Towards m-Traveling Salesmen Problem in Time-dependent Road Networks



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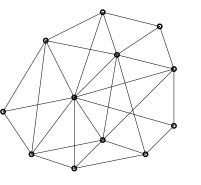
Introduction

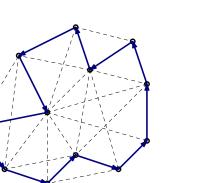
• *Traveling Salesman Problem* (TSP): Given a set of cities and their pairwise distances, find the shortest possible route for a salesman such that each city is visited exactly once and finally returns to origin.

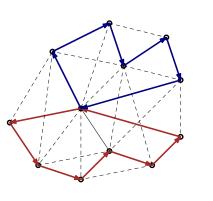
• *m-Traveling Salesman Problem* (m-TSP) is an extension of TSP where multiple salesmen are involved.

• Each node is visited by at least one salesman;

 The total traveling time of each salesman is ≤ h, where h is a predefined parameter







Related Work

- Time-dependent shortest path computation
 - First solved by Dreyfus (JOR'69) with a variant of Dijkstra's algorithm.
 - First attempt on time-dependent k-NN query processing: Demiryurek et al, (DEXA'10).

Traveling salesman problem

- Fist defined in 1800s by mathematicians W. R. Hamilton and Thomas Kirkman.
- First approximation algorithm proposed by Christofides, H. (1976) where the approximation ratio is upper bounded my 1.5.

Algorithm

• Due to the NP-hardness of time-dependent m-TSP, we focus on an approximation solution using greedy approach.

| Road network | TSP |
|--------------|-----|
| | |

m-TSP

Time-dependent m-TSP

- Time-dependent m-TSP is a variant of m-TSP in the sense that optimal routes are computed in the context of a *time-dependent road network* instead of a static road network.
 - A time-dependent road network is a road network where the traveling time of the road segments varies with time.



```
Algorithm: add_salesman(S, C, c_0)
Algorithm: m_TSP(G, C, c_0)
                                                   Input: S a set of salesmen. C a set of delivery
Input: Road network G (V, E).
                                                   centers. Origin c_0.
Delivery centers C \subseteq V. Origin c<sub>0</sub>
                                                  Output: null.
 ∈ C.
                                                  1: s \leftarrow a salesman and his route r \leftarrow \{c_0\};
Output: S, a set of salesmen with
                                                   2: c \leftarrow the closest delivery center to c_0 in C;
their routes.
                                                   3: while c can be added to r without violating
1: S \leftarrow \{\};
                                                   time constraint
                                                          r \leftarrow r \cup the path from c0 to c;
                                                   4:
2: while C is not empty;
                                                          C \leftarrow C - \{c\};
                                                   5:
        add_salesman(S, C, c_0);
3:
                                                          c ← the closest delivery center to c in C;
                                                  6:
4: return S;
                                                  7: S ← S \cup {s};
```

Experiments

- Dataset
 - Los Angeles road network with $\sim 1.5 \times 10^5$ nodes and 258 delivery centers.
 - Time-dependent edge travel-times are generated based two-years of historical data collected from 6300 traffic sensors. The sampling rate of the data is 1 reading/sensor/min.

•Evaluation

- How much total transportation time is saved (hours) by time-dependent shortest path planning compared with traditional shortest path planning.
- Average time for computing delivery routes: ~150 seconds.
- Experiments show that in average we need ~1-2 less drivers to cover all LA delivery centers by using time-dependent route planning instead of static route planning.

Challenges

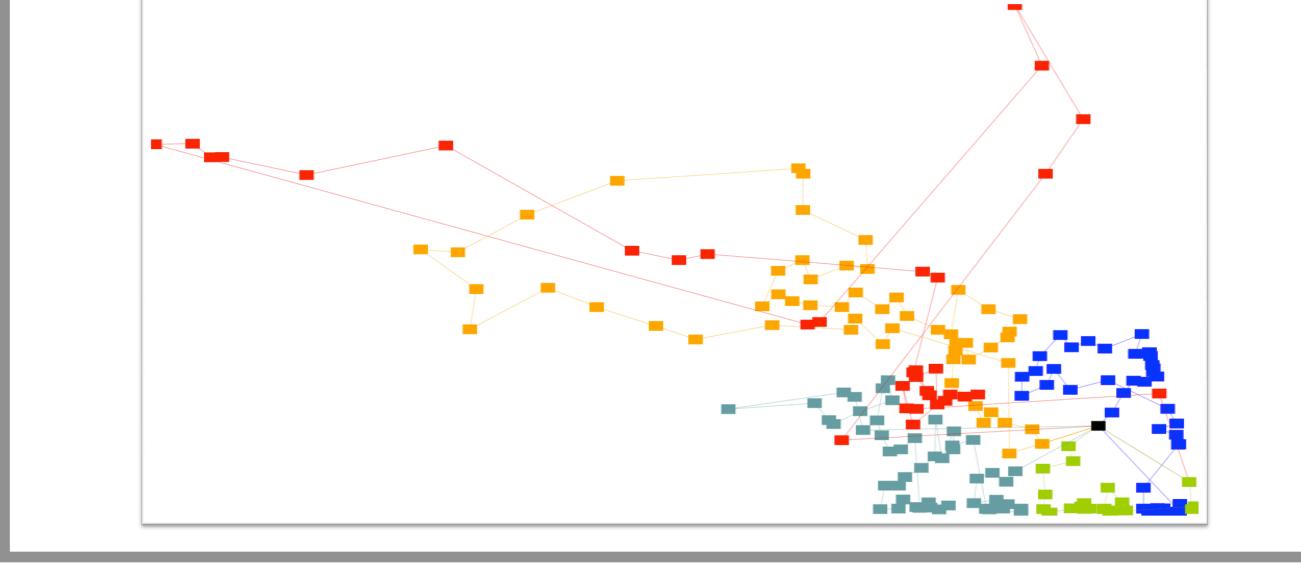
 m-TSP is an NP-hard problem, which derives directly from the NP-hardness of TSP.

• Difficult to find a solution to an NP-hard problem that is both optimal and efficient.

• Challenges also arise from time-dependency:

- Unpredictability of future traffic condition. We solve this problem was solved by analyzing historical traffic information and using this knowledge to predict future traffic data.
- Expanded search space. Since the time dimension is added to shortest path computing, we will have a much larger search space. It was proved that there might be an exponential number of shortest paths when traffic condition changes with time (Foschini et al, SODA'11).
- Existence of multiple shortest paths. The shortest path also depends on the departure time from the source.

| Problem size | 50 | 60 | 70 |
|-----------------|----------------|---------------------------|----------------|
| Easy problem | 0.0025 seconds | 0.0036 seconds | 0.0049 seconds |
| NP-hard problem | 3855 centuries | 2×10^8 centuries | ∞ |



Conclusion

• Time-dependent m-TSP is a variant of m-TSP, where the traveling time depends on not only the distance but also the traffic condition of the roads that varies with time.

- Time-dependent m-TSP is more general and realistic approach than traditional m-TSP; time-dependent route planning saves ~25.9 hours' total delivery time compared with static route planning that does not consider time dependency.
- m-TSP is NP-hard. It takes a intolerably long time to find the optimal routes.
 Therefore, we focus on approximate solutions using a greedy approach.



