Spatio-temporal networks: reachability, centrality, and robustness (Supplementary material)

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1. EMPIRICAL DATASETS

a. Data sources for network construction

In the interest of reproducibility, here we detail the publicly available data sources used to construct the four spatio-temporal networks used in this work.

London Metro: Comprehensive data on timetabled journeys by London Underground trains each week is provided by *Transport for London* (TfL) at the TfL Open Data Hub¹. These are provided in TransXChange² format, an XMLbased standard maintained by the UK Department for Transport.

Paris Metro: Data on timetabled Paris Metro journeys each week is provided by *Régie Autonome des Transports Parisiens* (RATP) at the RATP Open Data website³. These are provided in GTFS format⁴.

New York Metro: Data on timetabled New York City Subway journeys each week is provided in GTFS format by the *Metropolitan Transportation Authority* (MTA) at the MTA Developer Data Downloads website⁵. The overall MTA GTFS covers both New York City Subway and the Staten Island Railway, comprising 440 stations. Reducing this to New York City Subway stations active in the chosen observation window yields a network of 417 nodes.

US Flights: Flight take-off and landing times are compiled by the US Department of Transportation⁶. Flights are provided in monthly CSV (comma-separated values) files.

C. Elegans: Coordinates of neurons in the adult worm are provided by the Dynamic Connectome Lab⁷ and described in Ref. [1]. Neuronal connectivity data are available at WormAtlas⁸ and were originally compiled from the work in Ref. [2] and Ref. [3].

StudentLife: The complete StudentLife dataset has been made publicly available by the Dartmouth StudentLife Team⁹. The dataset is described in Ref. [4].

b. Geospatial embedding

LONDON METRO, US FLIGHTS, PARIS METRO, NEW YORK METRO, and STUDENTLIFE represent geospatial systems. Here we describe how we obtained the geospatial embedding of these data.

For the LONDON METRO network, we associated Underground stations in the *Transport for London* Journey Planner Timetable dataset with their corresponding geographic coordinates in the UK NaPTAN transport stop metadata¹⁰, giving nodes embedded in the WGS-84 geodetic coordinate space. We merged multiple platforms corresponding to the same station into one single entity, yielding 270 stations overall. Each station is represented by the centroid of its one-or-more platforms. We calculate geospatial distance over the surface of the Earth using the Vincenty formula.

Station and sub-station geo-coordinates are provided in the GTFS data for PARIS METRO and NEW YORK METRO. We reconcile PARIS METRO sub-stations to a single station by station name. NEW YORK METRO sub-stations are unified by GTFS parent station.

For US FLIGHTS, airport geographic coordinates are provided in the Aviation Master Coordinate Table¹¹.

Geo-location in STUDENTLIFE is obtained by the WGS-84 geographic coordinates provided by each student's Android smartphone. The Android API tracks location via a mixture of WiFi geo-location, GPS, and and cellular

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¹ Available on 05/05/2016. http://www.tfl.gov.uk/info-for/open-data-users/

² https://www.gov.uk/government/collections/transxchange

³ Available on 05/05/2016. http://www.ratp.fr/fr/ratp/r_70350/open-data/

⁴ Available on 05/05/2016. http://dataratp.opendatasoft.com/explore/dataset/offre-transport-de-la-ratp-format-gtfs/

 $^{^5}$ Available on 05/05/2016. http://web.mta.info/developers/download.html

⁶ Available on 05/05/2016. http://www.rita.dot.gov/bts

 $^{^7}$ Available on 05/05/2016. http://www.dynamic-connectome.org

⁸ http://www.wormatlas.org

⁹ Available on 05/05/2016. http://studentlife.cs.dartmouth.edu

¹⁰ UK National Public Transport Access Nodes (NaPTAN) metadata: http://data.gov.uk/dataset/naptan. Available on 05/05/2016.

¹¹ The Master Coordinate Table is among the aviation support metadata managed by the US Department of Transportation: http: //www.transtats.bts.gov. Available on 05/05/2016.



(e) C. ELEGANS (static)

FIG. 1. (Colour) Example snapshots in LONDON METRO, PARIS METRO, NEW YORK METRO, US FLIGHTS, and C. ELEGANS. STUDENTLIFE not presented due to sparsity. Edge colour represents propagation speed. Darker shades of blue indicate higher propagation speeds relative to other edges in the same network.

network geo-location. To reliably infer an individual's position from these multiple location sources we follow the same clustering approach described in Ref. [5].

c. Spatio-temporal visualisations

Fig. 1 presents example snapshots of each dataset. Each dataset's original spatial embedding is projected on to a Euclidian plane. Animations [6] of the whole duration of all five datasets can be found at http://mattjw.net/pub/stnets. MP4 video files are available via the Figshare Repository (doi:10.6084/m9.figshare.1452976).

d. Processed datasets

The LONDON METRO, US FLIGHTS, PARIS METRO, and NEW YORK METRO networks [7] generated for use in this paper are provided via the Dryad Repository (doi:10.5061/dryad.3p27r). Further description of the file format can be found at http://mattjw.net/pub/stnets. These datasets include the networks' time-varying topological and propagation information, as well as the geographic coordinates of the nodes. For privacy considerations we cannot redistribute our derivative of the STUDENTLIFE dataset, although the same spatio-temporal construction can be built from the source data. Construction of the C. ELEGANS network is described in the main article.

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- [2] B. L. Chen, D. H. Hall, and D. B. Chklovskii, "Wiring Optimization Can Relate Neuronal Structure and Function," Proc. Natl. Acad. Sci. USA 103, 4723–4728 (2006).
- [3] L. R. Varshney, B. L. Chen, E. Paniagua, D. H. Hall, and D. B. Chklovskii, "Structural Properties of the Caenorhabditis Elegans Neuronal Network," PLoS Comput. Biol. 7, E1001066 (2011).
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- [5] F. Tsapeli and M. Musolesi, "Investigating Causality in Human Behavior from Smartphone Sensor Data: A quasiexperimental Approach," EPJ Data Science 4 (2015).
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