Cross City Urban Transit Networks Analysis

Group 16 Mid Term up<u>date</u>

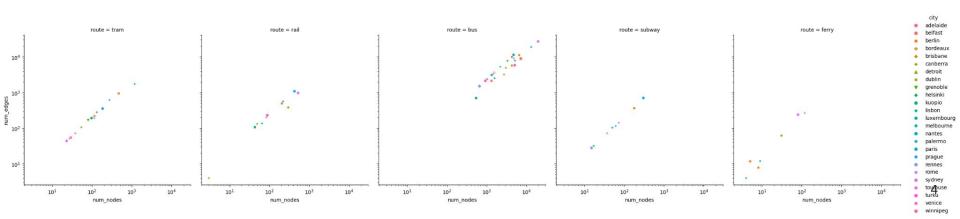
Aims

- How do the topological properties of public transport networks differ across different cities?
- Which public transport stops are the most important in each city based on their centrality measures?
- How do public transport networks compare across different cities (similarities and differences)?

Exploration

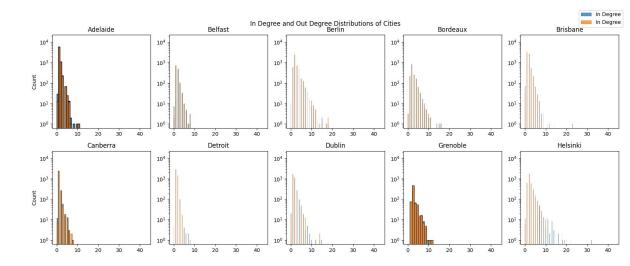
Preprocessing

- Two Main files for each city
 - Network_nodes.csv: For the nodes
 - Network_combined.csv: For the edges
- Merged nodes with same name and removed self loops between the nodes



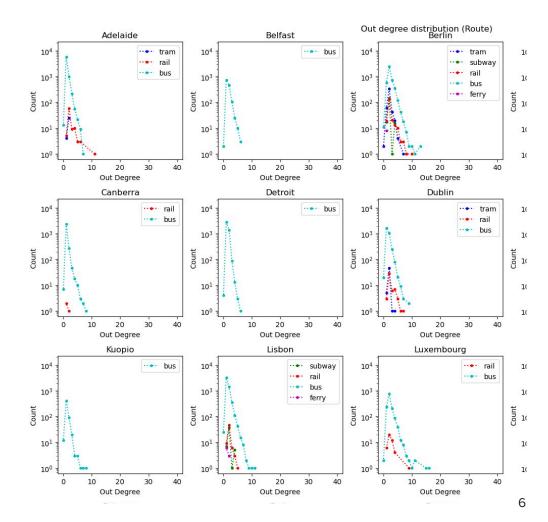
Tasks - Degree Distribution

- In-degree follows a very similar pattern to the out-degree distribution of graph nodes
- The average degree has a median of 2.03 and mean of 2.0 telling us that each station is connected with two other stations.



Tasks - Mode wise degree distribution

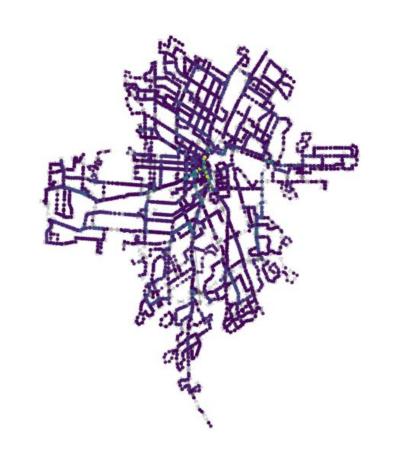
- The existence of a large number of nodes with low degrees and a smaller number of nodes with higher degrees is observed.
- This implies that the networks cannot be modelled by a random network model



Full network of Winnipeg with betweenness centrality of nodes

Task - Centrality Measures

- Average clustering coefficient
- Betweenness centrality
- Central/Transit stations have high betweenness centrality



- 0.14

0.12

- 0.10

- 0.08

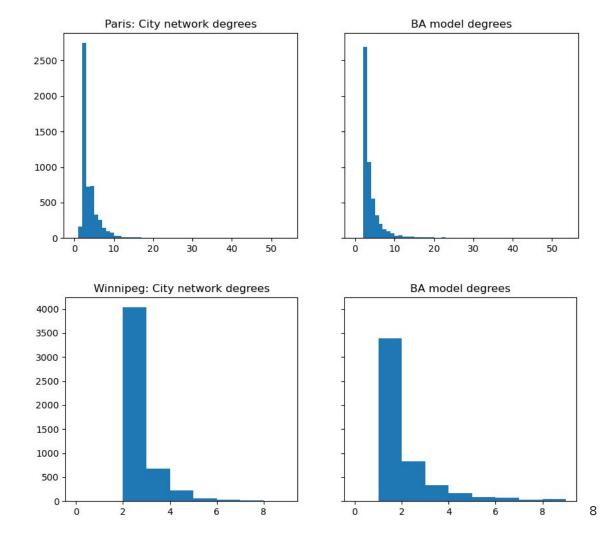
0.06

0.04

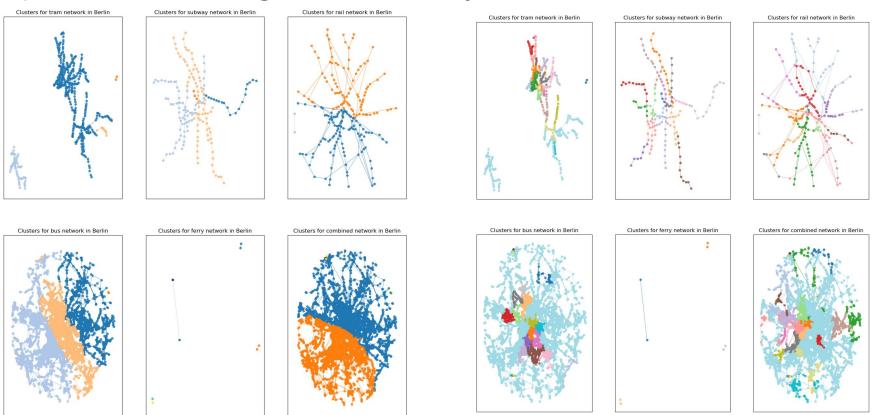
- 0.02

Task - Modelling networks

- Looking at the degree distribution above, we concluded that it cannot be modelled by a random network model.
- Comparing it with the
 Barabasi Albert generation
 model with an appropriate
 choice of q (from the degree
 distribution of network
 nodes).



Spectral clustering & Community detection

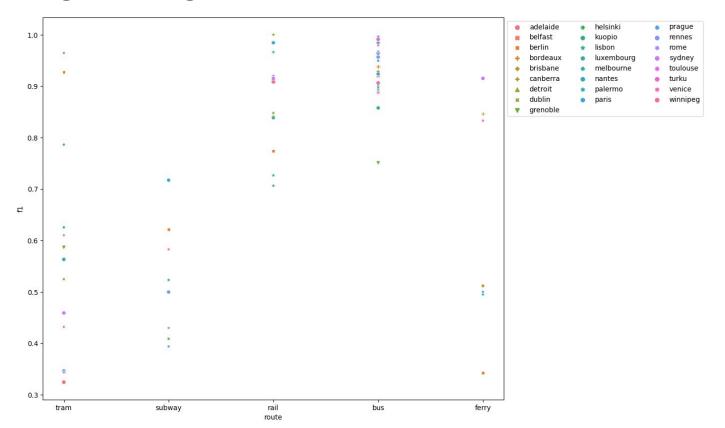


Exploitation

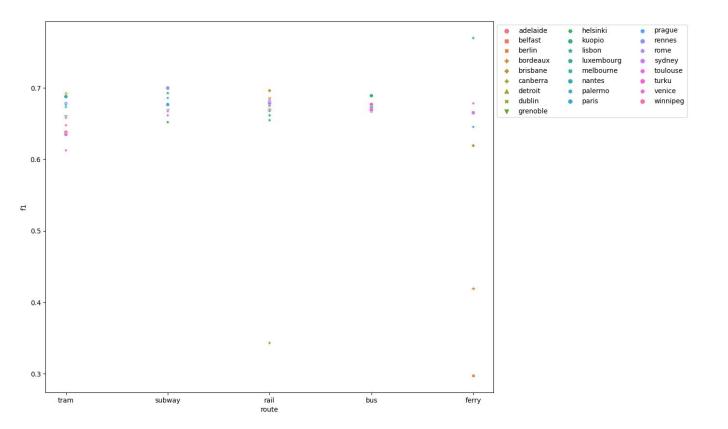
Task - Edge Prediction

- Given two nodes, predict if an edge exists from node 1 to node 2 (directed).
- Models
 - Logistic Regression
 - Node 2 Vec.
 - Deep Learning (may be)
- We do transport mode wise analysis
- Only finished LR with 50:50 train/test split.
 - Hand crafted features
 - In degrees centrality
 - Out degree centrality
 - Betweenness centrality
 - Katz centrality
 - Number of Common predecessors
 - Number of common successors

Logistic Regression - Results



Node2Vec (3 features) - Results



Next steps & Issues

Task - Edge Prediction

- Try with Node2Vec models
- Edge prediction for whole network
- Use Deep Learning models for edge presence and category prediction for full network

Issues

- Storing the edge-level features is very expensive in terms of memory required.
 - Is it possible to access any computing clusters?
 - This will also be needed for using GNNs.
- Should we look at alternative exploitation tasks/methods to mitigate memory problem?

Thank you