ALGORITHMS FOR TRAFFIC MANAGEMENT IN THE INTELLIGENT TRANSPORT SYSTEMS

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Traffic jams interfere with the drivers and cost billions of dollars per year and lead to a substantial increase in fuel consumption. In order to avoid such problems the paper describes the algorithms for traffic management in intelligent transportation system, which collects traffic information in real time and is able to detect and manage congestion on the basis of this information. The results show that the proposed algorithms reduce the average travel time, emissions and fuel consumption. In particular, travel time has decreased by about 23%, the average fuel consumption of 9%, and the average emission of 10%.

Keywords: traffic jams; intelligent transport systems; detection and control of congested roads; automated control system; control; simulation; algorithm.

Introduction

Traffic jams are a major problem in large urban centers and directly impact on the economy due to increase fuel consumption. In addition, it affects the productivity of society as congestion dramatically increase travel time from one point to another and causing irreparable damage to the environment due to the increase of the emission of polluting gases. According to statistics, 85% is associated with time wasted in traffic, 13% due to the increase in fuel consumption and 2% can be attributed to an increase in emissions of polluting gases [1, 2].

To prevent congestion and improve the efficiency of transport systems, there is an active implementation of intelligent transport systems (ITS). Intelligent transport systems use advances in technology in the area of processing, sensing and communication in order to monitor
traffic conditions in a particular region to manage traffic and reduce congestion as well as reduce the number of road accidents. In addition, they can be used to provide information and entertainment services to drivers and passengers in order to make their journey a more pleasant process [2, 3]. A Vehicular Network is an important component in an ITS. In this network, vehicles are equipped with processors, sensors and wireless communication interfaces so that they can communicate with one another and with the elements in the network infrastructure (RSU – Road Side Unit), thus creating an ad hoc network while vehicles move through roads. Thus, when driving vehicles on the roads equipped with ITS, it created a special network (4, 7, 10).

Some researchers [8, 9] proposed the architecture for the design of ITS, which range from centralized solutions to distributed decision making solutions. These solutions use both the information provided by vehicles and the characteristics of the infrastructure of the roads to detect and control congestion. Thus, after detecting a congested area or road, a rerouting mechanism is applied to the vehicles to prevent more vehicles from entering the affected area, which results in a more efficient traffic. Thus, this work is aimed at the implementation of a distributed ITS to detect congestion and traffic management.

The proposed system consists of a set of roadside controllers. Each controller is responsible for roadside vehicles management and congestion detection only within the coverage area of its range. Moreover, the system includes a congestion control mechanism that periodically performs routing of all vehicles so they do not pass through congested areas. The system is divided into four stages, the distribution of the roadside unit RSU collection and transmission of data, congestion detection and congestion control [5].

The algorithm of placing of roadside controllers (RSU) on the map

Allocation of road side controller (RSU) is uniform depending on their radius and size of the map. Thus, the larger the range, the smaller the number of roadside controllers is used to cover the map. In al-
algorithm 1, when placing roadside controllers are taken into account: Width – the total width of the card, Length – total length of the card and Radius – range roadside communication controller. Algorithm 1 based on the size of geographic area and radius roadside communication controller (line 3) calculates the number of roadside controllers that are needed to cover the entire map. Then the coordinates of the placement of roadside controllers in relation to the map.

The algorithm of placing of roadside controllers (RSU) on the map

Input: Width // Total width of the map
1 Length // Total length of the map
2 Radius // Operating range of RSU

Output: Coordinates of the points of each RSU
// returns the amount of RSUS to cover the entire map
3 rsus ← getNumberOfRSUs(Width; Length; Radius);
4 foreach γ ∈ rsus do
   // returns the coordinates (x; y) for each RSU
5 Rsus_coordinates.add(getCoordinates(r));
6 end

Communication range of roadside controllers influences the cost of implementing the system, i.e, the smaller the radius of the connection, the greater the number of roadside controllers required to cover the entire map.

Algorithm to collect and transmit data in the RSU

This algorithm provides data collection from RSU to obtain information about the events taking place within their communication range, as well as information about the behavior of the vehicle in their coverage area. In which vehicles periodically send information about the position, speed, distance covered, time and direction of travel movement in the near RSU. This information is transmitted via the long-range communication (LTE or 3G).

Algorithm 2 describes a process in which each vehicle sends the information to the nearest RSU. The algorithm takes as input the set of vehicles and a set of coordinates of each RSU, which are respective-
ly represented N and RSUs-Coordinates. Thus, each vehicle collects information and identifies the nearest RSU in relation to its position (Line 4). Moreover, each vehicles create messages containing the collected information and sends it to the nearest RSU (Lines 5–6). After collecting and transmitting data to the RSU, it is necessary its analysis and interpretation. Each roadside controller creates a graph $G = (V, A)$, where the vertices $V$ – a set of transitions within the reach of roadside controller (representing the top) and the arc $A$ – is a set of roads that connect the top. Each road (edge) of graph $G$ has a weight defined by the average speed with which the vehicles travel it and the maximum speed allowed on it. This weight is inversely proportional to the speed at which the vehicle moves along the road. This weight is inversely proportional to the speed at which the vehicle moves along the road. Thus, the closer to the speed of movement of the vehicle with speed limited, the lower the weight of the road. Conversely, the lower the speed of the vehicle relative to the maximum speed, the higher the weight.

Алгоритм of Collecting and sending data to the nearest RSU

Input: $N$ // Set of vehicles on the network
1 $RSUs\_Coordinates$ // Set with the coordinates of each RSU
2 foreach $v \in N$ do
   // returns information about vehicle V
3 status $v.getInformations();$
   // returns the nearest RSU to vehicleV
4 nearestRsu $RSUs\_Coordinates.getNearstRSU(v.getPosition());$
   // creates a message with the information of vehicle v and sends it to the nearest RSU
5 message $\leftarrow$ CreateMessage(status);
6 message.sendTo(nearestRsu);
7 end

The algorithm of alternative routing of vehicle

The change of route of vehicle movement occurs periodically depending on the current traffic. Each roadside controller is responsible
for the selection of a new route within its range. The distortion depends on the intensity of vehicle movement and the weight of the road. When an overload vehicle are forwarded to another route to avoid congested areas. As the base routing algorithm is used to find the shortest route in time. However, the set of K shortest paths is calculated as alternative ways in which K can be easily adjusted to adapt better to the behavior in the network. Of these alternative routes, the only way is selected by using the probability distribution of Boltzmann.

Algorithm 3 describes the process of alternative routing of the vehicle. The algorithm takes as input a set of vehicles within a radius of communication of each RSU (N). The graph is created with the appropriate functions described previously (represented in the algorithm as G, and the variable K, which represents the number of shortest routes that must be calculated). Thus, the RSU computes the path of each vehicle and its current coordinates (lines 4–5). Then, the RSU calculates the destination vehicle V is located within its coverage area (the last part of the vehicle ways within the RSU). After the detection of the last section of the path within the RSU, k shortest paths are calculated starting from the current point to the final vehicle In the path (line 7). From the K shortest past, a new path is selected based on the probability distribution of Boltzmann (line 8).

Algorithm of alternative routing of vehicles
input: N // a Set of vehicles in range of each RSU
1. G // the Graph is created for each RSU
2. K // Number of alternative ways
3. For v ∈ N do
   // sets of edges that make up the route of vehicles V
4. route v.getRoute();
   // returns the current coordinates of the vehicle and the distance to
   // the last segment of the path V, contained in the graph G for a given RSU
5. source ← v.getPosition();
6. lastEdge ← G.getLastEdge(route);
   // calculates K shortest paths from the starting point to the last
   // point of movement of the vehicle
7. // calculates K shortest paths from the starting point to the last
   // point of movement of the vehicle
alternativeRoutes ← G.getKShortestPaths(source; lastEdge;K);
// select a path from a set of alternative paths
newRoute ← boltzmann(alternativeRoutes;G);
if lastEdge! = route.getDestination() then
    // returns the remaining part of the route of vehicle V from the
    remaningEdges ← route.split(lastEdge);
    newRoute.add(remaningEdges);
    v.setRoute (newRoute);
  end
else
  v.setRoute(newRoute);
end

Select the path among \( K \) shortest routes is based on the calculation
of weights, defined as the sum of the weights of all segments of the path
included in the route. For load balancing across shortest routes uses a
lot of \( R_j \). For route determination uses the following rules:

\[
J = \text{set of vehicles on the map} \\
W_j = \text{set of paths of the vehicle } j \ (j \in J) \\
\gamma_j^i = \text{path } i \text{ of vehicle } j \ (j \in J) \text{ and } (\gamma_j^i \in R_j) \\
\omega_j^i = \text{weight of path } \gamma_j^i \\
M(\omega_j^i) = \text{normalized value of } \omega_j^i \left( \omega_j^i \in [0, 1] \right) \text{ defined by Equation. 1:} \\
M(\omega_j^i) = \frac{P(\gamma_j^i)}{\max\{ R(\gamma_j^i) \mid \forall i \in W_j \}} \quad (1)
\]

The \( H_j^i \) is the Boltzmann constant of vehicle \( j \) for temperature \( T \),
according to equation 2:

\[
H_j^i = \sum_{i \in R_j} e^{-\left(U(\omega_j^i)/T\right)} \quad (2)
\]

\( P_j^i \) is the probability of choosing path \( i \) of vehicle \( j \) with the parameter of temperature \( T \),
where: $T \longrightarrow \infty$ all candidate paths have the same probability of being chosen, i.e., the process approaches a uniform random distribution. When $T \rightarrow 0$, the path with less weight has a high probability of being chosen.

$F(R_j)$ is the path chosen ($F(R_j \in R_j$), the choice is made according to equation 4:

$$F(R_j) = \max \left\{ X \times Q_j^{i} (\gamma_j^i) \mid \forall \gamma_j^i \in R_j, X \approx \cup([0,1]) \right\}$$

After chosing the path, it is checked whether the last segment of the path the vehicle $V$ is in range of the roadside controller is its ultimate destination (line 9). If it is not like that, then this is the last segment of the path in the area of that RSU (line 10). The remaining stretches of road are added to the new Vehicle (lines 11–12). However, if the last segment of the path the vehicle $V$ in the range of the RSU is the ultimate destination, then it is the end of the route (line 15).

Results

This section presents the results of using the proposed system compared to the traditional approach where TC is not automated and does not interact with RSU, but the system does not have an alternate routing. In Fig. 1 shows the results obtained by the system. As can be seen from the figure, the proposed system reduces the effects, caused by congestion (increased travel time, fuel consumption and emissions $O_2$).

Figure 1 (a) shows that the proposed system reduces the travel time by 23% in comparison with the traditional approach. This confirms the efficiency of the algorithm of alternative routing. Figure 1(b) shows that the proposed system reduces fuel consumption by 9% compared to the traditional approach. This reduction is due to the fact that the proposed system detects and eliminates congestion by using alternative paths to avoid congested areas. The proposed mechanism for congestion control that contributes to more effective and continuous movement that leads to a reduction in fuel consumption. In addition, it reduces braking and acceleration, which also contributes to increased fuel consumption.
Consuming less fuel, less $O_2$ is emitted, as we can observe in figure 1 (C). The proposed system introduces a reduction of up to approximately 10% in emissions compared to the traditional approach.

**Fig. 1.** The results of modelling

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**Fig. 2.** Diagram of the results of travel time with the change of radius of communication RSU
Figure 2 shows the average travel time when changing the radius of communication in RSU (2, 4 and 6 km), the path to the shortest (3, 5 and 8) and the frequency of re-routing (180, 360 and 600 seconds). Figure 2 (a) shows the results of the K shortest paths with a fixed frequency of re-route in 180 seconds and varying the communication range of RSU from 2; 4 and 6 km. As we can observe, the higher communication range of RSUs, the larger the reduction in travel time. This is because RSUs have a greater knowledge of the area of the map and the calculated path is better from the standpoint of the time when calculated for a smaller area of the map. In addition, we can observe that the number of K shortest paths has an impact on the results obtained using the proposed system. For varying K, we can note that K = 3 has better performance compared to other values for K, thus , the result will be reduced to 9% against k = 8 and the reduction to 4% to = 5. Figure 2 (b) shows the results with frequency of re-routing 180; 360 and 600 seconds, and K shortest paths recorded in 3 (higher performance shown in the comparison figure 2 (a)) and the different range of communication from RSUs 2; 4 and 6 km. As we can observe, the frequency of re-routing has a positive effect on reducing the average travel time in the proposed system. The results show that the more the frequency of re-routing, the greater the reduction in the average time of passage for the proposed system. Figure 3 shows the fuel consumption of the proposed system by changing the communication range of RSU in (2, 4 and 6 km), K shortest paths (3, 5 and 8), and the frequency of re-routing (180; 360 and 600 seconds). Figure 3 (a) shows the results of the K shortest paths (3; 5 and 8) with the frequency of re-route fixed in 180 seconds and varying the communication range of RSUs from 2, 4 and 6 km. As we can observe, the higher communication range of the RSU, the greater the reduction in fuel consumption. This is because the RSU have a greater knowledge of the area maps and design way better when calculated for the smaller area of the map. In addition, it is possible to notice that the value of K shortest paths has an impact on the results obtained using the proposed system. The results show the difference in the amount of up to 7% for the evaluated configurations.
Consuming less fuel, the vehicle produces less emissions, which is shown in figure 3. The proposed system provides a reduction of approximately 10% of emissions compared to the traditional approach. Figure 3 (b) shows the results with frequency of rerouting 180, 360 and 600 seconds in the presence of K shortest paths (figure 3 (a)) and the different range of communication from RSU 2, 4 and 6 km (figure 3(b)). As we can observe, the range of re-routing positively contributes to reducing fuel consumption in the proposed system. The results show a difference of up to 4% in relation to the intervals of re-routing from 180 to 600 seconds. The results are presented in figures 2, 3 and 4 show that the higher the communication range of RSUs, the larger the reduction in travel time, fuel consumption and \( O_2 \) emissions. The obtained results allow to conclude that the smaller the number K shortest paths, the smaller the effect caused by congestion management. We note that large values of K also directly affect travel time, fuel consumption and \( O_2 \) emissions. Another important factor is the frequency of re-routing: the longer the interval, the more reduced travel time, fuel consumption and \( O_2 \) emissions. In this case, the vehicle does not need to make unnecessary changes of the path, thereby causing unnecessary braking and acceleration, because of high frequency of re-routing contributes to a more efficient and continuous movement, as you can see in figures 2 (b), 3 (b) and 4 (b).
The results obtained using the proposed system for $O_2$ emissions presented in figure 4, when changing the communication range of RSU (2, 4 and 6 km), in the presence of K shortest routes (3, 5 and 8) and the frequency of re-routing (180, 360 and 600 seconds). Emissions associated with fuel consumption are shown in figures 4 (a) and 4 (b) are the result of fuel consumption. Emissions associated with fuel consumption are shown in figures 4 (a) and 4 (b) are the result of fuel consumption, graphs of which are depicted in figures 3 (a) and 3 (b). The results are presented in figures 2, 3 and 4 show that the higher the communication range of RSUs, the larger the reduction in travel time, fuel consumption and $O_2$ emissions.

![Fig. 4. Diagram of the results of $O_2$ emission](image)

**Conclusion**

In this work, the algorithms for traffic management in intelligent transportation system for use in metropolitan areas. The proposed algorithms allow to reduce travel time, fuel consumption and emissions. The results show that the proposed system can significantly reduce average travel time, fuel consumption and emissions.

This work is aimed at implementing a distributed intelligent transport system for detection and control of overloads. For this purpose, RSUs are distributed throughout the city to provide full coverage of the region. In addition, each RSU is responsible for managing congestion only in the area covered by its range. Thus, the vehicle can interact with a number of shares of the way to reducing the harmful impact of an overload, such as increased travel time, fuel consumption and $O_2$ emissions.
References


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