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▶ To cite this version:

Nadia Haddadou, Abderrezak Rachedi, Yacine Ghamri-Doudane. Advanced diffusion of classified data in vehicular sensor networks. IWCMC'2011, Jul 2011, Istanbul, Turkey. pp.777-782, 10.1109/IWCMC.2011.5982645. hal-00620395

HAL Id: hal-00620395 https://hal.science/hal-00620395

Submitted on 21 Jun2012

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Advanced Diffusion of Classified Data in Vehicular Sensor Networks

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In this paper, we propose a new distributed protocol called ADCD to manage information harvesting, and distribution in Vehicular Sensor Networks (VSN). The concept of ADCD is based on the characterization of sensed information (i.e. its importance, location and time of collection) and the diffusion of this information accordingly. Furthermore, ADCD uses an adaptive broadcasting strategy to avoid the submersion of users with messages for which they have no interests. Thanks to this adaptive broadcasting strategy, ADCD limits the generated overhead avoiding network congestions as well as the long latency to deliver the harvested information, which are the main limits of other existing protocols. Moreover, it is designed to be flexible regarding the use of roadside units or not, which is not the case of other schemes in the literature. To reaches its objectives, ADCD operations are divided into three steps: (i) classification of data and the identification of their target area of diffusion, (ii) data-centric election of the set of broadcasters for avoiding the broadcasting redundancy, and (iii) iterative process for data dispatching in a targeted area. Performance evaluation shows that the ADCD protocol allows mitigating the information redundancy and its delivery with an adequate latency while making the reception of interesting data for the drivers (related to their location) more adapted. Moreover, the ADCD protocol reduces the overhead by 90% compared to the classical broadcast and an adapted version of MobEyes [4]. The ADCD overhead is kept stable whatever the vehicular density.

Index Terms—Vehicular Sensor Networks, Data Dissemination, Performance Evaluation.

I. INTRODUCTION

Vehicular sensor networks have been widely investigated during the last few years as they offer applications for road safety and driver's/passenger's comfort. They are considered as the most effective and cheapest way to avoid congestion on the roads leading to minimizing the consumption of fuel, and the time spent on the road. They also have fewer operational limitations (memory, processing, energy...) than basic Wireless Sensor Networks (WSN). However, their large amount of generated information and the frequent topology changes (density, topology, neighboring nodes...) make the existing solutions in the WSN field ineffective for this new kind of networks.

The main issues in VSN reside in the data harvesting and dissemination in such a large scale network, characterized by the frequent topology changes and network partitioning. In this paper we will focus on the data harvesting and distribution based on targeting the concerned nodes, i.e. the nodes that should receive the information and can be interested by its content according to their geo-location and the specificities (importance, location and time of collection) of the harvested data.

Our main motivation in this work comes from the fact that the congestion and the redundancy induced in order to ensure the reception of relevant messages by the concerned nodes, is an important drawback of previous works. Thus, we propose a new protocol for Advanced Diffusion of Classified Data (ADCD). ADCD targets the receptors to avoid both redundancy and network congestion. This is performed by inserting intelligence in the diffusion process. Therefore, ADCD differentiates the sensed data according to their relevance and period of validity, in order to better predict its importance for the other nodes in the network. Once the first step achieved, the collector node customize the diffusion by electing a number of broadcasters from its neighborhood according to the importance of the message, while reducing the redundancy by binding the election process to different criteria (node density, node positions). Finally, the third step consists on verify whether the limits of the broadcasting of a message, in terms of targeted broadcasting area and content validity in time, had been reached or not. By doing so, we show in this paper that the overhead/redundancy induced by ADCD is lower than the one induced by other existing schemes while the reception probability by concerned nodes is maintained high.

The remainder of this paper is organized as follows: Section II presents and summarizes the main related works. Section III focuses on the description of the proposed Advanced Diffusion Protocol of Classified Data (ADCD) protocol. Then, Section IV presents the performance evaluation of ADCD and compares it to two other existing protocols: classical broadcast and an adapted version of the MobEyes [4] protocols. Finally, Section V draws our conclusions and future works.

II. RELATED WORK

Many studies have recently been realized in the field of vehicular sensor networks. Among the main topics addressed by the research community in this domain, we find: data harvesting and data diffusion. In the following, we explain the limitation of the related literature. These limitations had motivated our current work.

A. Data Harvesting

Data harvesting is considered as the first step of most of VSN applications. The aim of data harvesting in VSNs is to

facilitate the two following steps in the VSN traditional lifecycle: i.e. data search and data dissemination.

The main data harvesting approach proposed in the literature is the DataTaxis [5] algorithm. This algorithm is inspired from the behavior of insects during their search for food. In Data Taxis, this behavior is thus adapted to vehicles allowing to effectively cover a large area and to harvest the maximum amount of data. This is realized by placing a large number of agents in concentrated information areas. This is performed by the use of stigmergy, a communication mechanism used by insects such as ants. The redundancy of collected data is decreases. This idea is difficult to adapt to insure safety applications with time constraints, due to the important mobility of VSNs. Indeed, in the case of safety applications, it is difficult to forecast where the events will happen. So, it is difficult to predict where the data harvesting agents need to be placed unless a large number of agents is used. In the latter case the data harvesting scheme will be very complex to deploy.

B. Data Diffusion

Defining a data sharing strategy depends on the type of application that uses this data as well as the required quality of service by this application. Among the existing works, two methods are frequently used. The first one consists in an immediate sending of the harvested data to support real-time applications. The inconvenience of such a method is the important redundancy that is induced and the risk of network congestion. The second method performs first data processing, and vehicles only exchange summaries at regular periods of time [6]. The goal of this method is to reduce the amount of exchanged information on the network to avoid any congestion. However, the deployment for real-time applications is impossible and an incomplete data diffusion is probable [4]. We position ourselves in the middle of these two ideas, the pre-processing made in ADCD avoid sending messages repeatedly and limits the rebroadcasting using different concepts to ensure good latency and reception ratio.

To perform a dissemination that takes into account the importance degree of each individual data, a classification between different kinds of messages, according to their quality of service requirements in terms of delay and throughput, is established. The first class concerns emergency messages, like accidents notifications. These messages are small and need to be transmitted with a high propagation velocity to insure a real time service [1] [2]. For other situations, particularly the less urgent ones, a warning message is used to catch the driver's attention. These messages concerns for instance Driver Assistance Applications. The last type is used to develop collaborative driving; where drivers share information about the density and the average speed of vehicles within the road allowing reducing the risk of traffic jams. Warning messages are less important than the emergency one but a higher rate is required for them. However, how to introduce this classification in the dissemination phase remains vague. That is why ADCD protocol classifies the information before their diffusion in the network and specifies the corresponding diffusion strategy to each one within the broadcasted message. This is made so as to guarantee that the diffused messages will be received by a maximum number of concerned nodes.

Another futuristic application of interest to VSNs is the content sharing as proposed in [7]. The technology used for that is the peer to peer technology which remains innovative for this kind of networks. Its content will include delay-sensitive and delay-tolerant data. These two concepts are also taken into account in our proposed protocol ADCD through the delay "Mode" parameter. In addition, content sharing applications among vehicles needs an efficient distribution mechanism.

The approach proposed in [3] differs from previous ones which use classic criteria for data classification. The authors develop generic network architecture to support futuristic VSN applications on focusing on the space, time and user interest related to information during its distribution. The paper assumes different kind of applications, such as safety applications, location-based services, city wide alerts and interactive services with their spatial, temporal and interest scope which define the area and time of deployment, followed by the type of targeted receiving vehicles. The realization of these applications supposes the use of roadside units which limits the deployment of such applications. Furthermore, there is no specific distribution protocol proposed in this paper. Although our protocol, ADCD, shares the same philosophy about the data's classification, it proposes a new adapted strategy for data dissemination which takes full advantage from the data classification. Another very interesting approach for data diffusion in VSNs is the Smart Mobs for Urban Monitoring with a Vehicular Sensor Network, MobEyes [4]. MobEyes ensures a proactive urban monitoring by taking advantage vehicles mobility. The exchanged messages can include 2 to 10 minutes of summaries regarding the captured data. This method, target a specific surveillance application where data are harvested, by police officers. A bloom filter is applied to retrieve the missing information tightly linked to this avoiding redundancy. However, MobEyes is proposed for specific application and its generalization is not straightforward. That is why, we propose a new protocol ADCD with more parameters which make it more general and thus it can be applied to different applications.

C. Discussion

Future application constraints for VSNs are tightened, with more requirements for network performance. As explained above, existing solutions cannot always meet their needs. A new vision has to be established to respond to the remains interrogations such as the dilemma of receiving or not information according to its pertinence and the delivery time required for its reception while avoiding overhead.

Our concept fits the identified constraints thanks to the data characterization and the corresponding enhanced diffusion.

III. ADVANCED DIFFUSION OF CLASSIFIED DATA

In this section, we describe our proposed protocol called Advanced Diffusion of Classified Date (ADCD). Its aim is to empower safety applications in vehicular sensor networks. For more robustness, the protocol is agnostic on the existence of roadside units. It just supposes that a minimum number of vehicles are equipped with sensors and communication capabilities, accepting to collaborate with other vehicles for safety information distribution. This protocol allows a regular harvesting along with smart information sharing between vehicles to avoid any risk of congestion or starvation in isolated areas.

We consider that each vehicle has to receive all emergency and local information messages available within a certain perimeter and under a corresponding time of validity. This is realized with the aim of keeping it continuously aware of current traffic conditions without flooding inappropriately other areas with information of non-interest them. Our approach is generic enough to make it possible to do that but also to distribute information on the entire network is possible if needed.

ADCD is based on three main parts: (i) harvesting and data classification, (ii) election of broadcaster, and data sharing, and (iii) iterative rebroadcast with corresponding scope.

A. Data Harvesting

A vehicle equipped with different types of sensors can collect several kinds of data. To illustrate our approach, we take the following message/data classes of interest: accident, traffic jam, landslide, risk of slipping, car crash, roadwork, vehicles number, failure of a traffic light, road's density. We consider that each piece of information depends on the region where it was collected. This way, its diffusion is only useful in its surroundings during a fixed period of time to avoid the transmission of old information.

In order to carry out this concept, we characterize information with two parameters: *class* and *mode*. A class represents the importance level of information; it is used to define the broadcasting area within the VSN. The *mode* is a value in a scale representing the period of validity of the data.

ADCD defines an interval $[\partial_{min}, \partial_{max}]$ for the classes and modes. The most urgent messages and the longest in validity are characterized as class ∂_{max} . A piece of information is represented by Cxy, where x represents the class and y the mode.

The vehicles concerned by particular information are those belonging to the targeted broadcast area and for which we advocate interest in receiving this information. In order to target these vehicles during the transmission, we attribute a diffusion perimeter, as a square centered upon with the coordinates of the collected data. The length of each square's side corresponds to the class of data. Table 1 gives an example of how ADCD can be used for a traffic safety application. In this example five classes are defined. For each class, a square is associated for each one defining its targeted diffusion area (200*200m², 300*300m², 400*400m², $600*600m^2$ and $800*800m^2$ to the first, second, third, fourth and fifth class, respectively). Hence, in this example, a road density message should not exceed the perimeter of 200*200 m².

The same principle was followed to define the period's validity corresponding to the mode of the data.

Information	Class	Mode
Accident	5	3
traffic jam	4	4
Landslide	3	5
risk of slipping	3	4
car crash	3	2
Roadwork	3	5
failure of a traffic light	2	3
density' s road	1	2
number's vehicles	1	1

Table 1: Information characteristics of shared data

B. Data Sharing

Each vehicle cooperates in the network by sending its collected data to the other vehicles. To avoid redundancy, the information is shared only if it meets the following conditions: - The vehicle has recently collected the data.

- The vehicle can resend the same data only if the previous message regarding it has reached its time validity and the information is always valid.

- None of the vehicle's neighbors already distributed this information (i.e. a message from a neighbor containing the information to share hadn't been received earlier).

If the conditions are met, the vehicle waits for a random time (backoff), bounded by the importance of the information. This mechanism aims at avoiding the simultaneous sending leading to redundancy and resource wastage.

Our concept of *class* and *mode* allows broadcasting harvested information by the vehicles according to their coordinates and date of collection. Each message will contain both values in its header.

To share information among vehicles in VSNs a basic and straightforward approach is to use classical broadcast (i.e. flooding). This will ensure the reception of the message by all the members but inevitably causes what we call a broadcast storm [10]. We decide to proceed differently. First, we choose a single hop broadcast, the message is received by all the direct neighbors of the source. Then, the continuation of the rebroadcasting procedure depends on the class and mode associated to the message (i.e. the data).

The source vehicle is also in charge of the election of the nodes among its neighbors to rebroadcast the message. The list of elected vehicles is thus inserted in the message, so that only these vehicles are authorized to rebroadcast the message. Thus allows avoiding broadcasting redundancy at the reception.



Figure 1: Sending diagram

The number of broadcaster nodes to elect depends on the information class. For the example depicted in Table 1, we can have for instance three configurations. One is for class 1 where a simple single hop broadcast is sufficient while respecting the perimeter of $200*200 \text{ m}^2$. The second, for classes 2 and 3, where the source selects three broadcaster nodes to cover the larger area covered ($300*300 \text{ m}^2$ and $400*400\text{m}^2$, respectively). This can be implemented by choosing the node with the largest number of neighbors, and then we choose two other nodes with a wide coverage located at almost 120 ° and -120 ° from the current source node. Similarly, the same method can be applied to classes 4 and 5 where four broadcaster nodes and an angle of 90° are used.

Figure 2, represents an example of use of the election algorithm for class 3 of depicted application example, in which the number of broadcaster is fixed to three. As we can see below, the first elected node is the neighbor with the highest density. Then, the two other elected nodes are chosen according to their rotation angle relatively to the previous elect and their density.



Figure 2: Example of election process for three broadcasters to elect

The election algorithm has as an input the number of broadcaster nodes we would like to have, and as an output their coordinates. It is depicted in Figure 3.

Input: elect_number; ⊖=360° / elect_number; L=0: Y=1:			
Elect_table[0]=Look_for_the_node_with_highest_densi ty (the list of neighbors sender);			
Delineate_areas_with_angle/according_to(Θ ,coordinat es_of_Elect_table[0]);			
While (Y < elect_number){ Elect_table[Y]=Look_for_the_node_with_highest_dens ity (the_list_of_area(y)); Y++; } Output: Elect_table;			
Figure 3: Election algorithm			

C. Iterative Rebroadcasting

At the reception of a message, the vehicle checks the validity of the information in time and space. This is performed according to the *mode* and *class* associated to the data in the message. If the information crossed the boundaries of its targeted area or the time validity is passed, the receiving node deletes it.

Otherwise, the receiving node consumes the information and verifies if it is elected as a broadcaster for it or not. If so, it broadcasts the message while safeguarding its characteristics, such as: class, mode, collection date and collection location coordinates.



Figure 4: Message consumption and rebroadcasting

IV. PERFORMANCE EVALUATION

In order to evaluate the performance of our protocol, ADCD, and compare it to other existing protocols, we used NS2 [9]. The parameter values for simulation are given in table 2

Number of	100, 200,	Mobility Model	SUMO [8]
Mac layers protocols	IEEE 802.11	traffic environment	urban
Bandwidth	11 Mbps	Area size	9000*9000m ²
Propagation model	Two ray ground reflexion	Maximum speed	13.9 m/s
Transmission range	250 m	Maximum simulation time	300 seconds

Table 2: Parameters values for the simulation

Using this evaluation scenario, ADCD is compared to two other diffusion algorithms. The first one is the classical broadcast where each receiving node performs one rebroadcast for message. The second one is adapted diffusion MobEyes, where a node transmits a message each 12 seconds containing a maximum of 5 received or harvested summaries. To increase the effectiveness of its sharing information, we set the k-hop to 3. We consider a message as redundant if the 5 summaries contained in it had been already received by the node, or if they exceed their interest perimeter or time validity.

For clarity, we measure the performance on a simple application, with different messages depicted in Table 1; their class and mode belong to the interval [1, 5].

In our simulation, a new message (i.e. a harvested data) is generated each 5 seconds and can be harvested (i.e. sensed) by at least one node. The events happen with certain probabilities according to their class, the probabilities of harvesting are 0.60, 0.20, 0.10, 0.05, 0.05 to the first, second, third, fourth and fifth class, respectively.

We select two performance metrics for our evaluation:

- The percentage of useful reception according to the target diffusion area.
- The overhead induced by each scheme.

The sensitivity of these two metrics regarding the vehicular density is also analyzed.

A. Effectiveness of the target diffusion

A message loses its relevance after a period of time which differs according to its content. In Figure 5, we represent the percentage of received messages by the concerned nodes as a function of time.



Figure 5: Reception rate according to the thresholds

We remark that ADCD quickly informs all concerned nodes compared to classical broadcast or adapted MobEyes. In terms of promptness, ADCD gives better results than both concurrent protocols. In addition, it reaches 100% of concerned nodes in the cases of class 1, 2 and 3 in less than 0.5 s. However, in the cases of class 4 and 5 the concerned nodes receive more than 60% of messages after 0.1s. We also note that extending the broadcasting time doesn't allow improving this percentage. The explanation of this difference among the different classes is mainly due to the size of the targeted area. The latter is more important in the case of classes 4 and 5. Classical broadcast reaches 75% reception ration on average (i.e. no differentiation between classes), which corresponds to 76%, 80%, 84%, 31% and 36% for the first, second, third, fourth and fifth class, respectively. MobEyes, however, requires more time to distribute its messages due to its design features, which also leads to a low percentage of informed vehicles.

B. Overhead

Each message transmitted to a node that is not concerned by it or after the expiry of its validity period is uninteresting and considered as overhead. Figure 6, shows the important gap between the overhead generated by the three algorithms.



We remark that ADCD provides less overhead compared to the broadcast and adapted MobEyes protocols. These results are mainly due to the target diffusion using the concept of class and period of validity according to the type of information. Moreover, ADCD allows deleting useless messages where the validity is expired and preventing their rebroadcast. In addition, the selection of the broadcaster nodes based on the election process allows to significantly reducing the redundancy. Vehicle density impact

In order to evaluate the impact of the vehicle nodes density on the ADCD performance, we varied the number of nodes from 100 to 300. The obtained results are plotted in Figure 7, which illustrates the average percentage of concerned nodes' received messages according to the nodes' density. We remark that, even if the nodes density increases the percentage of concerned nodes remains stable greater than 90% in the case of ADCD. However, in the case of both protocols (Broadcast and adapted MobEyes) with 100 nodes, the percentage of concerned nodes cannot reach 50% and 40% in the case of Broadcast and MobEyes respectively. When the nodes' density increases the Broadcast and MobEyes' concerned nodes cannot exceed 75% and 58% respectively. These results confirm the efficiency of ADCD for the target diffusion even with a small density which can limit the possibilities of rebroadcast in a large area.



Figure 7: Percentage of reception with 100, 200 and 300 nodes

Figure 8 shows the overhead according to the nodes density. We remark the linear increasing of the ADCD overhead compared to the exponential of Broadcast and MobEyes. An important overhead, causes the congestion of a network with the impossibility of inject new messages. This can have very serious consequences on safety application for the VSNs. In our case, it was impossible to push simulations with larger density because of the overhead.



Figure 8: The overhead generated by 100, 200 and 300 nodes

V. CONCLUSION AND FUTURE WORKS

The main objective of our study was to develop a new protocol to ensure a tradeoff between the reception gain of a data and the latency since it is harvested. This needs to be performed while avoiding redundancy and the implied network congestions. Our protocol called ADCD, for Advanced Diffusion of Classified Data, achieves this goal by characterizing the data to be diffused (distributed area range, validity period) and using an adapted broadcasting strategy. Our performance evaluation showed that ADCD allow reducing the overhead by up to 90% while compared to other existing protocols. This is performed while keeping the stability and reactivity of ADCD at an acceptable level for VSN applications. Our future works include the adaptation of ADCD to different traffic environments (sub-urban, rural, highways ...). Taking into account security concerns, such as confidentiality and message integrity, represent another important field of investigation.

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