Planes, Trains, and Afflictions

Agent-based modeling for the spread of disease through transportation networks

Frank Acquaye
Ho Lum Cheung
Dimas Muñoz-Montesinos
Elie Wanko

Faculty of Computer Science Higher School of Economics Moscow

27 JUN 2020

Table of Contents

- Motivation
- Prior Research
- ABM Framework
- Examples
- Deep Dive: New York City (NYC) Subway
- Discussion
- Conclusion
- Credits

Motivation

Why were certain areas of NYC more affected by COVID-19? How were the subways involved?

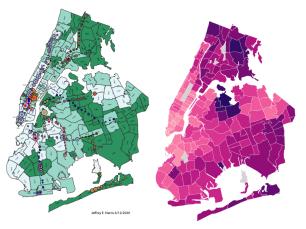
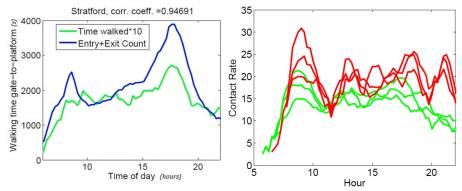


Figure: Left: Case Rate And Subway On April 12, 2020 [1] Right: NYC COVID Case Rate As Of June 23, 2020 [2]

3/26

Prior Research (London Underground)

Data: Oyster (Card), CASA (Timetable), PHE Data (Demographics)

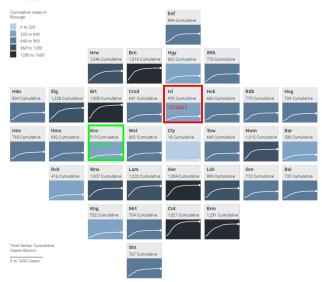


Results show a correlation between the use of the underground and ILI cases in London, specifically they show that higher numbers of ILI cases arise in those boroughs where the population spend more time in the Underground and/or incur in a higher number of contacts when travelling. [3]

COVID in London

Covid-19 Cases by London Borough (2020-03-23 to 2020-06-21)

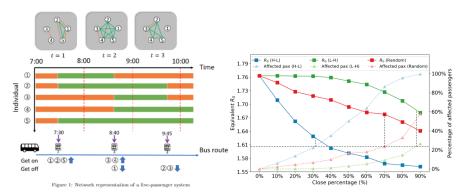
Displaying cumulative count (all historic) at: 2020-06-21



Source: https://coronavirus.data.gov.uk/ - Note: Data for most recent 5 days may be incomplete. Graphic by GLA City Intelligence | London Squared Format by After The Flood

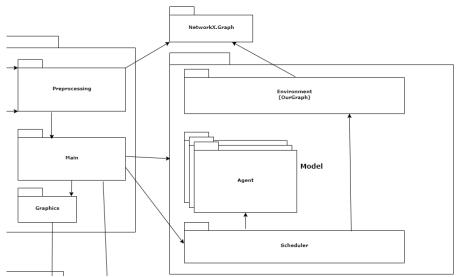


Prior Research (Singapore Buses)



The direct contact in trains is, however, difficult to obtain from smart card data because the transactions are recorded at the station level [4]

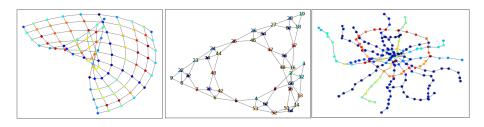
ABM Framework



Components: Python, MESA, Networkx

Base Classes: Transportation Model, SEIR Agent

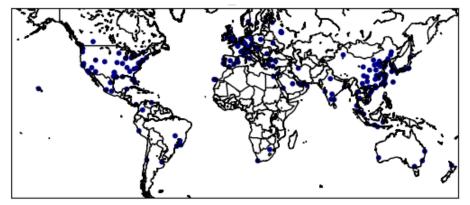
Simple Geometries



- Grid
- Sierpinski's Triangle
- Moscow

World Airline Network (Passenger Flow)

- Investigate similarities and differences with subways
- Make framework more robust



World Airline Network (Passenger Flow)

Main Idea: Detailed route data can help model spread of diseases.

- Regional / Domestic / EU Domestic / International Airports
- Regional / Domestic / EU Domestic / International Air Routes

Problem: This data not widely available. Solutions [5]:

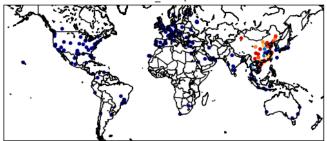
- Commercial Providers
- Open Data
- Statistical Modeling

World Airline Network (Passenger Flow)

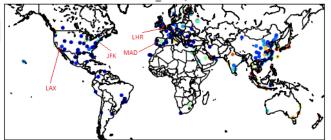
- List of Airports OpenFlights [6]
- List of Air Routes OpenFlights [6]
- Airport Passenger Flow Numbers [7]
- Top Air Routes [8]
- Other Air Routes
- Process airports, air routes
- Add all known passenger flow numbers
- Add all known airport and air route labels
- Classify remaining airports and air routes
- Interpolate missing air route data

World Airline Network Visualization (gif)

normalized_hotspot t=62



normalized_hotspot t=79



Madrid Commuter Trains (Central Hubs)

How important are central transportation hubs to disease spread? Data Source (madrid renfe gtfs)

Madrid Commuter Trains (Central Hubs)

Methods (ABM)

Results (Madrid)



NYC Demographics And COVID Timeline

NYC residents working in Manhattan primarily travel by subway. This is also true for residents of the Bronx, Brooklyn, and Queens [9]

- March 9 Mayor holds press conference and notes that there have been 16 confirmed cases. (106 cases)
- March 12 Mayor declares a local state of emergency. (687 cases)
- March 15 Schools officially close. (2,986 cases)

Borough - A geographical region. NYC has 5 boroughs.

MODZCTA - Modified Zip Code Tabulation Areas. \propto postal codes.

Subway Systems

Complex, Station, Line, Route

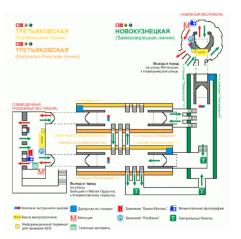


Figure: Floor Plan of Novokuznetskaya Metro Station [10]

NYC Subway Data

Stations, Map, Turnstiles (GTFS not considered)

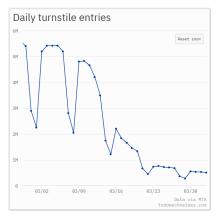


Figure: NYC Subway daily turnstile entries for March 2020 [11]

```
167,167,A32,IND,8th Av - Fulton St,W 4 St,M,A C E...
167,167,D20,IND,6th Av - Culver,W 4 St,M,B D F M...
```

NYC Subway Modeling

Algorithm 1 Simulation of Disease Spread on Subways

```
1: for i = 1; i < TIMESPAN; i + + do
       Check conditions (i, number of infected) to see if we should deploy
   COUNTERMEASURES
       for Station in SubwayModel.Environment.Nodes do
 3:
          Calculate 'Local Exposure' from infected and commute time.
 4:
          Calculate 'Route Exposure' from infected on the same route.
 5:
          Calculate 'General Exposure' due to city-wide infected.
 6:
          Update 'Exposure' at station based on above conditions
 7:
       end for
 9:
       for Agent in SubwayNetwork.Agents do
          Get 'Exposure' At Location
10:
11:
          Get City-wide COUNTERMEASURES
          Get Percentage of commuters
12:
          Calculate SEIR beta and gamma based on conditions
13:
          Update SEIR numbers
14:
       end for
15:
```

Subway Agent - SEIR Agent with additional exposure based on location.

Subway Graph - Nodes

Stations, Edges

Lines, Complex, Route Lookup

€

€

E

16 end for

NYC Model Parameters, Hyper-parameters

	Contact Rate (β)	Latent Rate (α)	Removal Rate (γ)	Start Trigger
Default	1.75	0.2	0.5	t = 0
Isolation Modifier	0.25	1	2	I >5000
Recommendation Modifier	0.67	1	1.5	I >500
Awareness Modifier	1 to 0.25	1	1	I >500

Other Parameters: Defiance, Global Exposure, Commuter Ratio By Station, Awareness Increase Rate, Borough Commuting Modifier...

Compartmental Modeling Results (NYC)

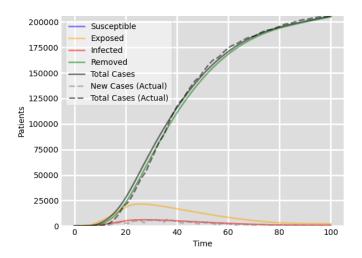
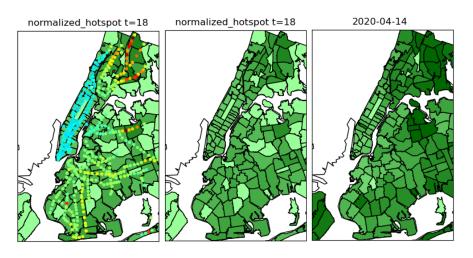


Figure: SEIR Fitting to NYC Case Total. *MAPE* ($t \ge 30$): 0.0145

27 JUN 2020

Results by MODZCTA (NYC)



Scale(0 - 0.11 infections/person, 0 - 0.11 infections/person, 0 - 0.20 cumulative cases/person)

MAPE ($t \ge 32$, Stations $\ge 1 = 0.311$)

New York City Timelapse



Discussion

- More granular passenger flow and population data.
- New York City is not a closed system.
- Control for population density, income, and other factors.
- Simplify away the feature bloat

Conclusions

All models are wrong, but some are useful -George Box

- Determining passenger flow is a difficult problem for all transportation networks at all granularities.
- Rush hour and hubs contribute strongly to disease propagation
- Commute time is correlated to prevalence of ILIs
- Subway routes are correlated to prevalence of ILIs

Acknowledgements (and Questions)

Frank Acquaye - WAN, Passenger Flow
Ho Lum Cheung - NYC, Organization, Research, Testing
Dimas Muñoz-Montesinos - Madrid, Hotspots
Elie Wanko - Theory, Consulting
Project on Github

References

Jeffrey E. Harris.

The subways seeded the massive coronavirus epidemic in new york city.

http://web.mit.edu/jeffrev/harris/HarrisJE WP2 COVID19 NYC 24-Apr-2020.pdf

Covid-19: Data.

https://www1.nyc.gov/site/doh/covid/covid-19-data.page.

Lara Goscé and Anders Johansson.

Analysing the link between public transport use and airborne transmission: mobility and contagion in the london underground. Environmental Health, 17(1), 2018.

Yu Shen Clarence Tam Daging Li Yafeng Yin Jinhua Zhao Baichuan Mo, Kairui Feng.

Modeling epidemic spreading through public transit using time-varying encounter network.

Preprint, ResearchGate.

https://www.researchgate.net/publication/340541290_Modeling_Epidemic_Spreading_through_Public_Transit_using_
Time-Varving Encounter Network

Liang Mao, Xiao Wu, Zhuojie Huang, and Andrew J. Tatem.

Modeling monthly flows of global air travel passengers: An open-access data resource, Sep 2015.

https://openflights.org/.

https://aci.aero/news/2019/03/13/preliminary-world-airport-traffic-rankings-released/.

https://www.routesonline.com/news/29/breaking-news/286313/busiest-routes-in-the-world-the-top-100/.

The ins and outs of nyc commuting.

 $https://www1.nyc.gov/assets/\overline{planning/download/pdf/planning-level/housing-economy/nyc-ins-and-out-of-commuting.pdf.}\\$

Plan of metro station novokuznetskaya.

http://www.karta-metro.ru/stations/96/428/.

New york city subway usage.

https://toddwschneider.com/dashboards/nyc-subway-turnstiles/.