

# The Team Orienteering Problem with Capacity Constraint and Time Window

Zhenping Li<sup>1,\*</sup>

Xianman Hu<sup>2</sup>

<sup>1</sup>School of Information, Beijing Wuzi University, Beijing 101149, China

<sup>2</sup>Department of Postgraduate, Beijing Wuzi University, Beijing 101149, China

**Abstract** The Team Orienteering Problem (TOP) is defined as to find a set of vehicle walks such that the total collected reward received from visiting a subset of customers is maximized and the length of each vehicle walk is restricted by a pre-specified limit. The team orienteering problem with capacity constraint and time window (TOPCT) is an extension of the TOP where each customer has a demand and a time window, each customer must be visited in the given time window and the capacity of each vehicle is limited. In this paper, we first formulate the TOPCT problem into an integer linear programming based on network flow theory and solve it to obtain the exact optimal solution. Then we give computational results which demonstrate the efficiency of the model and algorithm.

**Key words:** Team Orienteering Problem with Capacity Constraint and Time Window; Integer Linear Programming; Computational Result.

## 1 Introduction

The orienteering problem (OP) is a well established problem in combinatorial optimization which is introduced in [4]. In the orienteering problem, we are given an edge-weighted graph  $G(V, E)$ , where  $V = \{v_1, v_2, \dots, v_n\}$ , each node  $v_i \in V$  has an associated non-negative score  $w_i$ , each edge  $e_{ij} \in E$  has a weight  $d_{ij}$  representing the length or time taken to travel from node  $v_i$  to node  $v_j$ . Given a starting node  $s \in V$  and an ending node  $t \in V$ , a length or time limit  $D_{lim}$ . The goal is to find a walk beginning at  $s$  and ending at  $t$  of total length at most  $D_{lim}$  that maximizes the total score of distinct vertices visited by the walk.

In the past decades, researchers have proposed a large number of heuristics algorithms for the OP. Tsiligrides [11] proposed a stochastic algorithm for the OP. Wang et al [7] applied artificial neural networks to the OP and obtained high-quality results. Chao et al [1] applied deterministic annealing to the OP and also obtained high-quality results. Gendreau et al. [10] applied tabu search to the OP and obtained near-optimal solutions to instances with up to 300 nodes.

The Team Orienteering Problem (TOP) is a generalization of the OP in which a set of vehicle walks are constructed such that the total collected reward received from visiting

---

\*corresponding author: lizhenping66@163.com

a subset of customers is maximized and the length of each vehicle walk is restricted by a pre-specified limit.

The team orienteering problem with capacity constraint and time windows (TOPCT) is an extension of the TOP where each customer has a demand and a time window, each customer must be visited in the given time window and the capacity of each vehicle is limited.

The TOPCT arises in many applications. Consider, for instance, there are  $m$  transport vehicles to deliver goods for customers at geographically distributed locations. The capacity of each vehicle is limited. The demand of each customer is given. If the goods are delivered within a specified time window, the vehicle can get some rewards from the customer, otherwise, it can get no reward. As there is often a limitation on capacity of each vehicle, and each customer has different demand and service time window. Thus, it may not be possible to include all customers requiring service in the service schedules for a given day. Instead, a subset of the customers will be selected. The problem is which customers to choose for inclusion in each of the service vehicle schedules such that the total rewards obtained can be maximal.

Although there are many algorithms proposed for the OP or TOP, they can not be used for the TOPCT since their constraints are different. In this paper, the TOPCT is formulated into an integer linear programming model based on the network flow theory in section 2, then the optimal solution is obtained by solving the integer linear programming. Section 3 gives the computational results. Our model and algorithm are the first approaches for the TOPCT, and high-qualified solutions can be obtained by these model and algorithm on small size of example. Finally, the conclusions are given in section 4.

## 2 The Team Orienteering Problem with Capacity Constraint and Time Window

The Team Orienteering Problem with Capacity Constraint and Time Window (TOPCT) can be formulated as follows: Given an undirected weighted graph  $G = (V, E)$  with node set  $V = \{v_1, v_2, \dots, v_n\}$ , edge set  $E$ . Each node  $v_i$  has a demand  $q_i$ , a time window  $[e_i, l_i]$ , a serving time  $\tau_i$  and a weight  $w_i \geq 0$ . Where  $\tau_i$  can be interpreted as the time needed to serving for node  $v_i$  and  $w_i$  can be interpreted as reward or benefit obtained by serving for node  $v_i$ . The starting node is  $v_1$  and the ending node is  $v_n$ . Each edge  $e_{ij}$  has weight  $t_{ij}$  which can be interpreted as time for a vehicle to travel through edge  $e_{ij}$ . The TOPCT is to find  $m$  walks  $P_1, P_2, \dots, P_m$  in graph  $G$  which satisfies the following conditions. (1)  $P_i$  ( $i = 1, 2, \dots, m$ ) is starting from node  $v_1$  and ending at node  $v_n$ ; (2) the total demands of each walk  $P_i$  ( $i = 1, 2, \dots, m$ ) is no more than the given capacity  $C_i$  ( $i = 1, 2, \dots, m$ ); (3) the total distance of each walk  $P_i$  ( $i = 1, 2, \dots, m$ ) is no more than the given limit  $L_i$  ( $i = 1, 2, \dots, m$ ); (4) for each node in a walk, the time arriving at it satisfies its time window constraints; (5) the total reward or benefit of nodes in all walks  $P_i$  ( $i = 1, 2, \dots, m$ ) are maximal.

Using the network flow theory, suppose there are  $m$  flows of size  $n - 1$  entering into the network from source node  $v_1$ . Then each flow flows through one edge and enters into another node. Whenever a flow flows into a node, a unit of flow is absorbed by the node, the other units will flow out through another edge. Finally, when a flow enters into node  $v_n$ , all the units will be absorbed which means that node  $v_n$  is the ending node where all the

flows entering into it will be absorbed. The nodes and edges along each flow consists of a walk from  $v_1$  to  $v_n$ . The TOPCT can be formulated into an integer linear programming model.

Let  $K = \{1, 2, \dots, m\}$  be the set of vehicles;  $D = \{v_1, v_2, \dots, v_n\}$  be the set of nodes, where  $v_1$  is the starting node and  $v_n$  is the ending node;  $t_{ij}$  is the time passing through edge  $e_{ij}$ ;  $\tau_i$  is the serving time of node  $v_i$  where  $\tau_1 = \tau_n = 0$ ;  $C_k$  is the capacity of vehicle  $k$ ;  $q_i$  is the demand of node  $v_i$ ;  $w_i$  is the reward obtained from serving for node  $v_i$ ;  $[e_i, l_i]$  is the time window of node  $v_i$ ;  $L_k$  is the time limit for vehicle  $k$  to arrive at node  $v_n$ .

Set  $x_{ik}$  be a binary variable,  $x_{ik} = 1$  indicates that vehicle  $k$  serving for node  $v_i$ , while  $x_{ij} = 0$  otherwise.  $y_{ijk}$  be a binary variable,  $y_{ijk} = 1$  indicates that after vehicle  $k$  serve for node  $v_i$  then it directly go to serve for node  $v_j$ ;  $y_{ijk} = 0$  otherwise.  $r_{ik}$  be the time of vehicle  $k$  arriving at node  $v_i$ . If vehicle  $k$  do not serve for node  $v_i$ , then  $r_{ik} = 0$ .  $M$  is a very large positive number.

The TOPCT can be formulated into the following integer linear programming model.

$$\max \sum_{i=2}^{n-1} \sum_{k=1}^m w_i x_{ik} \quad (1.1)$$

$$s.t. \begin{cases} \sum_{j=2}^n y_{1jk} = 1, k = 1, 2, \dots, m & (1.2) \\ \sum_{i=1}^{n-1} y_{ink} = 1, k = 1, 2, \dots, m & (1.3) \\ \sum_{i=1}^{n-1} y_{ijk} = \sum_{l=2}^n y_{jlk} = x_{jk}, j = 2, 3, \dots, n-1, k = 1, 2, \dots, m & (1.4) \\ \sum_{k=1}^m x_{ik} \leq 1, i = 2, 3, \dots, n-1 & (1.5) \\ \sum_{i=2}^{n-1} q_i x_{ik} \leq C_k, k = 1, 2, \dots, m & (1.6) \\ \sum_{i=1}^{n-1} \sum_{j=2}^n y_{ijk} (t_{ij} + \tau_i) \leq L_k, k = 1, 2, \dots, m & (1.7) \\ r_{1k} = 0, k = 1, 2, \dots, m & (1.8) \\ r_{ik} + \tau_i + t_{ij} - r_{jk} \leq M(1 - y_{ijk}), i, j = 1, 2, \dots, n; k = 1, 2, \dots, m & (1.9) \\ e_j x_{jk} \leq r_{jk} \leq l_j x_{jk}, j = 2, \dots, n-1; k = 1, 2, \dots, m & (1.10) \\ r_{nk} \leq L_k, k = 1, 2, \dots, m & (1.11) \\ x_{ik} \in \{0, 1\}, y_{ijk} \in \{0, 1\}, i, j = 1, 2, \dots, n; k = 1, 2, \dots, m & (1.12) \\ r_{jk} \geq 0, j = 1, 2, \dots, n; k = 1, 2, \dots, m & (1.13) \end{cases}$$

The objective function (1.1) maximizes the total reward in all walks.

Constraints (1.2) and (1.3) means that every walk must start from node  $v_1$  and end at node  $v_n$ . Constraint (1.4) means that for every vehicle  $k$ , if it enters into node  $v_j$ , then it must leave node  $v_j$ ; Constraint (1.5) means that for every node  $v_i$ , there is at most one vehicle provide serve for it; Constraint (1.6) is the capacity constraint of vehicle. Constraint (1.7) means that for vehicle  $k$ , the total time used for passing through its walk is no more than  $L_k$ ; Constraint (1.8) means that for every vehicle, the start leaving time is 0; Constraint (1.9) means that the arriving time relationship for two sequence visiting node in a vehicle's walk; Constraint (1.10) means that for every node  $v_j$ , if vehicle  $k$  passing through it, then the arriving time of vehicle  $k$  must satisfy its time window constraint; Constraint (1.11) means that for vehicle  $k$ , the arriving time at node  $v_n$  should not exceed  $L_k$ ; Constraint (1.12) indicates that  $x_{ij}, y_{ijk}$  are the binary variables; Constraint (1.13) indicates that  $r_{jk}$  are nonnegative continuous variables.

Table 1: The demand , serve time, time window and the reward of 14 customers

Customer	Demand	Server time	The earliest serve time	The latest serve time	Benefit
1	0.12	0.3	1	5	80
2	0.32	0.5	1	3.5	250
3	0.15	0.5	2	6	100
4	0.2	0.5	1.5	5	180
5	0.06	0.4	6	8.5	50
6	0.3	0.5	0.5	3	200
7	0.25	0.5	5	8	200
8	0.35	0.5	4	7	400
9	0.1	0.3	1	5	80
10	0.08	0.2	5	7	60
11	0.1	0.2	0	5	70
12	0.12	0.3	0.5	4	80
13	0.2	0.4	3	6	200
14	0.1	0.4	6	8.5	100

### 3 Computational Results

For the small size of the TOPCT problem (for example the number of network nodes is no more than 30), we can directly solve the integer linear programming model to obtain global optimal solutions.

We use an example of 2 transportation vehicles to deliver goods from one distribution center to 14 customers. The sites of the distribution center and 14 customers are illustrated in figure 1. The capacities of two vehicles are 1 and 1.5 tons respectively. For each customer, the demand , the serve time, the time window, the reward are listed in table 1. The time distance between any pair of distribution center and customers are listed in table 2.

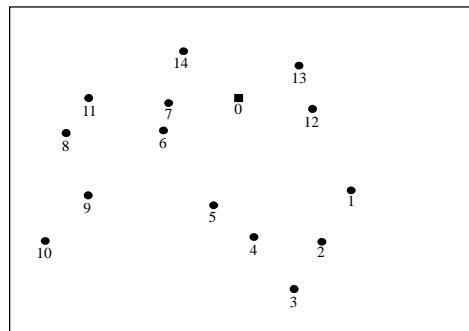


Figure 1: The site of distribution center and customers, where node 0 is the distribution center and nodes 1-14 are customers.

Table 2: The time distance between distribution center and 14 customers

Node	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
0	0	1.04	1.24	1.5	1.1	0.84	0.56	0.48	1.18	1.26	1.66	1	0.52	0.48	0.52
1	1.04	0	0.42	0.84	0.74	0.9	1.32	1.38	1.92	1.7	2	1.91	0.69	1.05	1.58
2	1.24	0.42	0	0.42	0.46	0.78	1.36	1.46	1.88	1.56	1.78	1.96	1.15	1.4	1.79
3	1.5	0.84	0.42	0	0.44	0.82	1.48	1.62	1.9	1.48	1.62	2.06	1.43	1.77	2.03
4	1.1	0.74	0.46	0.44	0	0.38	1.04	1.18	1.48	1.1	1.32	1.08	1.52	1.43	0.73
5	0.84	0.9	0.78	0.82	0.38	0	0.66	0.84	1.12	0.8	1.1	1.2	1.01	1.24	1.24
6	0.56	1.32	1.36	1.48	1.04	0.66	0	0.2	0.64	0.7	1.12	0.57	1.03	1.05	0.65
7	0.48	1.38	1.46	1.62	1.18	0.84	0.2	0	0.7	0.88	1.28	0.54	0.98	0.93	0.43
8	1.18	1.92	1.88	1.9	1.48	1.12	0.64	0.7	0	0.52	0.82	0.33	1.68	1.66	1.03
9	1.26	1.7	1.56	1.48	1.1	0.8	0.7	0.88	0.52	0	0.4	0.79	1.66	1.75	1.31
10	1.66	2	1.78	1.62	1.32	1.1	1.12	1.28	0.82	0.4	0	1.17	2.08	2.19	1.76
11	1	1.91	1.96	2.06	1.08	1.2	0.57	0.54	0.33	0.79	1.17	0	1.52	1.43	0.73
12	0.52	0.69	1.15	1.43	1.52	1.01	1.03	0.98	1.68	1.66	2.08	1.52	0	0.36	1
13	0.48	1.05	1.4	1.77	1.43	1.24	1.05	0.93	1.66	1.75	2.19	1.43	0.36	0	0.79
14	0.52	1.58	1.79	2.03	0.73	1.24	0.65	0.43	1.03	1.31	1.76	0.73	1	0.79	0

In this section, we select the distribution center (node 0) as both the starting and ending nodes of two vehicles' walks. Solving the integer linear programming model by Lingo software, we can obtain the optimal walks of both vehicles, which are  $0 - 12 - 6 - 11 - 9 - 13 - 14 - 0$  for vehicle 1 and  $0 - 1 - 2 - 4 - 3 - 10 - 8 - 7 - 0$  for vehicle 2. The maximum total reward is 2000.

## 4 Conclusion

In this paper, we proposed the team orienteering problem with capacity constraint and time window (TOPCT), and formulated it into an integer linear programming model. By solving the integer linear programming, we can obtain the global optimal solution of the TOPCT. We applied our model and algorithm to an example and obtain the optimal result.

Although we have used our model and algorithm on an example and obtained the optimal solution. We have not test the algorithms on large size of examples with more than 30 nodes. In the future, we will explore some more algorithms for the large size of TOPCT examples.

## Acknowledgments

This work is supported by National Natural Science Foundation of China under Grant No.60873205 and Beijing Natural Science Foundation under Grant No. 1092011. It is also partially supported by the Funding Project for Academic Human Resources Development in Institutions of Higher Learning Under the Jurisdiction of Beijing Municipality (No.PHR201006217), and Funding Project for Base Construction of Scientific Research of Beijing Municipal Commission of Education (WYJD200902).

## References

- [1] Chao I-M., Golden B.L., Wasil E.A., A fast and effective heuristic for the orienteering problem. *European Journal of Operational Research*, 88:475-489,1996.
- [2] Wang X., Golden B.L., Wasil E.A., Using a genetic algorithm to solve the generalization orienteering problem. *The Vehicle Routing Problem: Latest Advances and New Challenges*, 263:273, Springer Science+Business Media,LLC 2008.
- [3] Silberholz J., Golden B.L., The Effective Application of a New Approach to the Generalized Orienteering Problem. *MIC 2009: The VIII Metaheuristics International Conference*
- [4] Golden B.L., Levy L.J., and Vohra R., The orienteering problem. *Naval Research Logistics*, 34(3): 307-318, 1987.
- [5] Chekuri C. and Pal M., A Recursive Greedy Algorithm for Walks in Directed Graphs. *Proc. of IEEE FOCS*: 245-253, 2005.
- [6] Frederickson G. and Wittman B., Approximation algorithms for the traveling repairman and speeding deliveryman problems. *Proc. of APPROX*, 119-133, 2007.
- [7] Wang, Q., X. Sun, B. L. Golden, and Jia J., Using Artificial Neural Networks to Solve the Orienteering Problem, *Annals of Operations Research*, 61, 111- 120,1995.

- [8] Wang Q., Sun X., Golden B. L., Using Artificial Neural Networks to solve generalized orienteering problems. *Intelligent Engineering Systems Through Artificial Neural Networks: Volume 6*, ASME Press, New York, 1063-1068, 1996.
- [9] Liang Y.C., Smith A.E., An ant colony approach to the orienteering problem. *Journal of the Chinese Institute of Industrial Engineers*, 23(5): 403-414, 2006.
- [10] Gendreau M., Laporte G., and Sémét F., A tabu search heuristic for the undirected selective travelling salesman problem. *European Journal of Operational Research*, 106(2- 3):539-545, 1998.
- [11] Tsiligirides T., Heuristics methods applied to orienteering. *Journal of the Operational Research Society*. 35(9):797-809, 1984.
- [12] Chao I-M., Golden B.L., Wasil E.A., The team orienteering problem, *European Journal of Operational Research*, 88: 464-474, 1996.
- [13] Li Z., Zhang S., Zhang X.S., Chen L., Exploring the Constrained Maximum Edge-Weight Connected Graph Problem, *Acta Mathematicae Applicatae Sinica*, 25(4), 697-708, 2009.