Acceleration of the reliable shortest path algorithm in a time-dependent stochastic transport network

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Abstract

In this paper, we propose a modification of the reliable routing algorithm in a stochastic time-dependent network. We consider a stochastic ontime arrival problem. Reliability means maximization the probability of arrival at a destination within a given period of time. Modification of the shortest-path algorithm is aimed to decrease the computation time of the algorithm. The base idea of the proposed modification is to select a certain subset of nodes and links of the graph which can be used for calculating the shortest path. We propose two methods for selecting subset of nodes: based on a bounding box and based on the k-shortest path algorithm. Experimental studies of the base and modified algorithms are carried out on the transport network of Samara, Russia.

Keywords: reliable shortest path; adaptive route; time-dependent network; k shortest path algorithm

1. Introduction

Road congestion is a serious problem of modern society, which has several significant consequences. For traffic participants congestion reduces quality of life, consuming their free time. For organizations, congestion reduces employee productivity and increases freight transportation costs. For the society as a whole, traffic jams lead to the disruption of emergency services, and negatively affects the quality of the environment, causing a large amount of exhaust emissions. Traffic jams threaten traffic safety, raising the level of stress and fatigue among drivers. Thus, it becomes increasingly important to solve the optimal routing navigation problems.

In recent decades, a large number of papers have been devoted to the shortest path problem in transport networks. However, most of them focused their attention on finding the least expected travel time. Several types of models are distinguished depending on the weight type of the road segment. In classical models [1,2,3], the travel time of a road segment is considered to be constant or time-dependent. At the same time, the real situation shows that the travel time is continuously changing and depends on many factors such as time of day, weather conditions, traffic situation, etc. In models with stochastic travel time [4,5,6], the time is represented by a random variable with a time-dependent distribution function. In either case, the optimality condition for routing can be determined in different ways depending on the used objective function. The following types of objective functions can be used:

- 1) minimization of the least expected travel time [4 7];
- 2) maximization of the probability of arriving at the destination at a predetermined time interval [8, 9];
- 3) maximization of the probability that travel time is less than a given threshold [10, 11];
- 4) minimization of the worst possible travel time [12];
- 5) minimization of the travel time to guarantee a given likelihood of arriving on-time in a stochastic network [13].

These types of objective functions can be divided into two groups: the least expected time (LET) (1) and the reliable shortest path (RSP) problem (2-5). The LET problem has been well studied, there are many effective algorithms for different types of the problem [6,7]. Nevertheless, in a number of practical tasks, the path with the least expected time may not be suitable, since it does not take into account the dispersion of the travel time and does not give any guarantees of reliability. In many cases, road users decided to increase travel time and to choose a more reliable route [8]. The problem of finding a reliable shortest path was investigated in [10, 11]. In [14, 15], a shortest path algorithm was proposed, taking into account current and predictive information about the transport flows in the network, which is a modification of the algorithm from [8]. However, the algorithms described in the works are computationally complex and cannot be used to determine the shortest path in large-scale networks in real time.

In this paper, we investigate the method for reliable routing in a time-dependent stochastic transport network. We consider the following optimality criterion: maximizing the probability of arrival at a destination within a predetermined time interval. The aim of this work is to decrease the computation time of the existing algorithm. Acceleration of the algorithm is achieved by choosing a subset of the graph nodes used to find the shortest path. The algorithm proposed by the group of authors in [8] is used as a basic algorithm.

The paper is organized as follows. The second section introduces the basic notation, the problem statement, and describes the base algorithm. In the third section, we propose modifications of the reliable routing algorithm. In the fourth section, experimental studies of the proposed algorithms are presented. Finally, we provide our conclusions.

2. Stochastic on-time arrival problem

The transport network is defined as an oriented, time-dependent stochastic graph: G = (N, A, P) where *N* is a nonempty set of nodes that correspond to road intersections, |N| is the number of nodes, *A* is a set of links that correspond to road segments, |A| is the number of links, *P* is the probabilistic description of the links travel time. We assume that the graph has a spatial reference, i.e. each node of the graph $i \in N$ has coordinates $(x, y)_i$ that are determined by the physical location of the corresponding intersection in the real road network.

Let $T_{ij}(\tau)$ be the travel time of the link $(i, j) \in A$. Travel time $T_{ij}(\tau)$ is represented as a random variable with a probability density function $p_{ij}^{\tau}(t)$ that depends on the time at which the vehicle enters this link.

Let $r \in N$ be the origin node, $s \in N$ be the destination node, T be the maximum amount of time allowed to reach the destination, i.e. the time budget.

The optimal routing policy is defined as the policy of maximizing the probability of arrival at the destination s in a time less than T. This problem is abbreviated as SOTA (Stochastic On Time Arrival) [8, 14, 16].

In papers [8, 19] for the definition of the optimal routing policy, an additional notation is introduced. Let $u_i^{\tau}(t)$ be the probability of reaching the destination node *s* from the node *i* at the time τ with a time budget *t*. $u_i^{\tau}(t)$ is called reliability of the path.

Definition 1 [8]. The optimal routing policy for the SOTA problem can be formulated as follows

$$u_i^{\tau}(t) = \max_j \int_0^t p_{ij}^{\tau}(\omega) u_j^{\tau+\omega}(t-\omega) \, d\omega$$

$$\forall_i \in N, \ i \neq s, \ (i,j) \in A, \ t \in [0,T], \ \tau \ge 0;$$

$$u_s^{\tau}(t) = 1 \ t \in [0,T], \ \tau \ge 0.$$
(1)

The discrete algorithm for solving the problem (1) was described in [8]. We will consider it as the base algorithm. The algorithm can be formulated as follows:

Step 0. Initialization.

$$k = 0,$$

$$u_i^k(x) = 0, \forall i \in N, i \neq s, x \in N, x \in [0, T / \Delta t],$$

$$u_s^k(t) = 1, x \in N, x \in [0, T / \Delta t].$$

Step 1. Update.
FOR $k = 1, 2, ..., L = \lceil T / \Delta t \rceil$

$$\tau^k = k\delta,$$

$$u_s^k(t) = 1, x \in N, x \in [0, T / \Delta t],$$

$$u_i^k(x) = u_i^{k-1}(x), \forall i \in N, i \neq s, (i, j) \in A, x \in N, x \in [0, (\tau^k - \delta) / \Delta t],$$

$$u_i^k(x) = \max_j \sum_{h=0}^x p_{ij}(h)u_j^{k-1}(x - h), \forall i \in N, i \neq s, (i, j) \in A, x \in N, x \in [((\tau^k - \delta) / \Delta t + 1), (\tau^k / \Delta t)]$$

END FOR

where Δt is a sampling interval, δ is the minimum travel time across the transport network.

The decision rule for selecting the next node j in the transport network graph for a given time budget t and calculated probabilities $u_i(x)$ is as follows:

$$j = \arg\max_{i \in \mathcal{N}} u_i(t) \tag{2}$$

In the described algorithm, all graph nodes are used to construct the shortest route that makes this algorithm computationally complex and not applicable for finding the shortest path in large-scale networks in real time. In the next section we propose modifications of the algorithm, consisting in selecting a certain subset of the graph nodes that will be used to find the shortest route.

3. Modified algorithm

The main idea of the modification is to achieve the acceleration of the base algorithm by reducing the number of nodes that are considered as candidates in the shortest path. We propose two methods for selecting a subset of nodes and links of the graph used to construct a reliable shortest path: on the basis of the bounding box and on the basis of the k-shortest path algorithm.

3.1. Subset based on a bounding box

The obvious way to reduce the number of nodes used to find the shortest route is to select only those nodes whose coordinates is inside the bounding box between the origin and destination nodes

³rd International conference "Information Technology and Nanotechnology 2017"

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Let $(x, y)_{r \in N}$ bet the coordinates of the origin node and $(x, y)_{s \in N}$ be the coordinates of the destination node.

The bounding box coordinates can be written as follows:

$$\{x_{\min} = \min(x_r, x_s) - \Delta, y_{\min} = \min(y_r, y_s) - \Delta, x_{\max} = \max(x_r, x_s) + \Delta, y_{\max} = \max(y_r, y_s) + \Delta\},\$$

where Δ is the buffer distance chosen experimentally.

Then the subsets of nodes and links of the graph can be defined as follows:

$$N = \{i \in N : x_{\min} \le x_i \le x_{\max} \land y_{\min} \le y_i \le y_{\max} \},\$$

$$\overline{A} = \{(i, j) \in A : i \in \overline{N} \land j \in \overline{N} \}.$$

This method of selecting subsets is computationally simple, but the resulting subsets may be redundant or, conversely, insufficient, depending on the network structure and the location of the origin and destination nodes..

3.2. Subset based on the k-shortest path algorithm

The k-shortest path algorithm is an extension algorithm of the shortest path routing algorithm in a given network [17]. Depending on the problem statement, the shortest paths may or may not contain the same nodes and links. We suggest that the nodes and links that are included in the shortest path are marked as passed and cannot be used to construct the next shortest paths. To find different shortest paths, various algorithms can be used; in this paper we use the Dijkstra algorithm [1].

For a formal description of the method, we introduce additional notations.

Let $d_{rs}^{k} = \{v_{1}^{k}, v_{2}^{k}, ..., v_{M}^{k}\}$ be the kth shortest path between nodes *r* and *s*,

where $v_m^k \in N, m = \overline{0, M^k - 1}$ is the graph node included in the kth shortest path:

 M^k is the number of nodes in the shortest path;

k = 1, K is the shortest path index;

K is the number of shortest path chosen experimentally.

Then the subsets of nodes and links of the graph can be defined as follows:

$$\overline{N} = \left\{ i \in \left\{ d_{rs}^{k} \right\}_{k=\overline{1,K}} \right\},\$$
$$\overline{A} = \left\{ (i, j) \in A : i \in \overline{N} \land j \in \overline{N} \right\}.$$

The method of selecting a subset of nodes based on the k-shortest path algorithm has a greater computational complexity than the method based on the bounding box, but it does not depend on the network structure and allows to adjust the size of the subset by selecting the appropriate parameter K.

3.3. Modified reliable routing algorithm

The modified reliable routing algorithm uses subsets of nodes \overline{N} and links \overline{A} of the graph. In the form of a pseudo code, the algorithm can be written as follows:

k = 0, $u_i^k(x) = 0, \forall i \in N, i \neq s, x \in \overline{\mathbb{N}}, x \in [0, T / \Delta t],$ $u_s^k(t) = 1, x \in \overline{\mathbb{N}}, x \in [0, T / \Delta t].$

Step 1. Update.

FOR
$$k = 1, 2, ..., L = \left\lceil T / \Delta t \right\rceil$$

 $\tau^k = k\delta$,
 $u_s^k(t) = 1, x \in \mathbb{N}, x \in [0, T / \Delta t]$,
 $u_i^k(x) = u_i^{k-1}(x), \quad \forall i \in N, i \neq s, (i, j) \in \overline{A}, x \in \overline{\mathbb{N}}, x \in [0, (\tau^k - \delta) / \Delta t]$,
 $u_i^k(x) = \max_j \sum_{h=0}^x p_{ij}(h) u_j^{k-1}(x-h), \quad \forall i \in N, i \neq s, (i, j) \in \overline{A}, x \in \overline{\mathbb{N}}, x \in [((\tau^k - \delta) / \Delta t + 1), (\tau^k / \Delta t)]$
END FOR

The subsets \overline{N} and \overline{A} are selected by one of the methods described above.

4. Results and Discussion

The purposes of the conducted experiments were to compare the base and modified algorithms according to the criteria of the computation time of the algorithm and the reliability $u_i^{\tau}(t)$ of the calculated route.

The experiments were carried out on the transport network of Samara, Russia. The transport network consists of 9721 nodes and 26088 links. As the weight of the road link, we used travel time data averaged over a ten-minute interval. As the probability density function on the link (i, j) we used the lognormal distribution.

To conduct experiments, 10 initial and 10 final vertices of the transport network have been chosen, located at a considerable distance from each other. For each pair of vertices, the shortest paths have been found using the base and modified algorithms. To study the algorithm based on the bounding box, the following values of the buffer distance were chosen: $\Delta = \{200; 300; 400; 500; 600\}$. The study of the algorithm based on k-shortest paths was conducted depending on the number of shortest paths $K = \{3;4;5;6;7\}$.

First of all, we compare the algorithms by the reliability criterion (the probability of reaching a destination node with a given time budget). The results of the comparison are presented in Table 1.

Table	1 A1	gorithm	's com	narison	by the	reliability	
raute	1. AI	gomunn	s com	parison	by the	remaining.	

Algorithms	Base algorithm	Bounding box				k-shortest paths					
		200	300	400	500	600	3	4	5	6	7
Reliability	0.99836	0.99568	0.99567	0.99624	0.996328	0.99642	0.99494	0.99512	0.99512	0.99523	0.995239

All algorithms have shown similar results for the chosen criterion for specified time budgets, modified algorithms slightly inferior to the base one.

The next stage of the experimental analysis was the comparison of algorithms by the criterion of the computation time. The computational time of the base algorithm was about 10 minutes in average, depending on the used time budget T. Figure 1 shows a histogram of the modified algorithms acceleration time in comparison with the base algorithm for different values of the buffer distance and the number of shortest paths.



Parameters of the algorithm

Fig 1. Acceleration of the modified algorithm.

The best results on this criterion have been shown by the algorithm based on k-shortest paths. The computation time in comparison with the base algorithm decreased by 10-35 times depending on the number of shortest paths K. Modification of the base algorithm based on the bounding box allows to decrease the computation time by 8-10 times. Note that the value of the buffer distance does not have a strong effect on the speed of the algorithm, but too much value can significantly increase the computation time.

The results show, that the proposed modifications allow to significantly decrease the computation time of the base algorithm (in 8-35 times) with a slightly decrease in the reliability of the shortest path. The modified algorithm can be used to find the reliable shortest route in large-scale networks in real time.

5. Conclusion

In this paper we have proposed modifications of the reliable routing algorithm in a time-dependent stochastic transport network. Reliability means maximizing the probability of arrival at a destination within a time budget. Two modifications are proposed for the purpose of decreasing the computation time of the algorithm: based on the bounding box and based on the kshortest path algorithm. The proposed modified algorithms are compared with the base algorithm on the Samara transport network. Image Processing, Geoinformation Technology and Information Security / I. Abdulganiev, A. Agafonov

The results of experimental studies have shown that the proposed modifications allow to significantly decrease the computation time (by 8-35 times) with practically identical reliability of the routes. The modified algorithms allow to find the shortest route in large-scale networks in real time.

Further research includes:

- studies related to the route choice depending on the traffic flows;
- studies related to the choice of a travel time budget.

Acknowledgements

This work was supported by the Russian Foundation for Basic Research (RFBR) grant 16-37-00055-mol-a.

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