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# Modelling urban networks: some results and their limitations

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## Introduction

#### What this talk is about

- How is human movement in an urban setting conditioned by the topology of the transportation networks?
- Are there any features common between large cities?
- How do quantitative results relate to the practice of urban planning?

#### My aim

- Present recent results on urban transportation, joint work with Saray Shai, Emanuele Strano, and Marc Barthélemy
- Real-world data driving a study using network science
- The limitations we hit when theory met practice

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## **Complex networks**

## Networks and processes



#### Living between regularity and randomness <sup>1</sup>

- Heterogeneous degree distribution, fragile notion of "neighbourhood"
- Evaluate processes at each node, affecting behaviour of neighbours, often with a stochastic component
- Canonical example is the SIR model of disease propagation

A.-L. Barabási and R. Albert. Emergence of scaling in random networks. Science, 286(5439):509–512, 1999.

 URL doi://10.1126/science.286.5439.509

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## **Multiplexes**

A *multiplex* (or more correctly a *multilayer network*) is a collection of two (or more) networks <sup>2</sup>

- Nodes in the different networks are *coupled*
- Study properties of the individual networks or of the ensemble
- One network may be "less wide" than the other, and so offer "shortcuts" for processes



<sup>&</sup>lt;sup>2</sup> M. Kivelä, A. Arenas, M. Barthélemy, J. Gleeson, Y. Moreno, and M. Porter. Multilayer networks. *Journal of Complex Networks*, 2(3):203–271, 2014. URL doi://10.1093/comnet/cnu016 > ()) > ())

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## Spatial networks

Planarity limits permissible topologies <sup>3</sup>

- Network embeds into Euclidean 2-space ( $\mathbb{R}^2$ )
- No crossings: all intersections form junctions
- (Doesn't work precisely for all cities, *e.g.*, Edinburgh, which have significant 3D structure)
- Limits the possibility for long-distance connections
- Typically quite *modular*, with highly-connected locales

#### Spatial multilayer networks

• Each *layer* is planar, but the *multiplex* typically isn't

<sup>&</sup>lt;sup>3</sup>M. Barthélemy. Spatial networks. *Physics Reports*, 499:1–101, 2011 ← □ → ← 클 → ← 클 → ← 클 → → 클 → ク < ↔

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## Urban networks

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## Cities of different sizes and complexities



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## The problem: Urban transportation

### Coupled transport networks <sup>4</sup>

- Street and tube/subway form a multilayer network
- How does the addition of the tube affect travel times?
- How does this change as the tube speeds up?



<sup>&</sup>lt;sup>4</sup>M. Batty. The new science of cities. MIT Press, 2013. ISBN 978-0-292-01952-1 ( 西 ) ( T ) ( T

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## Framing the problem

### Our study <sup>5</sup>

- Simplify to treat as a purely topological problem
- Don't model traffic congestion per se

#### **Metrics**

- Impact of tube speed on network usage, travel costs, and shortest paths
- Study the *betweenness centrality* of nodes as the relative speeds of the two networks changes
- How does outreach change?

<sup>&</sup>lt;sup>5</sup>E. Strano, S. Shai, S. Dobson, and M. Barthélemy. Multiplex networks in metropolitan areas: generic features and local effects. *Journal of the Royal Society Interface*, 12(111), October 2015. URL doi://10.1098/rsif.2015.0651

## Methodology

## **Topological properties**

- Compute metrics between all pairs (*i*, *j*) of nodes in the street network V<sub>s</sub>
- Compute ratio of metrics between travel using the streets only *versus* using the whole multiplex

#### Geographical properties

- Network is spatial, so nodes have location in space, and a distance d(i, j) between pairs of street nodes
- Often scale distances according to network diameter,  $\frac{d(i,j)}{\sqrt{A}}$
- Compare network metrics to geographical distances

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## Setting up the study – 1

Acquired street and tube data from Open Street Map

- Street network consisting of *v* ∈ *V<sub>s</sub>* nodes
- Tube network  $V_t$
- · Coupled at access points to form a multiplex

	Ns	$N_t$	Street diameter	Tube diameter
London	325K	263	89km	60km
New York	68K	454	55km	57km

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## Data hygiene

#### Needed substantial manual cleaning

• Streets don't meet, tubes don't come up where they should, ...

#### Choices to be made

- Tubes sometimes emerge mid-street, not at a junction
- Add a pseudo-junction for the tube to be coupled to
- Couple to junction at one end of the street or the other
- Do these choices make a difference?

## Setting up the study -2

#### **Travel costs**

- *τ*<sub>s</sub>(*i*, *j*) the travel cost (in time units) between *i*, *j* ∈ *V*<sub>s</sub> using only street edges
- $\tau_m(i, j)$  the travel cost using the multiplex (street and tube)
- 0 ≤ β ≤ 1 the ratio of speed between street and tube (tube is faster for smaller β)

#### Shortest paths

- σ<sub>i,j</sub> the number of shortest paths between i, j ∈ V<sub>s</sub> using only the street network
- Similarly define σ<sup>m</sup><sub>i,j</sub> the number of shortest paths using the mulitiplex

## How much does the tube affect travel costs?

#### Metric

 Ratio of travel costs from a node *i* ∈ *V<sub>s</sub>* to all other nodes using the multiplex *vs* using the streets only

$$q_{ms}(i) = \frac{1}{N_s - 1} \sum_{j \in V_s} \frac{\tau_m(i, j)}{\tau_s(i, j)}$$



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#### Impact

- Halving  $\beta$  reduces  $\langle q_{ms} \rangle$  by about 20%
- Most journeys have a large street component that can't be removed

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## Interdependence

### Ratios of shortest paths



Inter-modal connectivity

• 
$$\lambda(i,j) = \frac{\sigma_{i,j}^m}{\sigma_{i,j}}$$

• For  $\beta = 0.8$ ,  $\langle \lambda \rangle = 0.7$ : 70% of journeys use the tube

#### Compare to scaled distances

- Scale based on  $\frac{d(i,j)}{\sqrt{A}}$
- $Q_{\lambda}(d) = \frac{1}{N(d)} \sum_{i,j \in V_{\mathcal{S}} \mid d(i,j) = d} \lambda(i,j)$
- Spatially short journeys benefit from hopping on the tube

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## Outreach

- Spatial outreach of a node *i* is the average Euclidean distance to all nodes reachable with a travel cost *τ*
- $L_{\tau}(i) = \frac{1}{N(\tau)} \sum_{j \in \{k \mid \tau_m(i,k) \leq \tau\}} d(i,j)$
- How "commutable" is a city



## **Betweenness centrality**

#### Metric

Compute σ<sup>m</sup><sub>i,j</sub>(ν) the fraction of shortest paths that go through ν ∈ V<sub>s</sub>

• 
$$bc_m(v) = \frac{1}{(N_s-1)(N_s-2)} \sum_{i,j \in V_s} \frac{\sigma_{i,j}^m(v)}{\sigma_{i,j}^m}$$

#### Impact

- Shift congestion from roads to nodal points of tubes as  $\beta$  decreases
- Tubes "decentralise" congestion to the ends of lines
- Betweenness centrality doesn't move to the centre, as happens with mesh networks

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## Shifting spatial patterns



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## Topology only

#### No congestion "agents"

- Not modelling the traffic per se
- When we suck traffic into the tube, we assume that we can
- Either roads are "sufficiently big" or traffic "sufficiently light" – neither of which is actually the case

#### A more detailed model

- Make cost dependent on centrality?
- Limit capacity of edges?

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## Wrong metrics

#### Betweenness centrality is all-to-all

Shortest paths between all pairs i and j

## A commuter model would capture which routes were more important

- Probabilistically weight the routes that people actually use
- Drive from real data, *i.e.*, TfL turnstile measurements
- Recent results show we can estimate route weights from census data on living and working population densities

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## Universality

Want to know that ideas work everywhere

• Our paper was originally titled *"Multiplex networks in metropolitan areas: universal features and local effects"* 

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#### Getting more towards universality

- Can we synthesise cities with realistic (coupled) topologies?
- Generate plausible alternative topologies to explore

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## Faking it

#### How do cities form and evolve?

- Villages coalesce over time, interconnections grow, ...
- Certain topological structures seem to be very persistent over time

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## Faking it

#### How do cities form and evolve?

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#### Study historical events, for example the Black Death

- A combination of disease, healthcare, and social structure
- Is breaking one of these features sufficient to stop an epidemic?
- Do network features make some modern epidemics worse?

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## Conclusion

- Realistic to study urban-scale networks computationally using network science
- Data is publicly available, but needs care and cleaning
- Topology-driven analysis still shows useful results
- A commuter model would be useful, and seems to be possible from observed patterns (in London at least)
- Universal results are elusive and would require significant advances in synthetic urban network generation

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## Thank you

Topology! The stratosphere of human thought! In the twenty-fourth century it might possibly be of use to someone...

Alexander Solzhenitsyn, The First Circle

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### References



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