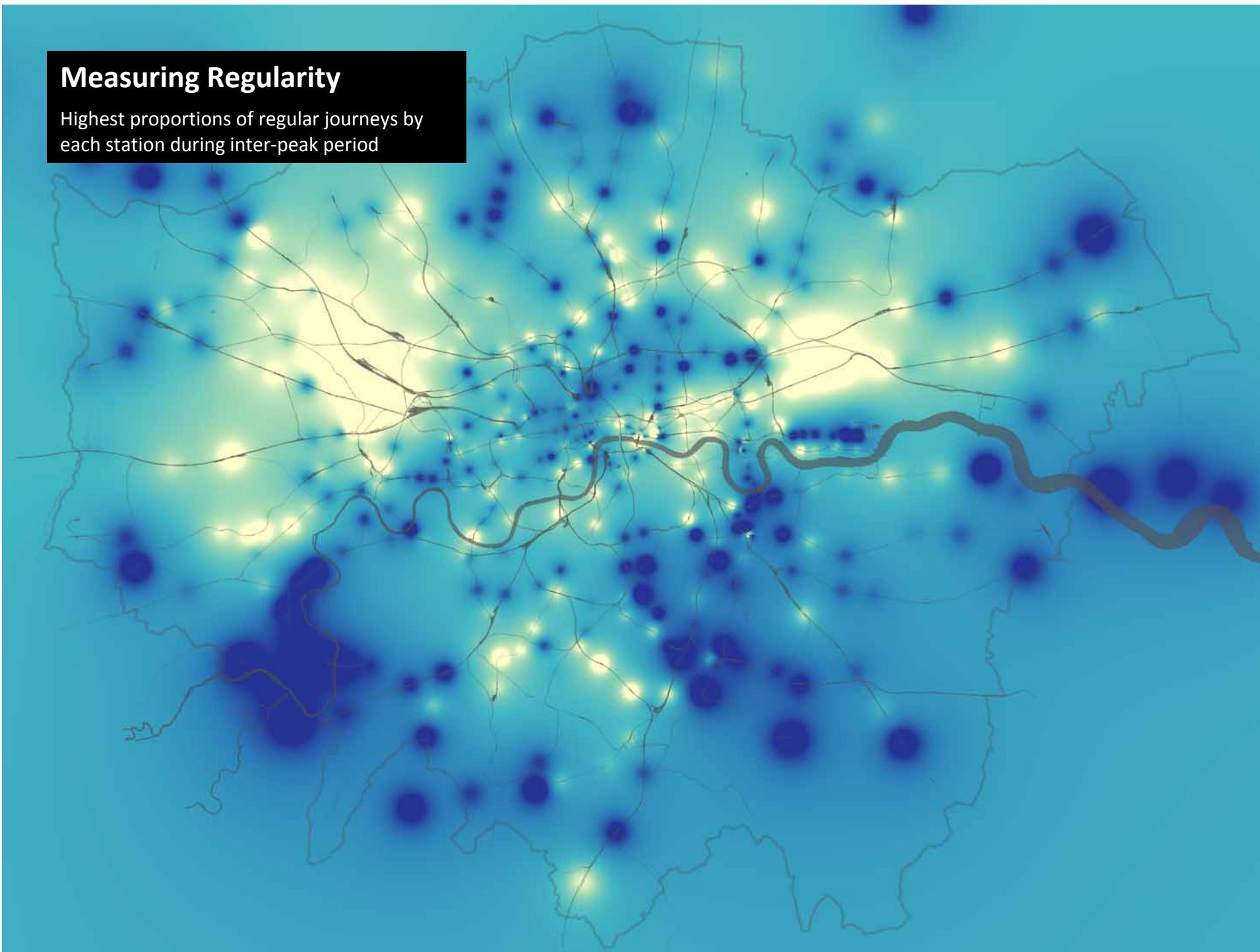
## Measuring Regularity

Highest proportions of regular journeys by each station during AM peak period



## Measuring Regularity

Highest proportions of regular journeys by each station during inter-peak period



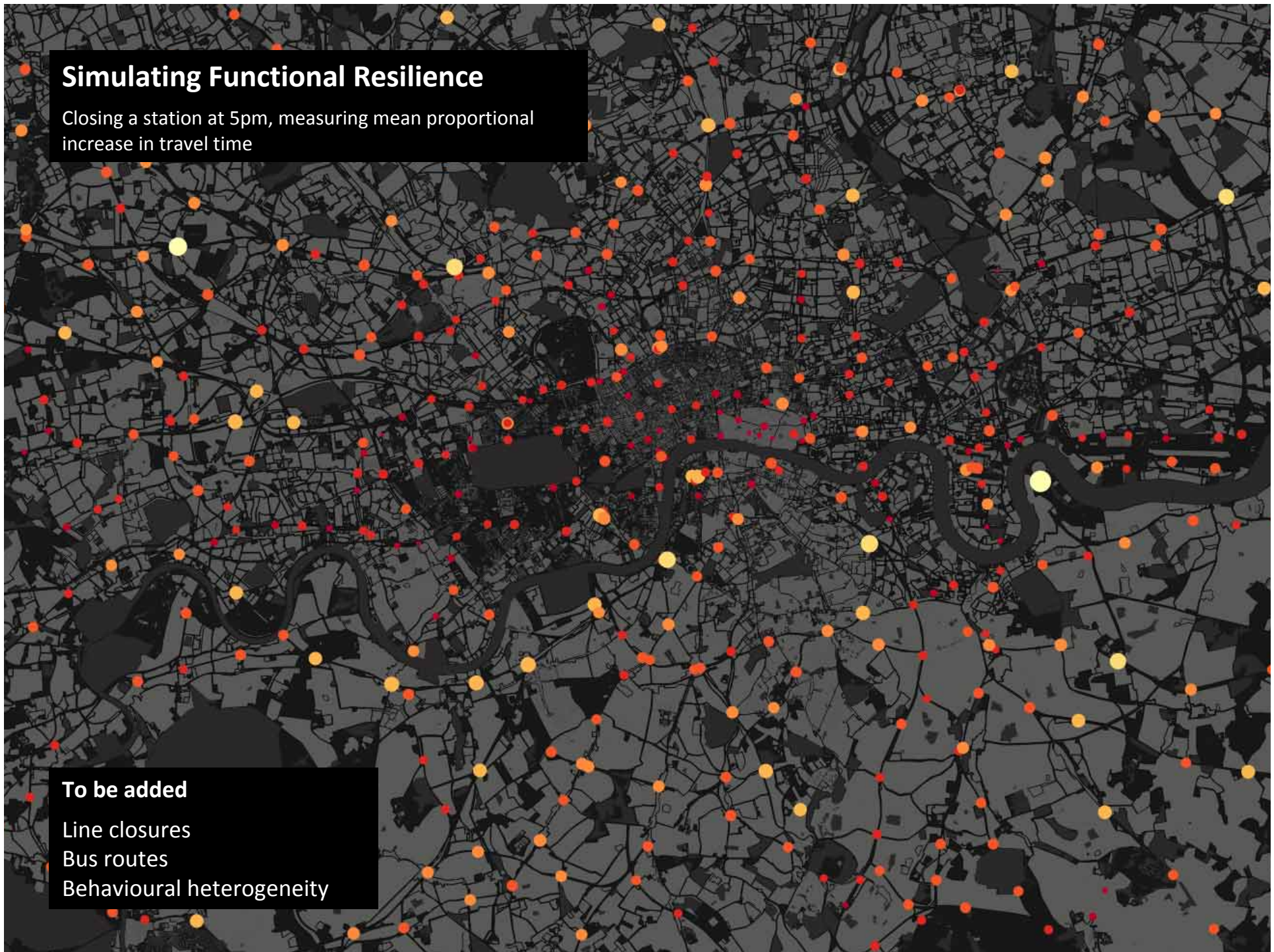


## Simulating Functional Resilience

Closing a station at 5pm, measuring mean proportional increase in travel time

### To be added

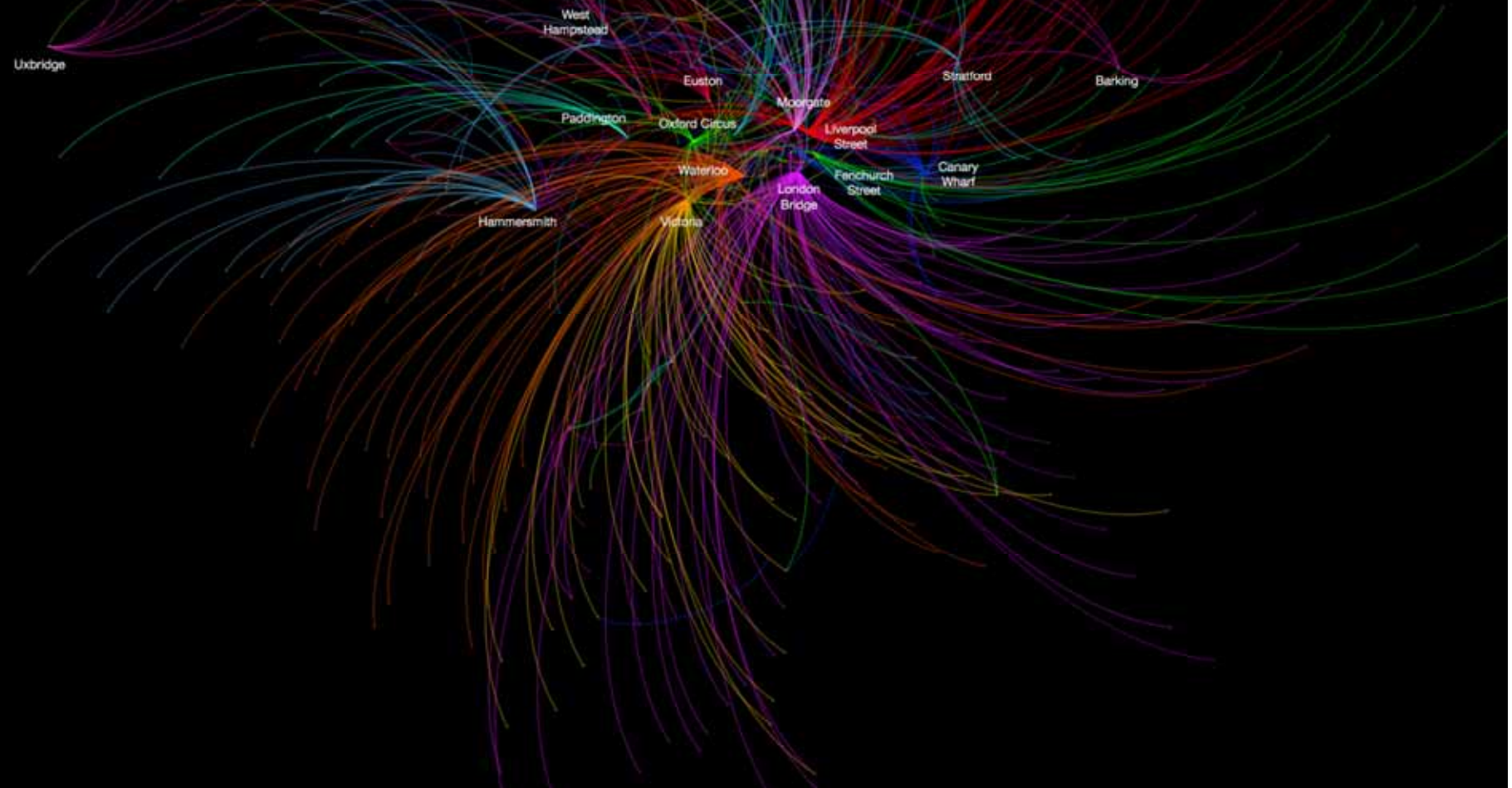
- Line closures
- Bus routes
- Behavioural heterogeneity





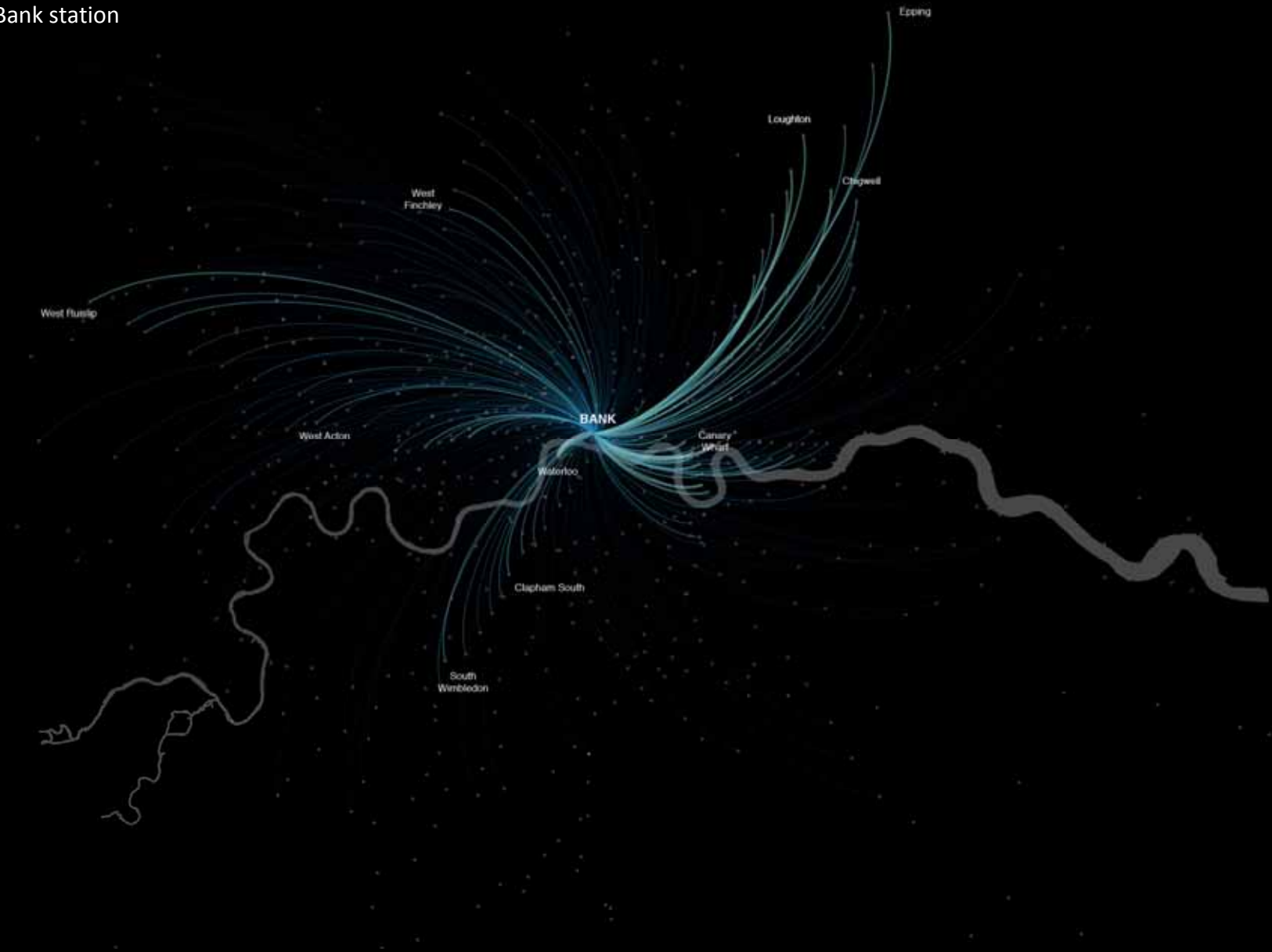
# Visualising Functional Resilience

Most Popular Destinations by Origin



# Visualising Functional Resilience

Reliance on Bank station



# Visualising Functional Resilience

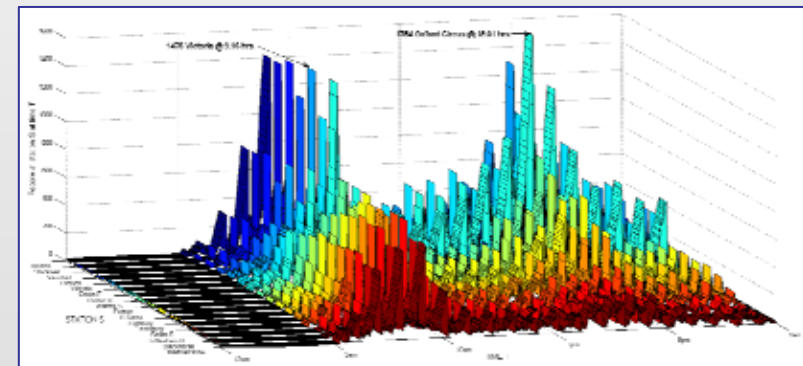
Reliance on Westminster station



# Daily Polycentricity: Information Flow

We are currently using information theory to figure out how much information from trips is transmitted from station to station through time by working out how many passengers are in stations or on trains in stations over time. We are using the concept of **transfer entropy** to do this. I don't have time to say much about this but here is a picture about this

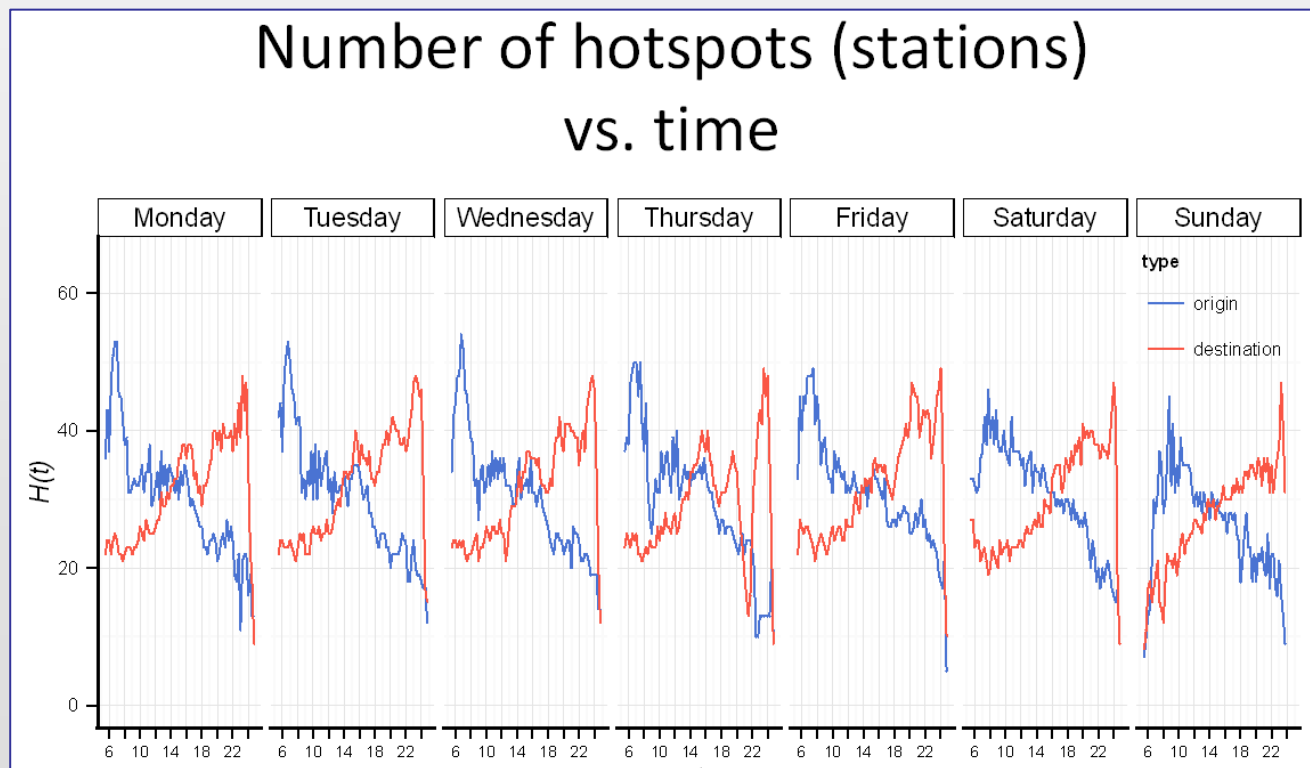
		Stations S																Euston @ 8.29am	
Time t		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16		
		196	61	211	132	400	152	488	358	252	252	188	170	128	180	144	159		
8.30am	181	164	199	138	304	144	489	415	210	647	228	157	218	76	55	160			
	178	165	222	179	288	188	395	280	604	274	211	137	178	178	111	137			
	173	128	232	91	276	167	326	206	299	246	180	243	227	65	72	153			
	183	114	217	132	322	279	342	562	244	278	143	153	111	135	148	130			
	252	142	230	87	275	212	614	313	214	264	262	218	222	103	72	156			
	171	85	216	162	375	203	469	257	251	255	136	183	82	201	140	128			
	162	205	206	161	299	372	424	286	206	158	193	248	116	132	60	165			
	124	170	215	176	366	765	422	301	246	312	181	200	106	241	121	158			
	143	171	143	138	470	200	427	262	262	170	202	254	223	147	46	207			
	206	117	236	161	381	315	431	190	178	241	201	133	142	213	120	186			
	250	125	224	138	320	753	310	325	223	231	292	144	239	149	42	202			
	224	86	277	163	367	221	467	216	202	281	173	123	170	209	112	149			
	210	176	206	173	313	210	400	301	242	232	164	203	206	121	56	162			
	209	224	216	178	297	273	447	266	192	361	138	177	176	200	164	94			
	243	172	185	102	299	143	452	292	313	223	181	257	176	90	73	110			
	256	189	275	127	348	216	520	263	220	218	163	211	95	179	162	112			
	327	200	265	80	271	220	475	376	191	203	254	194	211	110	62	145			
	271	118	198	148	312	239	549	270	198	221	213	179	119	208	119	142			
	259	250	218	210	258	188	499	276	235	163	216	183	213	92	61	141			
8.50am	310	204	263	163	338	286	508	285	170	288	252	135	162	176	97	171			

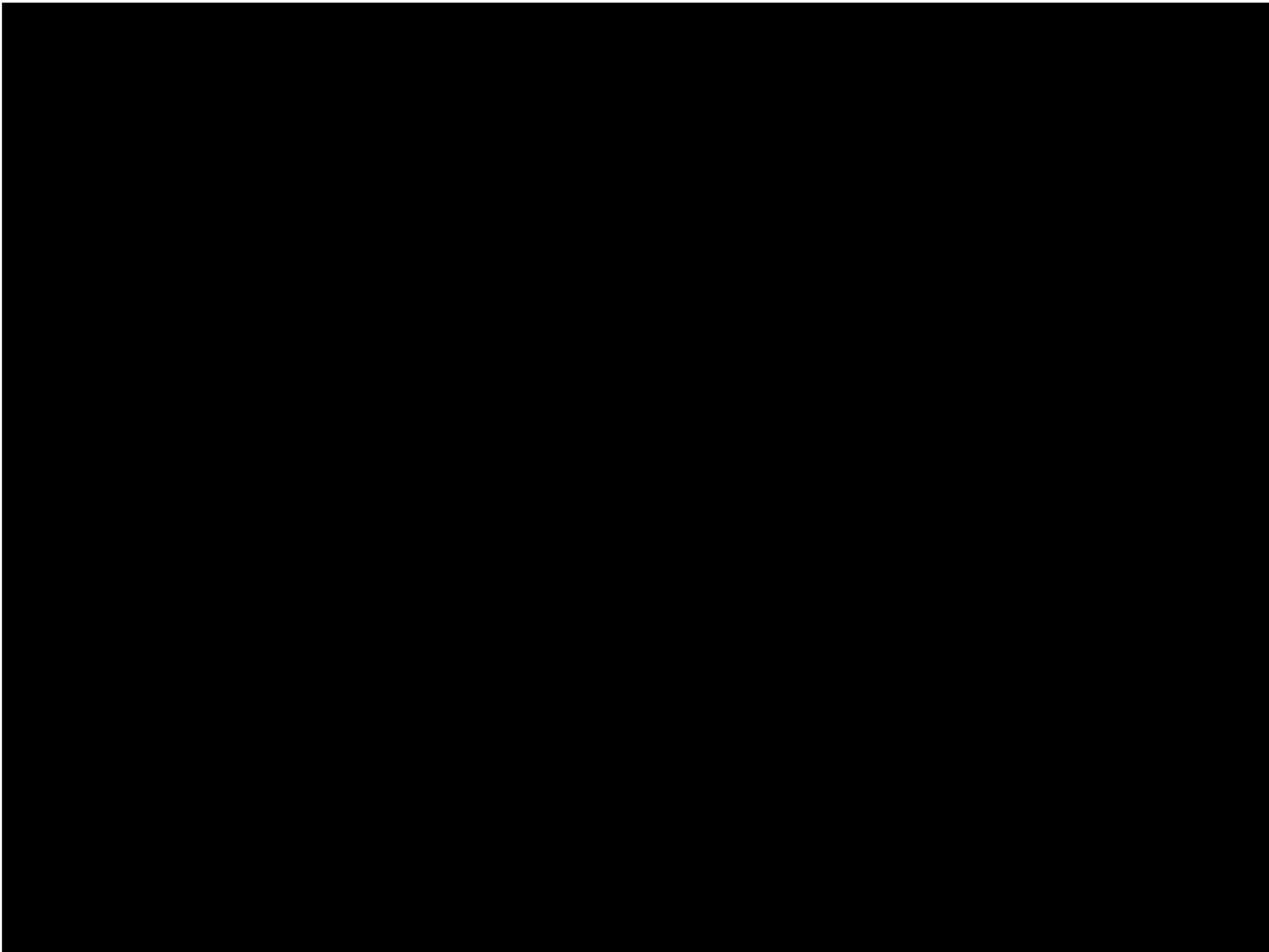


$$T_{YX} = \sum_{t=1} p(y_{t+1}, y_t, x_t) \log \frac{p(y_{t+1}|y_t, x_t)}{p(y_{t+1}|y_t)}$$



Second we are working with the Oyster data again with Melanie Bosredon in our group and Marc Barthelemy in Paris on extracting clusters from the travel data using a new method of defining intensity. I will show this as a simple movie of origin and destination intensities as they change over time of day.





And to finish this section – let me show you a movie where we combine this data with our 3D model and build a rather beautiful 3D moving environment that has all the problems of knowing how we should interpret different kinds of cycles in time and space

This is the kind of thing that is possible with this kind of big data and the current methods for visualisation – in fact in the live version you can query the movers and pull up some attribute data on them as the movie plays

*Courtesy of Gareth Simons, MRes Student CASA, 2014*

