

An approach to determine public facilities placement a case study based on Canberra bus network

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Catalogue



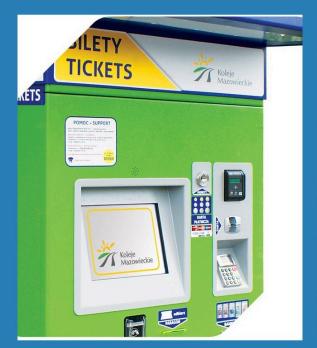
Motivation







What are public facilities?



Ticket vending machines



Electric vehicle charge stations



Sharing bike docks

ng_machines/

Tap on. Tap off.

Travel smart with MyWay.





How to distribute ticket vending machines in bus network?

We only have 5 ticket vending machines in Canberra. Not enough!

If the Canberra bus transportation company wants to set more auto charge machines to meet users's requirement. What's the placement plan to achieve most efficient distribution?

Also, we have to consider the robustness.

Nodes: machines randomly broke down;

machines in high-degree nodes are more likely not available (i.e. city center)

Edges: random edges failures (bus line broke down or road closed by accidents);

high-betweenness edge fails (i.e. traffic jam)

Dataset overview

Canberra	Canberra_merged
2520	1523
2908	1976
2.30794	2.59488
0.0118474	0.0760809
0.0210773	0.103448
10	16
1	1
51	44
26	22
0.0848577	0.118149
0.184854	0.582854
	2520 2908 2.30794 0.0118474 0.0210773 10 1 51 26 0.0848577

• Data pre-processing:

- merging nodes: decreasing the size of our network and make our approach more efficient.
- opposite stops
- close stops

• Interesting properties:

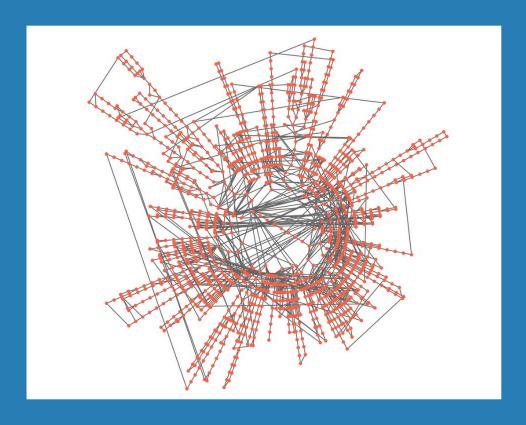
average degree

diameter

betweenness

• • •

Dataset overview



• Network topology:

red nodes represent merged bus stations edges represent trips

Distribution methods

Random

Randomly arrange a number of ticket vending machines in this network

High degree first

Firstly arrange ticket vending machines in the node with a high degree

Optimized solution

The best placement plan, but it's a NP-hard problem

Greedy algorithm

Every time arrange a ticket vending machine, minimizing the average shortest distance to the nearest machine

Our method: community + locally greedy

Based on communities detection method (greedy, Lovain etc.), and use greedy method within each community

Algorithms

How to measure the average shortest distance to the nearest machine?

- 1. The unit of distance is the stops.
- 2. We use breath first search to find the shortest distance between the node and the nearest machine. Then we compute the average shortest distance.

Greedy Algorithm

Assume we have n nodes(stops), m edges and we want to set b machines.

- 1. We assign a machine to a stop which can minimize the current average shortest distance most.
- 2. Repeat the step 1 until all machines are assigned.

Our Algorithm - Locally greedy by using community detection

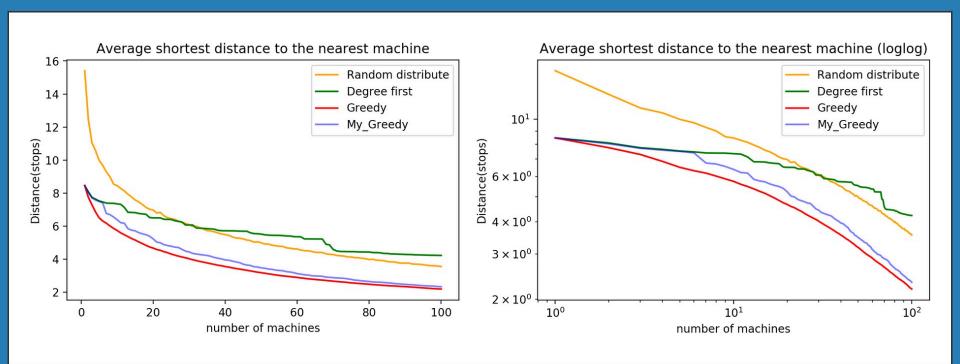
- 1. We use a community detection method (Greedy/Louvian, etc.) to find the community partition of the network, assuming the best partition has q number of community. The average number of nodes of each community is n/q and the average number of edges are m/n.
- 2. Assume the number of machines is greater than the number of communities.
- 2.1. First, we assign q machines to each community.
- 2.2. In each community, we trade it as a sub-graph, we assign a machine to a stop which is in the community and it can minimize the current average shortest distance within the community most.
- 2.3. Once step 2.1 finished, then we choose a community has the least dense of machines (b c/ size(community)), then we apply step 2.2.
- 2.4. Repeat step 2.3 until all machines are assigned.
- 3. If the number of machines is less than the number of communities, we just assign machines to those communities have the largest size.

Compare of algorithms

n, nodes; m, edges; b, number of machines; q, number of communities

Distribution methods	Availability	Computational time complexity	Result	
Optimised solution	No	N/A	unachievable	
Random	Yes	O(1)	bad	
High degree first	Yes	O(nlgn)	bad	
Greedy algorithm	Yes	$O(bn^3 + bmn^2) \rightarrow O(b^2n^2)$	good	
Our method: community + locally greedy	Yes	$O(\frac{bn^3 + bmn^2}{q^3}) \to O(\frac{b^2n^2}{q^3})$	good	

Result



Robustness analysis

machine random failure

- the machine randomly broke down
- under maintenance

machines in high degree failure (degree first)

- a long queue
- tend to have problem due to high using frequency

edge random failure

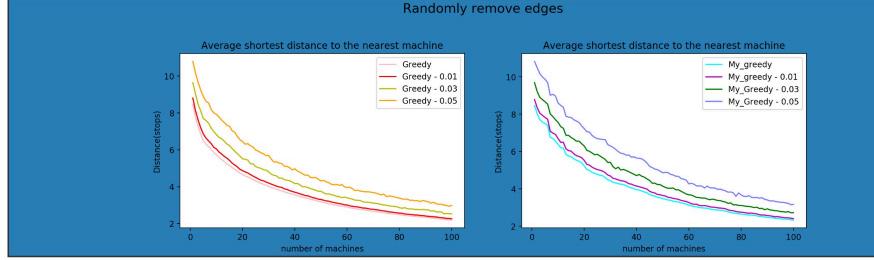
- bad weather
- road work

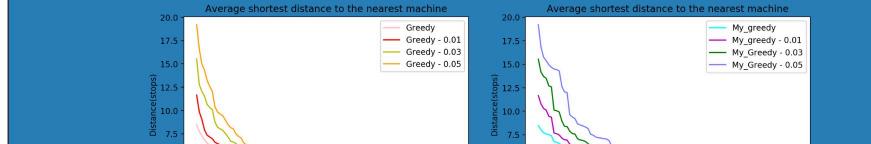
high betweenness edge failure

- traffic congestion at rush hours
- accident

Fail machines with probaility Greedy - random 3.8 -Greedy - degree first My_greedy - random My_greedy - degree first Distance(stops) 3.4 -3.0 0.05 0.15 0.10 0.20 0.25 0.00 failure probaility

Failure probability	0.00	0.05	0.10	0.15	0.20	0.25
Greedy-random	2.90	3.03	3.18	3.34	3.50	3.68
Greedy-DF	2.90	3.04	3.21	3.39	3.60	3.82
My_greedy-random	3.13	3.26	3.41	3.55	3.72	3.87
My_greedy-DF	3.13	3.26	3.39	3.54	3.71	3.90





5.0

2.5

number of machines

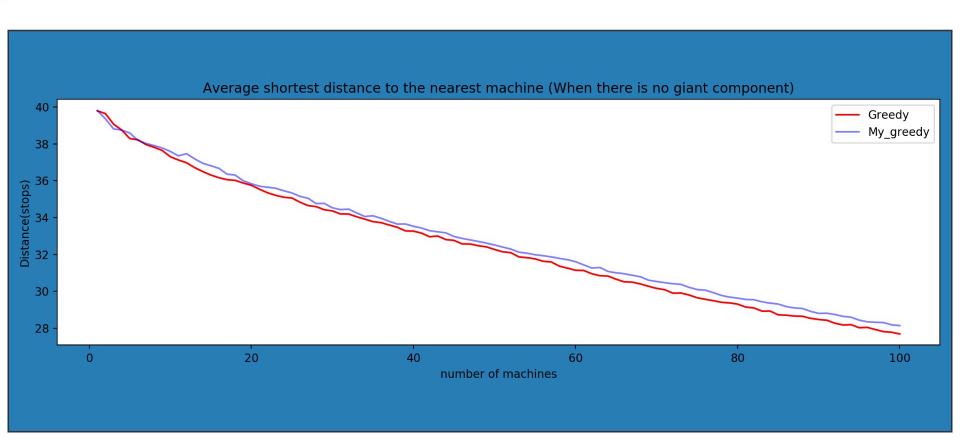
Remove edges with high betweenness

5.0

2.5

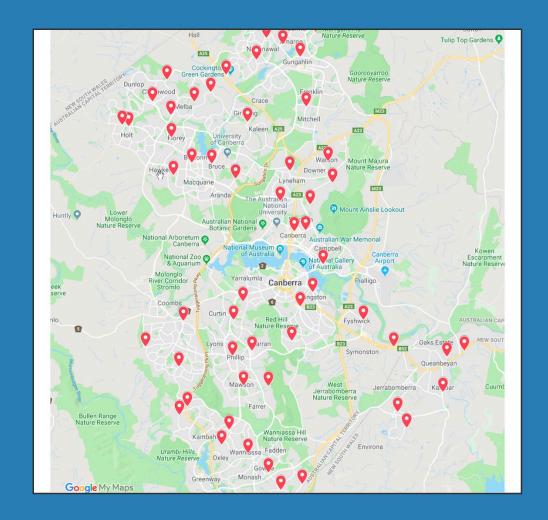
number of machines

Remove edges to no giant component...



Conclusions

- Our results are basically same as the original greedy algorithm
- But our method is much more fast than original greedy algorithm
- Our distribution has a high robustness



Applications

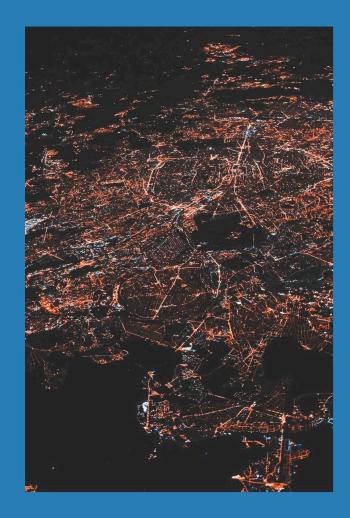
• Other city bus network..

ticket vending machines

• City planning.. self-help parcel lockers public toilets

New sharing-economy company..

sharing bike docks sharing power bank boxes



Limitation and future work:

- 1. The method of merging stops can be improved.
- 2. More random simulations are expected even though we already tested hundreds times.
- 3. Population density or people flow data can be added in future.

Thanks for listening! Vote us~