



AMMA Consulting

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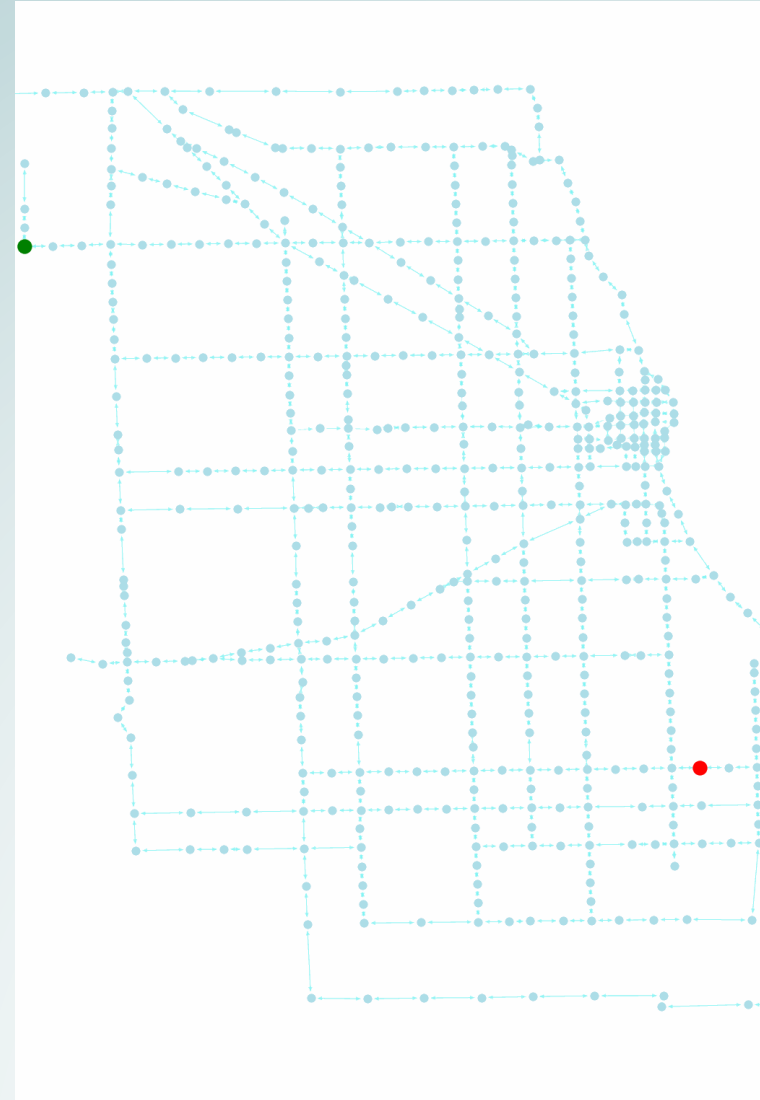
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18.07.2024

Topics

- 1. Similarities**
- 2. Algorithms: Dijkstra / A-Star**
- 3. Computational Results and Analysis**
- 4. Conclusion**
- 5. Difficulties / Complications**



Similarities

edge weight
*
(1-percentage)

$$\text{edge weight} * (1 - \text{percentage})$$

First approach

- percentage: frequency of found edges in all historical routes

$$\text{edge weight} * (1 - \text{percentage})$$

First approach

- percentage: frequency of found edges in all historical routes

Second approach

- percentage: frequency of found edges in historical routes taken in one hour, i. e. from 00_00_00 to 00_59_59

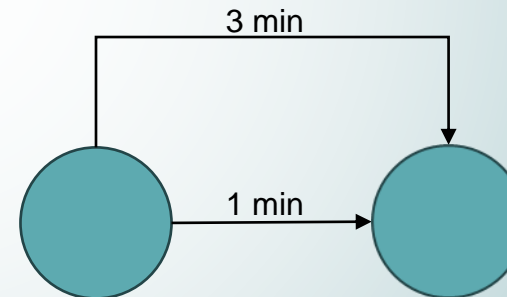
edge weight
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First approach

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Second approach

- percentage: frequency of found edges in historical routes taken in one hour, i. e. from 00_00_00 to 00_59_59



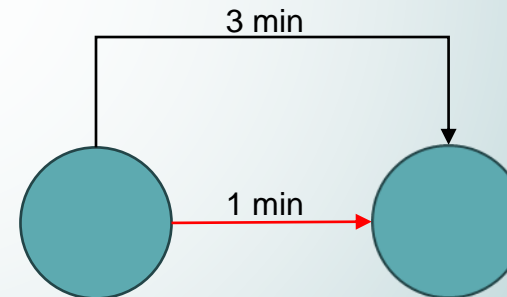
edge weight
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(1-percentage)

First approach

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edge weight * (1-percentage)

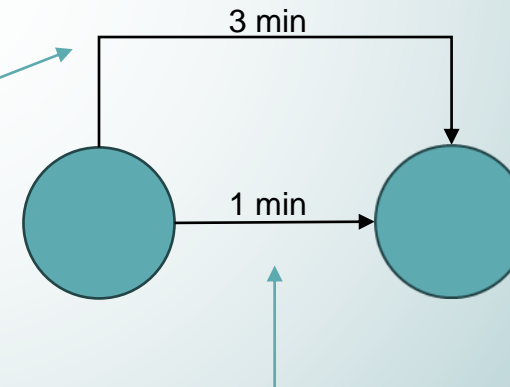
First approach

- percentage: frequency of found edges in all historical routes

percentage: 80%
→ new weight: $3 * (1 - 0.8) = 0.6$

Second approach

- percentage: frequency of found edges in historical routes taken in one hour, i. e. from 00_00_00 to 00_59_59



percentage: 10%
→ new weight: $1 * (1 - 0.1) = 0.9$

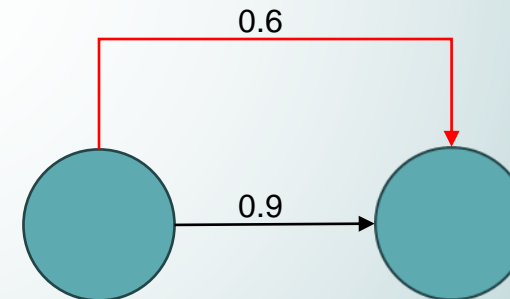
edge weight
*
(1-percentage)

First approach

- percentage: frequency of found edges in all historical routes

Second approach

- percentage: frequency of found edges in historical routes taken in one hour, i. e. from 00_00_00 to 00_59_59



→ new shortest, more optimal path

Algorithms

Dijkstra / A-Star

Dijkstra

Dijkstra

A*

Dijkstra

A*



edge weight: duration * similarity

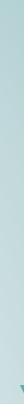
Dijkstra



A*



edge weight: duration * similarity



Heuristic: euclidean distance from
node 94 to node $Y \quad \forall Y \in \{0, \dots, 537\}$


```
def dijkstra_main(timestamp, per):  
    city_graph2: WeightedGraph[str] = WeightedGraph([str(i) for i in range(538)])  
  
    graph = pc.parse_csv()  
    for i in range(1, len(graph)):  
        fro = graph[i].get('from')  
        to = graph[i].get('to')  
        percentage = 1 - (per[i - 1])  
        city_graph2.add_edge_by_vertices(str(fro), str(to),  
                                         dur.get_edges_predicted_duration_new(timestamp)[i-1] * percentage)
```

Dijkstra

Implementation Similarity

```
lst = []  
for j in range(counter, len(p)-1):  
    if fro == p[j+1].get("from"):  
        percentage = 1 - per[j]  
        lst.append((str(p[j+1].get("to")),  
                    dur.get_edges_predicted_duration_new(timestamp)[j] * percentage))  
    if fro not in liste:  
        liste.append(fro)  
counter += 1  
Graph_nodes[str(fro)] = lst
```

A*

Implementation Similarity

Dijkstra

A*

edge weight: duration * similarity

Heuristic: euclidean distance from
node 94 to node Y $\forall Y \in \{0, \dots, 537\}$

Final approach for
optimal route with
this similarity

→ A*

Dijkstra

A*

edge weight: duration * similarity

Heuristic: euclidean distance from
node 94 to node Y $\forall Y \in \{0, \dots, 537\}$

Final approach for
optimal route with
this similarity

→ A*

Why: is a better Dijkstra algorithm and returns the best
optimal route out of all algorithms

Validating the similarity

Testing with the route-all.csv, validating with May instances from route-all-missing-last-day.csv

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Testing with the route-all.csv, validating with May instances from route-all-missing-last-day.csv

First approach

The percentage of the routes, that are $\geq 75\%$ similar to our optimal computed route, should lay between **5** and **10 %**

Validating the similarity

Testing with the route-all.csv, validating with May instances from route-all-missing-last-day.csv

First approach

The percentage of the routes, that are $\geq 75\%$ similar to our optimal computed route, should lay between **5** and **10 %**

Second approach

The percentage of the routes, that are $\geq 75\%$ similar to our optimal computed route, should lay between **10** and **30 %**

Validating the similarity

First approach

The percentage of the routes, that are $\geq 75\%$ similar to our optimal computed route, should lay between **5** and **30 %**

Second approach

The percentage of the routes, that are $\geq 75\%$ similar to our optimal computed route, should lay between **5** and **30 %**

Final approach:

Computational Results and Analysis

Results

Computational



Mine



Mengyuan



Maharshi



Final result

Mine

- A* optimal route
- Optimal in balance with similarity

Mengyuan

- 3 alternative routes to the computed optimal route
- As an addition to the optimal route
- Based on another similarity

Maharshi

- Dijkstra optimal route
 - One alternative route based on dtw-similarity
 - Historical route
- 3 routes the client can choose from

Final Result

- A* / Dijkstra optimal route
- 4 alternativ routes
- 1 historical routes

Analysis

Route

- Calculated route often also a historical route
- If not, then it is very similar to it

→ our route has with a high
kk possibility a good balance
kk between optimal and similar

Similarity

- although the optimal route with similarities is longer than the shortest route, this is completely fine
- Balance between real life and theoretical world
- Shortest path algorithm without similarity is not applicable in real life

Conclusion

What you can expect from AMMA consulting group:

- Optimal routes everytime
- Additionally alternativ routes, similar to the optimal one
- Precise predictions

Difficulties / Complications

- Neural Network:

- predicted speeds to high
- balance
- fine adjustment
- accuracy

- Similarities:

- adjustments to similarity approach
- implement past similarities in the algorithm
- right approach?
- How to improve:
 - over time, add more historical data to have a more accurate algorithm
 - more testing: is the similarity approach still working?

- Implementing and Coding

- bad runtime: A* and Dijkstra compute 30 minutes for one route
- How to improve?
 - implement more efficiently, hire skilled developer



Thank You!

www.amma-consulting.com