Research Article

Architecture for Resource Allocation in the Internet of Vehicles for Cooperating Driving System

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Internet of Vehicles (IoV) is a complex system that consists of resource types such as vehicles, humans, and sensors. Although the Internet of Vehicles is complex, it improvises communication among vehicles on the roads. Quality of service (QoS) enabled the cooperative driving system (CDS) based on 5G technology, enabling vehicles to communicate and cooperate to improve road traffic efficiency. Due to the high vehicle density and limited resources (bandwidth) of current network infrastructure, sometimes a better channel that meets the requirements of cooperative driving is not available that causes network congestion, which directly influences the overall QoS of the CDS. To overcome this, we proposed a 5G network-based architecture for CDS that incorporates a D2D technology-based resource allocation scheme. The proposed network architecture and cooperative behavior-based scheme helps in improving QoS for CDS. We implemented our proposed scheme by incorporating the density-based scattered clustering algorithm with noise for vehicular clustering. The proposed scheme’s performance shows significant improvement in terms of throughput compared with existing D2D approaches.

1. Introduction

The fifth generation (5G) desires to be a generation that is not extended only to meet the needs of mobile operatives and consumer infrastructures and opens up new paradigms and support for societies, health, vehicles, media, manufacturing, and the transport industry [1]. 5G is seen as the underlying technology allowing self-driving cars to talk with each other. Cooperative driving plays a vital role in traffic efficiency. The new development of information and communication technologies allows auspicious cooperative driving, which has been mediated to enhance traffic efficiency and traffic safety [2] expressively.

Cooperative driving systems (CDS) are primarily based on wireless communication between vehicle-to-vehicle (V2V) and among vehicles to infrastructure (V2I). Human behavior is modelled in CDS that helps vehicles while travelling on roads [2]. The exchange of information among vehicles enables them to make cooperative decisions regarding driving [3].

IoVs prepared with 5G interfaces are complex but give additional features because of D2D communication. The base station plays a role by directing vehicles with control information to schedule the resources as per priority. In this way, base station guarantees the available resources to meet the latency and data rate required for typical vehicles [4].
Resource allocation procedures in most fourth-generation telecommunication networks require continuous monitoring of channels but do not provide required resources to every vehicle as per their needs [5]. Although 5G is quite capable of the service requirement of V2V, its use must be controlled to make it cost-effective [6]. The main motive is the efficient utilization of the radio resources as per increasing QoS demand for CDS.

High vehicle density that shares a common channel for communication QoS is not filled in CDS. So, clustering for 5G users is better for efficient resource utilization and allocation. CDS for 5G-based V2V communication is necessary to provide better resources like bandwidth to improve QoS requirements [7]. Due to extreme automobile density and limited bandwidth of current network infrastructure, sometimes a better channel that meets the requirements of cooperative driving is not available that causes network congestion. Simultaneously, the ever-changing topologies of communicating vehicles are a big hurdle while allocating appropriate resources at the right time and place. These things can directly impact the overall QoS performance of the CDS. For better V2V communication to improve service requirements, it is necessary to develop a network infrastructure that improves the QoS of cooperative driving with a greater throughput.

This paper proposed a resource allocation scheme for a cooperative driving system for IoVs. First, an architecture of the cooperative driving system (CDS) is proposed based on 5G networks. Then, a clustering algorithm is proposed for resource allocation. Consequently, proposed architecture entities, how these entities work, and their functionalities for our proposed solution are outlined. Secondly, a clustering algorithm for IoVs to allocate resources for better transmission in CDS is formulated. The proposed scheme is implemented, and the results are validated against the existing scheme. Section 2 discusses the related work, and Section 3 presents the proposed CDS architecture in detail. Section 4 presents the proposed solution’s implementation, and the performance evaluation is presented in Section 5. Finally, Section 6 presents the conclusion of the paper.

2. Related Work

The automation of vehicles is increasing over time. Simultaneously, technological advancement in telecom networks increases the number of interactions among vehicles and between vehicles and other road users. As a result, the communication system’s reliability and capability become significant for improved performance of autonomous driving and road safety [8].

Three usual multivehicle cooperative scenarios, formation control, convoy-driving, and connection management are studied in [9]. A communication resource allocation solution was brought forth for these three scenarios. The concentration was on just one convoy when it crosses the intersection. A self-organized resource allocation scheme for V2V communication is suggested in [10]. To lessen the overhead dynamic, zone formation-based resource algorithm is proposed. This scenario’s problem is that multiple vehicle zones are connected with only one roadside units (RSUs). It causes transmission overhead for the base station and reduces the channel state quality. The problem here is that every V2I links bonds spectrum with multiple V2V associations. The BS has only access to gentle decreasing certificate status information (CSI) of vehicular associations, except those who are dismissing at the BS [11, 12].

A cooperative allocation strategy for bandwidth against each link is presented in [13] to achieve better communication performance among vehicles on the move. CAVDO and MA-DTR-based clustering approaches are used to organize network organization and resource allocation [14, 15]. It performs fine in clustering by dropping the number of clusters when there is very high vehicle density. However, if clustering is not appropriately done, resource optimization is mislaid.

Höhytiä et al. [16] analyzed the Third-Generation Partnership Program (3GPP) standards for power consumption management regarding 5G. Release 14 presents two types of innovative communication modes that are specified for V2V communications. In mode three, the cellular networks choose and achieve the cars’ radio resource for their direct V2V communication. In mode 4, vehicle automatically chooses the radio resources for their direct V2V communications. Mode 4 could function without cellular coverage. It has a different scattered scheduling system for vehicles that choose their radio resources and provides functioning for distributed congestion control [17, 18].

A comprehensive analysis of 5G [19] related to road traffic management is presented with different problems that occur during the resource allocation. Two main challenges for D2D communication are first interference for D2D links. Secondly, it improves QoS requirements using RRM and RRM scheduling algorithms in the neural Q-learning-based approach [20]. An ACORA [21] scheme (ant colony optimization-based resource allocation algorithm) is proposed, which promises the efficient use of accessible resources helping users by conferring their QoS parameters. A swarm intelligence algorithm is adopted to reduce computational complexity while understanding suitable performance. RRM [22] for V2X based on D2D communication is proposed which promises reliability and latency while allocating resources to vehicles.

A clustering scheme based on cooperation among vehicles is presented in [23], which is dependent on GPS. Performance is a lack in case of GPS signal interruptions. Another LTE-dependent clustering scheme is presented in [24]. It relies on the LTE network for location-based information required by the vehicles while travelling on the road, whenever the need to make driving-related decisions. There is no localized mechanism that helps vehicles in the absence of GPS signals. A Fermi learning model based on the prospect theory is developed in [25], which calculates the distribution density and the distance between the taxis. The authors in [26] proposed an optimal beacon transmission rate allocation for cluster control. An adoptive rate is defined for efficient intracluster communication that is useful in avoiding cluster congestion.
A 5G network-based control framework [27] is proposed and deployed in Sweden that minimizes communication delay in managing intersections. This framework works cooperatively but no focus on radio resources management. A comprehensive review presented in [28] radio resource management schemes proposed for LTE/LTE-A networks is still relevant. However, these schemes should be adaptive and futuristic that meets the user’s demands.

Most of the research done in IoVs for resource allocation that we find in literature used the Knap Sank algorithm [29] and the greedy algorithm. There is a large “network load” during communication that causes “traffic congestion,” thus the reason for trouble in large data handling. “Dynamic topology” of both communicating vehicles may reduce network performance [3]. We need such a system that allows vehicles to communicate in dense areas with fewer transmission delays and a better resource allocation system.

3. CDS Architecture

A resource allocation architecture and its entities for CDS are proposed in this section, as presented in Figure 1. Symbols and abbreviations used in our proposed design are described in Table 1.

A CDS-based V2V communication helps vehicles to cooperate and communicate with each other on the road to improve traffic efficiency. QoS provision is important for efficient communication. It is necessary to provide proper resources like bandwidth to improve QoS for CDS. To enhance CDS for V2V communication, we proposed a clustering algorithm that is DBSCAN for resource allocation in IoVs.

This system’s constraints are that devices used in communication should perform in a real-time scenario. If the system does not perform real time, the system working does not handle it. Our system should meet the computational time. The details of the proposed architecture and its algorithm are described below in Sections 3.1 to 3.7.

3.1. Channel State and Mobility Management Entity. In 5G networks, the mobility management entity (MME) covers 5G user location servers connected with the ENodeB access interface. It collects the vehicle’s location data. It checks how many vehicles are moving on the road, how many resources are available for each vehicle, and how much additional is needed. The channel state entity, along with the mobility management entity, collects channel state quality information of each ENodeB and exchanges channel state quality information with TES. It updates the information on available resources and the number of resources needed for each vehicle using the banker’s algorithm. It checks the resource availability (in our scenario bandwidth), how much resource is needed, how many resources available on each ENodeB, and how much each ENodeB request. The working of channel state entity is as follows:

\[ S = \{S_0, S_1, S_2, \ldots, S_{N-1}\}, \]

\[ R_P = \sum_{R_P} + S \]

\[ R_B = \sum_{R_B} + S. \]

The allocated resource in our scenario is bandwidth represented by \( n \). The range of resources is found by a comparison value in which bandwidth is less than the total resource block. For that total resource, blocks are subtracted by 1 and then divided by 2:

\[ \text{bandwidth} = n, \]

\[ \text{range} = n \leqslant \left[ \frac{B_{R g-1}}{2} \right]. \]

To allocate resources and to calculate the availability and need of resources, we formulated an equation in which \( \log_2 \) is taken. Resource allocation is represented by \( R_{a} \), and resource availability is denoted by \( R_{av} \):

\[ R_{a} = \left[ \log_2 \left( n^{B_{R g-1}} + \frac{1}{2} \right) \right], \]

\[ R_{av} = \log_2 \left[ n^{R_{PB}} + 1 \right]. \]

To find a need (\( N \)) of resources, available resources are subtracted from allocated resources. So, from equations (8) and (9), we get
Controlled and shared resources define the resources in the resource pool. Controlled are the available resources, and the shared resources are the allocated resource. When the resource allocation process is completed, this information is sent to TES for further processing.

3.2. Traffic Efficiency Server (TES). TES receives vehicle location and updates vehicle location messages received from vehicles with a mobility management entity. It stores and updates data exchanged from channel state entity. Based on the channel state information exchanged from the channel state entity, TES allocates resources to the ENodeB for V2V communication [30]. Due to the CDS' strict latency and reliability requirements for V2V communication, a traffic efficiency server must manage and accept the network promptly [31, 32], so TES and ENodeB must establish their connection in real time without delay.

3.3. Macrobase Station and ENodeB. Macrobase stations are fundamental elements in any heterogeneous network wireless infrastructure to provide coverage and support capacity. A small cell that is ENodeB is a major infrastructure part of the 5G network architecture. In our scenario, macro-BS, ENodeB, and TES are interconnected. Resources like bandwidth are assigned to different ENodeBs. In our scenario, the resource is bandwidth. These base stations (ENodeB) handover resources to vehicles. If the requested resource is not available in the base station, it can currently request TES for a particular service. TES server requests a macrobase station. From the macro-BS, these resources are handed over to ENodeB using LTE-A.

3.4. Clustering. Vehicular clustering is a technique of grouping nodes in several clusters. Each cluster has two components: CH and cluster member (CM) [33]. In V2V communication, CH’s role is to lead the group, and all the participants in the group are its members called CM [34]. In our proposed architecture, the closest to ENodeB is the clustering head, and all neighbor nodes are its members. These are active clusters that are involved in V2V communication.

3.5. Zone Formation. To cater to interference, each cluster forms its zone for V2V communication [10]. For D2D communication, we set the radius, which is the communication zone. The purpose of creating a zone is that each CM communicates within its range. When the nodes (vehicles) are in the radius range, the data transmission is high. As soon as they go away from the base station radius, their throughput decreases. Each vehicle communicates in its zone. When a vehicle moves from one zone to another, it makes its new cluster.

A problem arises when many clusters are in communication within their radius. Some intermediate nodes cause interruption because they are connected with both ENodeB.

![Proposed architecture for resource allocation in CDS.](image)

**Table 1:** Symbols used in proposed design and implementation.

| Symbol | Description |
|--------|-------------|
| CDS    | Cooperative driving system |
| BOLTZ  | Boltzmann constant |
| SNR    | Signal to noise ratio |
| BW     | Bandwidth |
| R_B    | Resource block |
| R_P    | Resource pool |
| N_RB   | Number of resource blocks |
| N_USER | Total number of users |
| N_SC   | Number of subscribers |
| R_a    | Resource allocation |
| R_av   | Resource availability |
| S      | Number of subframes |
| N      | Need of resources |
| n      | One user value |

\[
N = R_{av} - R_{al} \left( N = \log_2 \left[ n^{R_B - R_P + 1} \right] - \left[ \log_2 \left( n^{R_B - 1} + \frac{1}{2} \right) \right] \right).
\]
A value is set for each zone so that each node communicates within its specific range of ENodeB. It helps to avoid overlapping of the node between the clusters. In Figure 2, the vehicle in the middle of two cluster cause overlapping between the two zones and cause signal interruption for both clusters results in the wastage of the resources during communication.

3.6. Density-Based Spatial Clustering Algorithm for Resource Allocation. In V2V communication, mostly k means clustering technique is used in which the central node is considered as clustering head. We use the DBSCAN algorithm [35] for clustering. It consists of epsilon (Eps) that tells how close the points should be to each other in clusters. Minimum points (Min points) show the minimum number of points to form a dense region. The points which are near the border are included as noise. In our proposed solution, first, we create a resource pool for D2D communication, in which resources are calculated. A subframe is created along with the allotted bandwidth.

For the clustering algorithm, we made a dataset for vehicles. The allotted values in the resource pool are given along with throughput. This is real data gathered from Monte Carlo simulation [36], which is used to get data for D2D communication in Matlab [37]. The clustering algorithm is applied to a dataset that contains labelled data. For the clustering algorithm, we made a dataset for vehicles. The allotted values in the resource pool are given along with throughput. This is real data gathered from Monte Carlo simulation [36], which is used to get data for D2D communication in Matlab [37]. The clustering algorithm is applied to a dataset that contains labelled data. When the cluster formation is done, the resource allocation process is started. Equation (11) shows the DBSCAN algorithm’s working for distance calculation:

\[
\text{dist}(p, q) = \sqrt{\sum_{i=1}^{n} (p_i - q_i)^2}.
\]  

3.7. Pseudocode of Proposed Algorithm. For mode selection and transmission point for resource allocation for CDS is given in (Algorithm 1).

4. Implementation Scenario

In our implementation, the first resource pool is created, and then, the DBSCAN algorithm is implemented on the dataset. This dataset is created by taking the data from the D2D simulator. D2D throughput is calculated using Monte Carlo simulation. As per our system requirements, we design system architecture, two nodes, two base stations with TES, and MME-based V2V communication. Vehicle clustering using the DBSCAN algorithm is incorporated with system architecture.

We create a resource pool in which subframes and resource blocks are created. To allocate resources, equations (1), (2), and (3) are defined as presented in Section 3.1. Figure 3 illustrates the V2V resource pool for our resource allocation. Figure 3 shows that the V2V -controlled resource pool is dark blue, and the shared resource pool is yellow. The light blue color shows synchronization time. The brown color shows the baseband and the green color for bandwidth. After the simulation is done in Matlab, the results are compared with the D2D communication model [38].

4.1. Implementing the DBSCAN Algorithm. After creating a resource pool, we create a dataset for resource allocation and implement the DBSCAN algorithm for that dataset, as shown in Figure 4. The dataset is used for the clustering of data. It is a well-known clustering algorithm used in machine learning and data mining. It consists of Eps that tells how close the points should be to each other in clusters.

Min points show the minimum number of points to form a dense region. The points which are near the border are included as noise. This is a labelled dataset. Each label shows the functionality of every point.

4.2. Throughput Calculation. Equation (12) shows \(N_{RB}\) is the number of resource blocks and \(N_{USER}\) is the total number of users, where \(n\) is used for one user value. Equation (13) is used to calculate the total number of subscribed users (\(N_{SC}\)) per resource blocks:

\[
n_{user_{RB}} = \left( \frac{N_{RB}}{n} \right).
\]

\[
n_{user_{SC}} = n_{user_{RB}} * \frac{N_{SC}}{N_{RB}}
\]

\[
n_{RB_{-user}} = \frac{N_{RB}}{N_{USER}}
\]

Equation (14) calculates throughput, and the first SNR is calculated. In the following equation, BW is the bandwidth, and BOLTZ is a Boltzmann constant that is \((1.3806488 e^{-23})\). It is used when we calculate SNR with bandwidth:

\[
\text{SNR} = \left( \text{interference array} + \text{BOLTZ} * \frac{\text{BW}}{N_{RB}} \right)
\]

To calculate the D2D throughput, sum of the log is multiplied with the SNR value and then add one in it, as shown in the following equation:

\[
\text{D2D throughput} = \text{sum(log2 } (1 + \text{D2D SNR}))
\]

5. Performance Evaluation

We measure the performance of the proposed CDS against the D2D communication [38] scheme for vehicular networks. Throughput is measured against the number of RBs, various radius sizes, and the number of users. In the end, SNR against throughput is also presented for performance check. The parameter setting for the calculation of throughput is shown in Table 2.

Figure 5 shows the average throughput against the allocated number of radio RB. At the lower number of RBs, there is not much difference in the throughput of the existing D2D scheme and proposed scheme. However, this difference increases with the increase in the RBs at RB of
The throughput of the proposed scheme is almost double the existing D2D schemes. One of the reasons is that, in the D2D scheme, an increase in the network’s communication area involves more latency or packet drop.

Average throughput is calculated to find the QoS for V2V communication. For the D2D communication scheme, average throughput is calculated for several resource blocks, for the radius size, and the number of users. From the results, it is shown that our proposed algorithm has better results as compared to the existing D2D scheme. The average throughput increases with an increase in the RBs. The increase in the proposed scheme’s throughput is higher than the existing D2D schemes. This is because increase in RBs increases the chances for the mobile node to get more required resources on time.
Figure 6 shows an average throughput against various network radius sizes for both the D2D scheme and the proposed scheme. The average throughput decreases as the radius size is increasing. An increase in the signal range increases the cluster size as more vehicles are added, affecting overall communication among vehicles. By increasing radius and size, there is a negative impact on both protocols’ performance. However, our proposed protocol’s performance is still greater at each of the radius size values. The decrease in performance is because of the increase in transmission failures and an end-to-end delay. This is also evident in Figure 7, which shows an increase in the number of users decreases the throughput as more radius size allows more vehicles to become part of the communication. Thus,
as the network’s scalability is concerned, the proposed scheme has shown better performance. The average throughput remains almost double of the D2D scheme at radius sizes of 30, 40, and 50 meters.

Figure 7 shows the average throughput against the number of users. Our proposed algorithm’s average throughput is much higher than a simple D2D-based communication scheme. For the D2D communication scheme, when the number of users increases, the average throughput decreases because all users want to get all resources. In our case, the throughput is good, which provides better QoS for CDS.

SNR directly affects the performance of throughput. When SNR’s value is high, it means that the signal’s strength is also more robust relative to the noise levels; it allows higher data rates and fewer retransmissions. The lower the SNR value, the lower is the data rate, which decreases throughput.

Figure 8 shows the SNR concerning the throughput of our proposed schemes. In our scenario, the average SNR is high, which increases throughput; thus, the QoS for the CDS increases, which helps traffic efficiency applications. The increase in throughput against various metrics shows that our proposed architecture for resource allocation for CDS performed well against the existing D2D scheme. Throughput increases because of the efficient resource allocation that positively impacts data rate and reduces overall transmission delay.

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**Table 2: Parameters set to calculate throughput.**

| Parameters                        | Values               |
|----------------------------------|----------------------|
| Frequency                        | 5 (% GHz)           |
| No. of resources (n_rb)          | 20, 40, 60, 80, 100 |
| N_user                           | 25                   |
| Path loss parameter              | 3.5                  |
| Radius                           | 25 m                 |
| No. of antennas                  | 2                    |
| Transmission mode                | FDD                  |
| No. of the base station          | 2                    |
| CSI                              | On                   |

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![Graph showing average throughput against allocated no. of RBs.](image-url)
Figure 6: Average throughput against various radius size.

Figure 7: Average throughput against the number of users.
6. Conclusion

IoVs is a trend of the IoTs, used for communication among vehicles. Better V2V communication that meets the QoS requirements is essential for efficient road traffic management. A QoS CDS based on V2V communication helps vehicles communicate and cooperate reasonably to improve traffic efficiency over the road as there is a significant increase of more than 50 percent (on average) in the throughput of the proposed CDS than the existing D2D schemes. The results adequately appropriate the main motive of the efficient use of radio resources to meet CDS’s QoS requirements. This technique may help in the future to improve QoS for CDS by better allocation of resources.

Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

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