

Optimal dispatching method of traffic incident rescue resource for freeway network

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Abstract: An optimal resource dispatching method is proposed to solve the multiple-response problem under the conditions of potential incidents on freeway networks. Travel time of the response vehicle is selected, instead of route distance, as the weight to reflect the impact of traffic conditions on the decisions of rescue resources. According to the characteristics of different types of rescue vehicles, the dispatching decision-making time is revised to show the heterogeneity among different rescue vehicle dispatching modes. The genetic algorithm is used to obtain the solutions to the rescue resources dispatching model. A case study shows that the proposed method can accurately reveal the impact of potential incidents on the costs of rescues according to the variations in the types and quantities of rescue resources, and the optimal dispatching plan with respect to potential incidents can be obtained. The proposed method is applicable in real world scenarios.

Key words: optimal dispatching; potential incident; genetic algorithm; rescue resource; freeway network

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Traffic incident response plays an important role in traffic management for maximizing freeway efficiency. A traffic incident is a non-recurrent event that can lead to traffic flow disruptions and capacity reductions^[1], and the resource dispatching methods are required to reduce the decision-making time by separately analyzing various factors that contribute to incident responses. Especially, the potential incidents that may occur at the same locations will affect the resource dispatching decisions. Therefore, it is necessary to develop a model for optimizing the resource dispatching under the demand of potential incidents.

Targeting the methods for traffic rescue resource dispatching, Zografos et al.^[2-5] studied rescue resources in an area with dispatching events. These methods fit the dispatching type according to the sequence of events for

local sections along the road, but they are limited in that an incident corresponds to only one response vehicle. Subsequently, Zografos et al.^[6-7] studied the dispatching types based on the shortest distance from various rescue dispatching points to the current incident point. However, because of time-varying traffic congestion levels, the shortest distance does not mean the shortest travel time for the closest response vehicles to reach the incident point. In regard to the resource dispatching under potential incidents, Sherali et al.^[1,8] defined the rescue cost of potential incidents as an opportunity cost and developed an opportunity cost-based dispatching model. When there are potential incidents in a freeway network, the dispatching of the closest rescue point for a current incident does not necessarily achieve global optimization. This method incorporates the rescue cost of potential incidents into the total rescue cost and attempts to solve the problem of resource dispatching on the freeway network.

Focusing on the resource dispatching under the freeway sections characteristics, both domestic and international researches still lack the ability to tackle the following issues. First, current methods do not consider the differences in dispatching the response vehicles. Practically, the relationships between the types and driving characteristics of the response vehicles are not considered. Secondly, current methods usually use a constant dispatching decision-making time for a response vehicle and do not reflect the impacts of different incident levels and different types of response vehicles. Finally, according to the conflicts among the performance indicators for dispatching optimization, e. g., the conflict between response time and rescue cost, the conventional method cannot meet the requirements for optimized multi-incidents and multi-response dispatching problems^[9].

In summary, the minimum travel time or the minimum distance from the rescue point to the incident point cannot comprehensively reflect the impact of the incident level on the dispatching time and the differences among the driving speeds of different response vehicles^[10]. By using the minimized summation of the dispatching time and the travel time as the objective function, this paper first proposes a method to estimate the decision time for different incident levels and the travel time of different response vehicles. Then, a genetic algorithm is presented to solve

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the problem of optimized response vehicle dispatching for potential incidents with multi-rescue points in the freeway network.

1 Modeling of Incident Rescue Resource Dispatching for Freeway Network

1.1 Conventional incident rescue resource model

Let L denote the set of the rescue points; F denotes the set of the current incident points; N denotes the set of network nodes; r_i denotes the response vehicles that are arranged for road rescue points where $i \in L$; n_f denotes the number of response vehicles that are required for a current incident point where $f \in F$ and $F \subset N$; λ_{if} denotes the minimum section weight that is calculated as the distance from rescue point i to the incident point f ; x_{if} denotes the number of response vehicles that are dispatched from i to f ; H denotes the set of potential incident points where $H \subset N$; P_h denotes the probability of potential incidents at node h where $h \in H$; λ_{ih} denotes the minimum section weight that is calculated as the distance from rescue point i to potential incident point h , $\lambda_h \equiv \min \lambda_{ih}$; and y represents an arbitrary y_{ih} . If the response vehicles are dispatched from i to h , then $y_{ih} = 1$; otherwise, $y_{ih} = 0$. Let s_i denote the number of the remaining vehicles that can be dispatched for potential incident points after the dispatching of the vehicles from i to f , and the conventional rescue resource model^[11] can be formulated as

$$\min \left[\sum_{i \in L} \sum_{f \in F} \lambda_{if} x_{if} + \sum_{i \in L} \sum_{h \in H} P_h (\lambda_{ih} - \lambda_h) y_{ih} \right] \quad (1)$$

$$\sum_{f \in F} x_{if} + s_i = r_i \quad \forall i \in L, s_i \geq 0 \quad (2)$$

$$\text{s. t. } \sum_{i \in L} x_{if} = n_f \quad \forall f \in F \quad (3)$$

$$\sum_{i \in L} y_{ih} = 1 \quad \forall h \in H \quad (4)$$

$$y_{ih} \leq s_i \quad \forall i \in L, h \in H \quad (5)$$

$$x \geq 0 (\text{and integer}), y \text{ binary} \quad (6)$$

The objective function (Eq. (1)) is to minimize the total cost of incident rescues, including the direct cost of current incidents and the opportunity cost of potential incidents. Eq. (2) ensures that the number of vehicles dispatched from any depot does not exceed the number of service vehicles that are available at the depot. Eq. (3) stipulates that the number of vehicles dispatched to any current incident node should meet the service vehicle requirements at that node. Eq. (4) states that a service vehicle should be ready for dispatching rescue resources to potential incidents at any node in the network, whereas Eq. (5) states that such an additional vehicle in Eq. (4) can be dispatched from a depot only if it is available at that depot. Eq. (6) imposes the integrality and binary restrictions of the decision variables.

The limitation of the conventional model is that the rescue point and the section weight are fixed. Consequently, the impact of different types of resources on generating plans cannot be reflected. As is known, the dispatching decision-making time should vary with the type and the level of the incident. Therefore, the conventional model needs to be improved.

1.2 Modified incident rescue resource model

Previous studies have shown that the levels of incidents affect the dispatching decision-making time and that the driving characteristics of response vehicles are important factors in determining their travel time. Therefore, the dispatching time in this study is defined as the combination of the dispatching decision-making time and the travel time of response vehicles. The decision-making time refers to the time from the moment of determining the level of an incident to the moment of generating the dispatching plan. The travel time of response vehicles refers to the vehicles used on the rescue path. The dispatching time is given as

$$T_{if} = t_f + \frac{\lambda_{if}}{v_{if}}, \quad T_{ih} = t_h + \frac{\lambda_{ih}}{v_{ih}} \quad (7)$$

where T_{if} is the dispatching time from rescue point i to current incident point f ; T_{ih} is the dispatching time from rescue point i to potential incident point h ; t_f and t_h are the dispatching decision times determined by the levels of a current incident and a potential incident, respectively; λ_{ih}/v_{ih} is the travel time of response vehicles; and v_{if} and v_{ih} are the average driving speeds of the vehicles passing road sections from rescue point i to a current incident point f and a potential incident point h , respectively. According to the level of traffic incidents, the reference values of the dispatching decision-making time^[11] are given in Tab. 1.

Tab. 1 Dispatching decision-making time according to levels of incidents

Levels of incidents	Minor incident	General incident	Major incident	Serious incident
Decision time	3	5	7	10

The average driving speeds of different types of rescue vehicles^[12] are given in Tab. 2.

Tab. 2 Average driving speed of rescue vehicles on road

Types of vehicles	Average speed/(km · h ⁻¹)
Police patrol and administration vehicles	80
Small wrecker and cranes	60
Medium wrecker and cranes	50
Large wrecker and cranes	40

By replacing the section weights λ_{if} and λ_{ih} in Eqs. (1) to (6) with the dispatching times T_{if} and T_{ih} in Eq. (7), and defining $T_h \equiv \min T_{ih}$, the modified potential demand based model is given as

$$\begin{aligned}
& \min \left[\sum_{i \in L} \sum_{f \in F} T_{if} x_{if} + \sum_{i \in L} \sum_{h \in H} P_h (T_{ih} - T_h) y_{ih} \right] \\
& \sum_{f \in F} x_{if} + s_i = r_i \quad \forall i \in L \\
& \sum_{i \in L} x_{if} = n_f \quad \forall f \in F \\
& \text{s. t. } \sum_{i \in L} y_{ih} = 1 \quad \forall h \in H \\
& y_{ih} \leq s_i \quad \forall i \in L, h \in H \\
& x \geq 0 (\text{and integer}), y \text{ binary}
\end{aligned} \quad (8)$$

Compared with the conventional model where the rescue cost is computed as the product of rescue route distance and the number of dispatched vehicles, the proposed model calculates the rescue cost as the product of the dispatching time and the number of dispatched vehicles to reflect the actual situation for dynamic traffic conditions.

2 Dispatching Plan Based on Genetic Algorithm

Response vehicles include traffic police, road patrol cars, wreckers, fire engines, ambulances, etc. In this section, a genetic algorithm (GA)^[13], including chromosome encoding, fitness function, selection, crossover, and mutation, is developed to solve the modified dispatching model for the response vehicle dispatching. The algorithm is coded in Matlab 7 by installing the GATbx (genetic algorithm optimization toolbox) developed by the University of Sheffield.

2.1 Chromosome encoding

In the chromosome encoding, the digit position in a chromosome is defined as the number of response vehicles dispatched from rescue point L_i to incident point F_i . The general structure of the chromosome coding is shown in Tab. 3.

Tab. 3 General structure of chromosome coding

Point	L_1				L_2				\dots	L_n			
	F_1	F_2	\dots	F_{m1}	F_1	F_2	\dots	F_{m2}		F_1	F_2	\dots	F_{mn}
Number of response vehicles	X_1	X_2							\dots				X_{mn}

The numerical value X_1 on the first position of a chromosome is the number of rescue vehicles dispatched from rescue point L_1 to accident point F_1 . In a road network with multi-rescue points and multi-accident points, each rescue point has its specific coverage due to the limits of the rescue time, and only the rescue points and accident points within the rescue point coverage are selected for encoding. The number of rescue points contained in a chromosome is determined by the location of the current incident, which means that rescue point L_x will be included in the process of encoding if the current accident occurs within the coverage of rescue point L_x . According to the encoding characteristics, the encoding constraints are set as follows.

First, for the potential accident point, the total number of rescue vehicles obtained from different rescue points should not exceed the total demand for any given point; i. e. ,

$$\sum_{h \in H} X_i \leq n_h \quad (9)$$

where X_i is the number of response vehicles dispatched from the rescue point to the incident point and n_h is the total number of demand vehicles for the potential incident.

Secondly, for the current accident point, the total number of rescue vehicles obtained from different rescue points should be equal to the number of demand vehicles; i. e. ,

$$\sum_{f \in F} X_i = n_f \quad (10)$$

where n_f is the total number of demand vehicles for the current incident.

Thirdly, the total number of response vehicles dispatched from each rescue point should not exceed the number of response vehicles allocated for this rescue point; i. e. ,

$$\sum X_i \leq r_i \quad (11)$$

where r_i is the number of response vehicles equipped at this rescue point.

2.2 Fitness function

The fitness function is the objective function of the algorithm. The key point of dispatching is the formation of dispatching plans rather than the specific value of the rescue costs. Therefore, the fixed variables in the objective function in Eq. (8) can be eliminated, whereas the parameters that vary with the actual situation can be kept.

Due to the fact that $\sum_{i \in L} y_{ih} = 1, \forall h \in H$, if the potential incident points are fixed, the term of $\sum_{i \in L} \sum_{h \in H} P_h (-T_h y_{ih})$

$= \sum_{i \in L} P_h (-T_h)$ can be removed from the objective function in Eq. (8).

Therefore, the latter term of Eq. (8) can be simplified as $\sum_{i \in L} \sum_{h \in H} P_h T_{ih} x_{ih}^*$, where x_{ih}^* is the number of response vehicles dispatched from i to h . Set the occurrence probability of the current incident as 1, and the fitness function will be the product of the probability of the incidents, the dispatching time from the rescue point to the incident point, and the number of vehicles dispatched to the incident point. In this sense, the objective function is $\min \sum_{i=1}^n P_i T_i x_i$, where n is the chromosome coding length. When the genetic algorithm is used to calculate the maximum value, the objective function can be converted to the form as $\max(a - \sum_{i=1}^n P_i T_i x_i)$, where a

Tab. 6 Types and quantities of response vehicles required

Accident point	Freeway	Police patrol vehicles	Administration vehicles	Wreckers				Cranes	
				3 t	8 t	48 t	60 t	40 t	60 t
0	Ning-Hang	3	2	0	0	2	0	1	1

3.2 Generation of dispatching plan

The dispatching plan is divided into several independent parts according to the type of rescue vehicles. For example, a dispatching plan can be composed of five parts, i. e., dispatching of police patrol cars, road patrol cars, 48 t-class wreckers, 40 t-class cranes, and 60 t-class cranes.

Tab. 7 Individual coding structure of police patrol vehicles

Dispatching points	Current and potential incident points	Coding
A	0, 5, 6, 22, 23, 24, 25, 28	$X_1, X_2, X_3, X_4, X_5, X_6, X_7, X_8$
C	0, 5, 6, 17, 18, 22, 23, 24, 25	$X_9, X_{10}, X_{11}, X_{12}, X_{13}, X_{14}, X_{15}, X_{16}, X_{17}$
D	0, 17, 18, 22, 23, 28	$X_{18}, X_{19}, X_{20}, X_{21}, X_{22}, X_{23}$

In the individual coding, X_i is the value of the i -th position corresponding to the number of patrol cars that are dispatched from the dispatching point to the accident point.

3.2.2 Constraint conditions for developing fitness function

The constraint conditions represented in Eqs. (9), (10) and (11) are required to be determined when applying Eq. (12) for the fitness function. As shown in Tab. 6, a total number of three police patrol cars are required to be dispatched for the incident. If rescue points A, C and D are equipped with two police patrol cars, respectively, the total number of police patrol cars dispatched by each rescue point cannot exceed 2. Meanwhile, for the potential incident points, it is deemed that the demand of rescue vehicles should not exceed 1. Therefore, the constraint conditions to be met by the dispatching plan are formulated as $X_1 + X_9 + X_{18} = 3$, $X_2 + X_{10} \leq 1$, $X_3 + X_{11} \leq 1$, $X_{12} + X_{19} \leq 1$, $X_{13} + X_{20} \leq 1$, $X_4 + X_{14} + X_{21} \leq 1$, $X_5 + X_{15} + X_{22} \leq 1$, $X_6 + X_{16} \leq 1$, $X_7 + X_{17} \leq 1$, $X_8 + X_{23} \leq 1$, $X_1 + X_2 + X_3 + X_4 + X_5 + X_6 + X_7 + X_8 \leq 2$, $X_9 + X_{10} + X_{11} + X_{12} + X_{13} + X_{14} + X_{15} + X_{16} + X_{17} \leq 2$, $X_{18} + X_{19} + X_{20} + X_{21} + X_{22} + X_{23} \leq 2$.

3.2.3 Parameter design of genetic operators

Combined with the actual situation in this case, the number of chromosomes in the population is set to be 50;

Taking the dispatching of police patrol cars as an example, the process of obtaining the optimal dispatching plan by using the genetic algorithm is illustrated as below.

3.2.1 Creation of chromosome encoding

According to Fig. 1, the incident at point 0 lies within the coverage of dispatching points A, C and D. The potential incident points within A, C and D rescue points and the chromosome coding structure are shown in Tab. 7.

the largest genetic algebra is selected as 150; the generation gap value is set to be 0.9; the probability of a single-point crossover is set to be 0.7; and the mutation probability is set to be 0.004. By using the genetic algorithm, the output of the dispatching result is achieved after an iteration of 150.

3.2.4 Dispatching plan

The dispatching plan obtained through the genetic algorithm is shown in Tab. 8. It is observed that the optimal dispatching plan does not necessarily select the rescue point with the shortest dispatching time to the incident point when considering the potential incidents in the road network. As indicated in Fig. 1, point A is the closest rescue point from incident point 0, but there are some potential incident points in the vicinity of point A. Comparatively, the number of potential incident points is fewer for point D. In this sense, the preferred dispatching of resources from rescue point A is obviously not the global optimization dispatching plan. This is also consistent with the results from the improved dispatching model. As shown in Tab. 8, the optimal dispatching plan can be quickly found by applying the genetic algorithm. Moreover, the improved dispatching method in this study can reflect the impact of potential accident points in generating the resource dispatching plan.

Tab. 8 Comprehensive dispatching plan

Dispatching points	Road	Police patrol vehicles	Administration vehicles	Wreckers				Cranes	
				3 t	8 t	48 t	60 t	40 t	60 t
A	Hun-Ning	1	1	0	0	1	0	0	0
D	Ning-Hang	2	1	0	0	1	0	1	1

4 Conclusion

This paper proposes an improved dispatching model for road network rescue resources for incident response, in

which the dispatching decision-making time and the driving speed range of the rescue vehicles are given, and the calculation of the resource dispatching time is designed. The proposed model describes the resource dispatching

cost for both current incidents and potential incidents in the road network for fully reflecting the corresponding relationships between the levels of incidents and dispatching decision-making time. In addition, the model considers the impact of different response vehicles on the dispatching time which is more consistent with a realistic rescue dispatching process.

The improved model is solved by the genetic algorithm. First, the real coding is adopted and the objective function is changed. Secondly, the fitness function is designed. Finally, the selection, crossover, and mutation operators are designed for the genetic parameters. In addition, the genetic optimization steps to solve the improved model are given to obtain the optimal resource dispatching plan.

References

[1] Sherali H D, Subramanian S. Opportunity cost-based models for traffic incident response [J]. *Journal of Transportation Engineering*, 1999, **125**(3): 176 – 185.

[2] Zografos K G, Nathanail T, Michalopoulos P. Analytical framework for minimizing freeway incident response time [J]. *Journal of Transportation Engineering*, 1993, **119**(4): 535 – 549.

[3] Goldberg J, Dietrich R, Chen J M, et al. Validating and applying a model for locating emergency medical vehicles in Tucson, AZ[J]. *European Journal of Operational Research*, 1990, **49**(3): 308 – 324.

[4] Yamada T. A network flow approach to a city emergency evacuation planning [J]. *International Journal of Systems Science*, 1996, **27**(10): 931 – 936.

[5] Yuan Qian, Liu Shuyan. Research on the optimum of the

Hungarian method[J]. *Journal of Wuhan University of Technology*, 2007, **29**(3): 146 – 149. (in Chinese)

[6] Zografos K G, Androutsopoulos K N, Vasilakis G M. A real-time decision support system for roadway network incident response logistics [J]. *Transportation Research Part C*, 2002, **6**(10): 1 – 18.

[7] Haghani A, Tian Q, Hu H J. A simulation model for real-time emergency vehicle dispatching and routing [J]. *Transportation Research Board*, 2003, **1882**: 176 – 183.

[8] Carter G M, Chaiken J M, Ignall E. Response areas for two emergency units[J]. *Operations Research*, 1972, **20**(3): 571 – 594.

[9] Chai Gan, Zhu Canghui. Decision model and algorithm for traffic rescue resource dispatching on expressway[J]. *Journal of Southeast University: Natural Science Edition*, 2009, **25**(2): 252 – 256. (in Chinese)

[10] Zhang Ying, Zhou Gang, Huang Xiyue. Simulation of rescue model based on optimal plan[J]. *Computer Simulation*, 2006, **23**(1): 207 – 209. (in Chinese)

[11] Cheu R L, Huang Y, Huang B. Allocating emergency service vehicles to serve critical transportation infrastructures [J]. *Journal of Intelligent Transportation Systems*, 2008, **12**(1): 38 – 48.

[12] Wang Wei, Guo Xiucheng. *Traffic engineering* [M]. Nanjing: Southeast University Press, 2000. (in Chinese)

[13] Wang Ling. *Intelligent optimization algorithms with applications*[M]. Beijing: Tsinghua University Press, 2001. (in Chinese)

[14] Ghoseiri K, Ghannadpour S F. Multi-objective vehicle routing problem with time windows using goal programming and genetic algorithm[J]. *Applied Soft Computing*, 2010, **10**(4): 1096 – 1107.

道路路网交通应急事件救援资源的优化派遣方法

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摘要:为了解决潜在事故条件下路网交通救援多响应问题,提出了一个优化的应急救援资源派遣方法. 用车辆行驶时间代替传统方法中的行驶距离作为路段权值,以真实反映交通状态对救援资源派遣决策的影响;依据不同种类救援车辆行驶特征修正派遣决策时间,表现了不同种类救援车辆派遣方式的差异性;使用遗传算法获得了所建救援资源派遣模型的优化解. 应用案例结果表明,所提方法能够根据当前与潜在事故所需资源种类和数量的变化,准确揭示潜在事故对救援成本的影响,并获得潜在事故响应的路网优化派遣方案,具有工程适用性.

关键词:优化派遣;潜在事故;遗传算法;救援资源;道路路网

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