

# Heuristics for Master Thesis

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## 1 Introduction

We assume our graph is a complete graph. If it is not a complete graph, we induce a complete graph by computing the distances between all nodes. We assume a graph is located on a plane, such that given 3 nodes  $u$ ,  $v$ , and  $w$ :  $d(u, w) \leq d(u, v) + d(v, w)$ . This implies that there are no shortcuts to be taken in the complete graph, and the shortest path is always the direct one.

The original graph may be weighted both on the nodes and on the edges. The nodeWeights indicate the number of POI's located in the node, while the edgeWeights indicate the distance between the given nodes. We induce a non-node-weighted graph from the complete graph by deleting all nodes with 0 weight and for all nodes with weight  $n > 1$  we create a set of  $n$  nodes with edgeWeights set to 0 between them, and connect them to all other nodes. Thus any connected graph can be induced to a complete non-node-weighted graph.

### 1.1 Limit sum of length of trips, all trips are length $k$

Given a complete weighted graph  $G$ , a start node  $s$ , a trip-length  $k$  and bound  $X$ ; Is there a set of trips  $T$  which visits all of  $V(G)$  such that each trip (except the last trip) visits at most  $k$  nodes (excluding  $s$ ) and the total length all trips is at most  $X$ ?

*Input :  $G, w(E(G)), start s, trip - length k, constraint X,$   
and  $\forall u, v, w \in V(G), w(u, w) \leq w(u, v) + w(v, w)$*

*Problem :  $\exists$  set of trips  $T : t_1, t_2, \dots, t_n$  where a trip  $t_i \subseteq V(G)$   
with an ordering of nodes  $v_1, v_2, \dots, v_m$  s.t. the following holds :* (1)

1.  $\forall v \in V(G), v \in t_i \wedge v \notin T \setminus t_i$
2.  $\forall t_i \in T v_1 = v_m = s$
3.  $m = k + 2, |t_n| \leq k + |v_i = s|$
4.  $\sum_{i=1}^n \sum_{p=1, q=2}^m w(E(v_p, v_q)) v_p, v_q \in t_i \leq X$

This problem, we will call the  $k$ -Round-based Traveling Salesman Problem or  $kRbTSP$ . A trip should minimize the total distance it travels. It is trivial to

observe that this is an NP-Hard formulation as a specialized instance of this problem is the Traveling Salesman Problem. The problem becomes TSP if we set  $k = |V(G)|$ . Note that we require each trip to visit the maximum number of nodes capable as long as there are at least  $k$  nodes left unvisited. This is a potential requirement to look at, as we may compare a solution that does not prohibit smaller trips and see if there is a possibility of getting a smaller solution:

## 1.2 Limit sum of length of trips, all trips are length at most $k$

Given a complete weighted graph  $G$ , a start node  $s$ , a trip-length  $k$  and bound  $X$ ; Is there a set of trips  $T$  which visits all of  $V(G)$  such that the number of nodes visited excluding  $s$  is at most  $k$  and the total length all trips is at most  $X$ ?

*Input :  $G, w(E(G)), \text{start } s, \text{trip-length } k, \text{constraint } X,$   
and  $\forall u, v, w \in V(G), w(u, w) \leq w(u, v) + w(v, w)$*

*Problem :  $\exists$  set of trips  $T : t_1, t_2, \dots, t_n$  where a trip  $t_i \subseteq V(G)$   
with an ordering of nodes  $v_1, v_2, \dots, v_m$  s.t. the following holds :* (2)

1.  $\forall v \in V(G), v \in t_i \wedge v \notin T \setminus t_i$
2.  $\forall t_i \in T \ v_1 = v_m = s$
3.  $m \leq k + 2$
4.  $\sum_{i=1}^n \sum_{p=1, q=2}^m w(E(v_p, v_q)) v_p, v_q \in t_i \leq X$

This problem we call the Round-based Traveling Salesman Problem, or RbTSP, and is essentially the same as the previous with the only difference being the length of each trip no longer requires a total visit of  $k$  nodes if possible. Important to note, is the constraint of distance between nodes. By not having any "shortcuts", it is not yet clear if RBTSP will yield a better solution than KRBtSP.

## 2 Conclusion

The problems can be solved using various techniques, though it is uncertain if they yield a minimum solution. First, in order to obtain a complete graph we must obtain a distance matrix between all members of  $G$  using a path-finding algorithm. From this matrix we can build a complete graph.

KRBtSP has a simple greedy solution: Simply visit the closest non-visited node at each step, and then return to base once  $k$  nodes are visited. This is an approximation algorithm, in which the approximation factor is currently unknown.

RBTSP may require a more sophisticated solution, which we call a clustering solution: First use the distance matrix in a clustering algorithm to get a

grouping where the size of each group is at most  $k$ . Each cluster is grouped such that the average distance between members of the group, including the base, is minimized. This step can be approximated in polynomial time, but is NP-hard to optimize.

In the clustering step the grouping constraint takes the double of the distance to base, as it must be visited twice in a trip. This is a key element that may allow the clustering solution to beat the traditional greedy solution, as it does not take the distance to base into account. However, it is not yet determined if it makes a difference in the computation.

Finally we could use a traditional TSP solving technique for each group. Although the final step is exponentially hard to compute, so long as  $k$  is of a relative small size ( $k < 30$ ) it should not be too resource intensive. As  $k$  is a constant it does not grow as the problem size grows so it is not considered an exponential solution for that matter.

Hopefully the clustering solution yields a smaller approximation factor than the proposed greedy solution while staying polynomial in time.