

## “Architectural Dynamics in Internet of Vehicles”

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**Abstract** This abstract explores and compares key architectural frameworks in the context of the Internet of Vehicles (IoV). The Three-Layer Architecture establishes a foundational structure with distinct layers for perception, network, and application, ensuring scalability in IoV systems. Context-Based Architecture enhances decision-making by incorporating contextual information, while Lambda Architecture combines batch and real-time processing for diverse data needs. The integration of Cloudlets into IoV introduces a decentralized edge computing paradigm, minimizing latency and enhancing data processing efficiency. IoV's future lies in the harmonious coexistence of various architectures, which will support a networked ecosystem that prioritizes effectiveness, intelligence, and smooth user experiences.

### 1. Introduction

The seamless integration of automobiles into the huge linked web of the internet, known as the Internet of vehicles (IoV), forecasts a transformative era in the automotive sector. This innovative idea goes beyond conventional ideas of transportation and ushers in a new era of real-time communication between vehicles, infrastructure, and roadways to improve driving safety, efficiency, and overall enjoyment. Fundamentally, IoV uses cutting-edge communication

technology to build a dynamic network that allows cars to communicate with one another, with traffic signals, and even with their surroundings. In this interconnected ecosystem, vehicles become more than mere modes of transportation; they evolve into intelligent entities capable of exchanging crucial data, such as traffic conditions, road hazards, and vehicle performance metrics. Through the fusion of sensors, artificial intelligence, and wireless connectivity, the IoV promises to revolutionize the way we navigate our world. This paradigm shift holds the potential to significantly reduce traffic accidents, optimize traffic flow, and minimize environmental impact [1]. These days, there is also a noticeable trend in Intelligent Transportation Systems (ITS) applications toward higher data dimensions, which are defined by the "5 Vs of Big Data." It becomes essential to use big data analytics in order to fully utilize such data. The hub connecting ITS devices to cloud computing centers, where massive data processing occurs, is the Internet of Vehicles (IoV). However, the transfer of large amounts of data from geographically separated devices poses difficulties, resulting in network overhead, bottlenecks, and a high use of network resources. Furthermore, the traditional centralized method used to handle massive data in ITS leads to increased latency, which is not suitable with the time-sensitive nature of ITS applications. This sensitivity to

delay is especially important in situations where immediate reactions are essential. Furthermore, the traditional centralized method used to handle massive data in ITS leads to increased latency, which is not suitable with the time-sensitive nature of ITS applications. Fog computing is considered a promising technology for real-time big data analytics. Fog computing in big data analytics is a decentralized computing paradigm that extends cloud computing capabilities to the edge of the network, closer to the data source or "on the edge" devices. While traditional cloud computing relies on centralized data centers to process and analyze data, fog computing distributes these tasks to local devices, such as routers, gateways, and edge servers. This approach is particularly valuable in scenarios where real-time data processing, low latency, and bandwidth efficiency are crucial. Fog computing helps address challenges associated with transmitting large volumes of data to centralized cloud servers. Instead of sending all data to a distant cloud for processing, fog computing enables local analysis and decision-making at or near the data source. This not only reduces the latency in data processing but also alleviates network congestion and minimizes the need for massive data transfers [2].

The concept of Vehicular Ad Hoc Networks (VANETs) was introduced where vehicles equipped with wireless communication devices can form networks. Vehicular Ad Hoc Networks, or VANETs, support communication between automobiles and infrastructure, resulting in improved road safety, and traffic efficiency. VANETs, which function on the ad hoc networking principles, enable cars to create dynamic, self-organizing networks without the need for a fixed infrastructure. With sensors and onboard electronics, any car may talk with other cars in the

vicinity to exchange vital information about the state of the road, speed, and location. Cellular networks and dedicated short range communication (DSRC) make this communication possible. Applications supported by VANETs include real-time traffic updates, traffic signal optimization, and collision avoidance. Information exchanged is guaranteed to be authentic and has integrity due to security measures like cryptographic protocols. In this setting of dynamic connectivity, the network architecture changes as vehicles move, posing both opportunities and challenges for effective data distribution and network administration [3] [4]. The communication architecture of Vehicular Ad Hoc Networks (VANETs) can be categorized into three primary types: Wireless Access in Vehicular Environments (WAVE) based Wi-Fi, ad hoc networks, and hybrid architectures. In the WAVE-based Wi-Fi-driven architecture, Roadside Units (RSUs) are strategically placed alongside roads to function as wireless access points. These RSUs provide communication coverage to vehicles within their designated coverage area, enabling seamless connectivity and data exchange among the vehicles and the infrastructure. The ad hoc architecture operates differently by leveraging groups of on-road vehicles to form ad hoc networks using WAVE technology. These networks are characterized by their ability to perform operations independently, without relying on any fixed infrastructure support. The vehicles communicate directly with each other, establishing a dynamic and flexible network that can adapt to the changing positions and conditions of the vehicles on the road. The hybrid architecture combines elements of both cellular and ad hoc architectures, also utilizing WAVE technology. In this model, the operations of cellular networks and ad hoc networks are integrated and work in collaboration. This hybrid

approach aims to leverage the strengths of both architectures, providing robust and versatile communication solutions for vehicles in various scenarios. Each of these architectures offers distinct advantages and is suitable for different aspects of vehicular communication, contributing to the overall effectiveness and efficiency of VANETs [14].

A network model of the Internet of Vehicles (IoV) is proposed by identifying its major network elements. These building blocks of IoV, in terms of network elements, more effectively convey the meaning and functionalities of IoV as a comprehensive heterogeneous network. The first element, the 'cloud,' represents the brain of IoV. It provides a range of services related to intelligent computing and processing, which are offered as primary cloud services. These services are hosted on a cloud platform supported by cloud infrastructure. The cloud-based intelligent computing and processing services are accessed via a reliable 'connection,' the second element of IoV. A variety of wireless access technologies can be employed to establish this connection. The different types of vehicular communications within IoV represent different connections, owing to the use of diverse wireless access technologies. These various connections are utilized by smart 'client' applications, which constitute the third element of IoV. Each client application has specific service requirements that may differ from other clients. These service requirements are defined by the characteristics of the wireless access technology used. Therefore, client applications prioritize and prefer certain wireless access technologies over others [22].

## 2. IOV System Architectures

Three layer architecture: A comprehensive three-layer architecture has been identified to delineate the intricate interactions among various technologies within the Internet of Vehicles (IoV) environment. The first layer is focused on the vehicle's array of sensors. These sensors play a crucial role in meticulously gathering environmental data and identifying key events such as vehicular scenarios, driving patterns, and surrounding conditions. By continuously monitoring these factors, the sensors ensure that the vehicle remains aware of its environment, contributing to safer and more efficient driving experiences. The second layer is dedicated to communication, supporting an array of wireless interaction modes. This layer facilitates Vehicle-to-Pedestrian (V2P), Vehicle-to-Infrastructure (V2I), Vehicle-to-Vehicle (V2V), and Vehicle-to-Sensor (V2S) communications, ensuring that the vehicle remains connected with its surroundings. The communication layer integrates multiple networks, including IEEE 802.15.4, IEEE 802.11p, GSM, LTE, Wi-Fi, and Bluetooth, to provide seamless and reliable connectivity. This robust communication infrastructure is essential for maintaining continuous interaction between the vehicle and various elements within its environment. The third layer encompasses the intelligence resources of the IoV. This layer is responsible for decision-making, especially in critical situations such as hazardous road conditions and accidents. It utilizes statistical tools and provides storage and processing capabilities for the extensive data collected. By analyzing this big data, the intelligence layer can make informed decisions that enhance vehicle safety and performance. The integration of these resources ensures that the vehicle can respond appropriately to a wide range of scenarios, ultimately contributing to a more intelligent and

responsive IoV system. Together, these three layers create a cohesive architecture that enables the seamless operation of IoV, from data collection and communication to intelligent decision-making. This multi-layered approach is pivotal in advancing the capabilities of IoV, fostering a connected ecosystem that prioritizes safety, efficiency, and enhanced user experiences [5].

Context based architecture: An IoV paradigm that is context-aware and uses high-level contextual information to bring intelligence to the design in order to improve communication performance. Specifically, the researcher examined the effects of various contextual information on V2V communication performance using big data analytics on large-scale IoV communication traces gathered from a comprehensive experiment carried out in Shanghai. The NLoS link condition is a key contextual information type that has a considerable impact on V2V connection performance among many other types. Three major communication program designs with context awareness of V2V link are: (A) Smart medium resource allocation with context awareness in V2V (Vehicle-to-Vehicle) communication optimizes data transmission. By dynamically adapting to link conditions, such as vehicle proximity and environmental factors, it allocates resources judiciously. This enhances communication reliability and efficiency, mitigating interference and congestion. Context-awareness ensures that the allocation aligns with real-time conditions, fostering a responsive and adaptive V2V network that thrives on intelligent resource management for seamless and reliable communication between vehicles.

(B) Efficient routing in Vehicle-to-Vehicle (V2V) communication relies on context-awareness of link

conditions. Through real-time assessment of factors like vehicle speed, location, and traffic density, the system dynamically adapts routing strategies. This context-aware approach optimizes data transmission paths, minimizing latency and ensuring reliable communication in varying driving scenarios. By considering the dynamic conditions of the V2V links, the routing establishment becomes more agile, enhancing the overall efficiency of communication in connected vehicular networks.

(C) Reliable safety message broadcasting in V2V (Vehicle-to-Vehicle) communication employs context-awareness of link conditions. Vehicles assess factors like signal strength, interference, and network congestion to dynamically adapt message transmission. This context-aware approach optimizes communication reliability, ensuring critical safety information, such as collision warnings, reaches neighboring vehicles accurately. By adjusting broadcast strategies based on real-time link conditions, V2V systems enhance overall safety and responsiveness on the road [6] [7].

Lambda architecture: The generic Lambda Architecture for real-time big data processing is designed to maximize efficiency and provide comprehensive data insights by utilizing a dual-processing mechanism. This sophisticated architecture ensures that incoming data is handled through two distinct paths simultaneously. The batch layer is responsible for long-term storage and historical analysis. This layer collects and processes large volumes of data over time, allowing for thorough and detailed analysis of historical data trends. By doing so, it supports deep analytical queries and complex computations that require access to extensive data sets. The speed layer addresses the need for real-time

processing and immediate insights. This layer is designed to handle data as it arrives, processing it quickly to provide up-to-the-minute information. This real-time capability is crucial for applications that require instant feedback and timely decision-making, such as monitoring systems or dynamic content delivery. The results from both the batch and speed layers are then merged in the serving layer. This layer acts as a bridge, integrating historical data analysis from the batch layer with the real-time insights from the speed layer. The serving layer ensures that users have access to a comprehensive and current view of the data, combining the strengths of both processing approaches. This hybrid approach of the Lambda Architecture optimizes the handling of both historical and real-time data. By doing so, it offers a robust and efficient solution for dynamic big data processing needs. This architecture's ability to balance long-term data storage and immediate data processing makes it an ideal choice for systems that require both extensive historical analysis and rapid real-time insights [8].

**Cloudlets:** Cloudlets serve as localized mini-clouds that are strategically positioned closer to the edge of the network. These compact cloud platforms play a crucial role in processing the vast amounts of data generated by IoV devices more swiftly. By being situated near the source of data generation, cloudlets significantly reduce the latency typically associated with sending data to distant cloud servers for processing. This proximity enables real-time responsiveness, which is essential for the effective functioning of IoV applications. One of the key advantages of cloudlets is their ability to offload computation tasks from centralized cloud servers to local processing units. This offloading mechanism ensures that vehicles within the IoV ecosystem can

access critical services much faster, which is paramount for applications that demand immediate data processing and response. For instance, in scenarios involving traffic management, cloudlets can process real-time traffic data quickly, enabling dynamic adjustments to traffic signals and providing timely updates to drivers. Similarly, in the realm of safety protocols, cloudlets can analyze and respond to data from sensors and communication systems almost instantaneously, thereby enhancing the overall safety and reliability of vehicular operations. The implementation of cloudlets in IoV not only improves efficiency but also ensures seamless connectivity across various applications. By handling data processing tasks locally, cloudlets minimize the dependency on remote cloud infrastructures, which can be prone to delays and connectivity issues. This localized approach to data processing ensures that the performance of IoV applications, such as traffic management systems and safety protocols, is optimized, leading to more efficient and reliable operations [9].

### 3. Limitations

Cloud-based big data system architectures encounter limitations in meeting latency-sensitive requirements for numerous Intelligent Transportation System (ITS) applications. The reliance on centralized cloud servers introduces unavoidable delays in data processing and transmission. For time-critical ITS scenarios, such as real-time traffic monitoring or collision avoidance, this inherent latency becomes a hindrance. The geographic distribution of data sources exacerbates the challenge, causing network congestion and bottlenecks during data transfer. Consequently, the centralized nature of cloud-based architectures impedes the timely processing of information,

undermining the performance and responsiveness essential for latency-sensitive ITS applications, which demand instantaneous decision-making and real-time insights [10].

Cloudlets need to be more flexible than typical cloud data centers since user mobility is dynamic. As users migrate, virtual machine (VM) handoff technology becomes crucial to enable smooth offloaded service transfer from one cloud to another. When cloudlets are used as geographically dispersed small data centers in the context of the Internet of Vehicles (IoV), users must find, pick, and associate with the best cloudlet among several possibilities. These difficulties are made worse by the quick movement of transportation entities, which makes cloudlet utilization in Internet of Vehicles (IoV) contexts difficult and resource-intensive. Furthermore, there are significant maintenance expenses associated with distributing cloudlets over a large geographic area, and underutilized facilities may arise at specific times and locations due to the erratic demand for cloudlet services based on vehicle traffic patterns [11].

IoV faces significant security challenges, including the risk of unauthorized access, data breaches, and cyber-attacks. Safeguarding the communication channels and ensuring the integrity of data are crucial concerns. The extensive collection and sharing of personal and sensitive data within IoV raise privacy concerns. Protecting user information and ensuring proper data anonymization are essential but challenging tasks. Standardization and interoperability among diverse IoV components and systems are critical issues. Achieving seamless communication and collaboration between different manufacturers' vehicles and infrastructure remains a challenge. The abundance of data generated by IoV devices can lead to information

overload. Effectively managing, processing, and extracting meaningful insights from this vast volume of data is a significant challenge [12].

The framework of Vehicular Ad Hoc Networks (VANETs) presents several limitations that hinder the provision of global and sustainable services by Intelligent Transportation Systems (ITS) applications. One primary issue arises from the pure ad hoc network architecture. When a vehicle disconnects from an ad hoc network, it loses access to network services despite still being on the road. This disconnection is due to the inability of the vehicle to collaborate with other reachable networks, thus interrupting the service continuity. In the current VANETs framework, guaranteeing Internet connectivity is problematic. This lack of reliable connectivity directly impacts the availability of commercial applications for drivers and passengers, as these applications depend on consistent Internet access. Without this reliability, the commercial potential of VANETs remains untapped [14]. Despite the significant increase in the use of personal devices in daily life, these devices often fail to communicate with VANETs. This communication breakdown is due to the incompatible network architecture that separates personal devices from VANET systems, preventing seamless integration and interaction. Another challenge in VANETs is the inability to make intelligent decisions based on big data mining computations. This limitation stems from the constraints on computing and storage capabilities within vehicles, coupled with the unavailability of cloud computing services. Consequently, the potential benefits of data-driven decision-making are not realized in the current VANET architecture [15]. The accuracy of ITS application services in VANETs is also significantly lower than desired. This lower

accuracy is a substantial risk factor, as reliable service is crucial for enhancing driving experiences. The primary reason for this inaccuracy is that VANETs rely on local knowledge of traffic environments, which can be limited and insufficient for optimal decision-making [16].

Accurately localizing vehicles within vehicular communication environments presents significant challenges, primarily due to the stringent accuracy requirements that far exceed those offered by existing GPS-based systems. In these environments, precise vehicle positioning is crucial, and GPS-based localization, with its current accuracy level, falls short of meeting these demands. Firstly, while GPS-based localization typically offers an accuracy of about 5 meters, the vehicular communication environments require a much finer accuracy of around 50 centimeters. This discrepancy highlights a substantial gap between the current capabilities of GPS technology and the precision needed for effective vehicular communication [19]. Secondly, GPS-based localization systems do not account for the speed of the objects they are tracking. In contrast, the speed of vehicles is a critical factor in vehicular communication environments. The dynamic nature of vehicle movement necessitates a localization system that can accurately track and respond to rapid changes in position and speed to ensure safety and efficiency [20]. Lastly, the quality of GPS signals can significantly deteriorate or even become entirely unavailable in dense urban environments. Tall buildings and other urban structures can obstruct GPS signals, leading to inaccuracies or loss of signal. This is a considerable limitation in urban settings where precise vehicle localization is paramount for navigation and communication [21]. To meet the stringent accuracy

requirements in vehicular communication environments, these three critical issues: accuracy, speed consideration, and signal quality must be thoroughly addressed and resolved.

#### 4. Conclusion and Future Scope

As the Internet of Vehicles (IoV) continues to evolve, the choice of architecture is heavily influenced by the specific use cases and objectives at hand. The Three-Layer and Context-Based Architectures offer solid solutions for particular scenarios, providing reliability and efficiency where needed. Meanwhile, Lambda Architecture's flexibility makes it suitable for a wide range of data processing needs, demonstrating its adaptability in various contexts. The rise of Cloudlets highlights the crucial role of edge computing within IoV, presenting a distributed framework that caters to the increasing demand for applications requiring low latency and high performance. This innovation underscores the importance of integrating edge computing solutions to enhance the overall functionality of IoV systems. Looking ahead, the future of IoV hinges on the seamless integration of these diverse architectures. Achieving this integration will cultivate a connected ecosystem that emphasizes efficiency, intelligence, and an effortless user experience. Nonetheless, the journey towards this future is fraught with challenges. Security concerns, data privacy issues, and the necessity for a robust infrastructure are significant obstacles that require careful and strategic planning. Balancing the drive for innovation with the imperative to safeguard against potential risks will be essential in unlocking the full potential of IoV. Addressing these complexities with diligence and foresight will pave the way for a more connected, intelligent, and efficient IoV area. The Internet of Vehicles (IoV) presents a substantial

market opportunity not only for the automobile industry but also for various other sectors, including IT equipment manufacturing, the software industry, and Internet service providers. The global number of on-road vehicles is expected to rise significantly, driven by increasing motorization rates. This surge in vehicle numbers will likely lead to greater congestion and longer travel times. Even if just five minutes of the time wasted during travel globally is monetized, it is projected to generate approximately 25 billion euros in revenue annually by 2030. The automobile industry, in particular, is anticipated to see substantial profit growth, with profits expected to increase from 54 billion euros in 2012 to 79 billion euros by 2020 [17]. One of the primary objectives of IoV is to optimize travel time utilization effectively. Additionally, the recent advancements in the Internet of Things (IoT) and its high market penetration rate are significant drivers for the design and development of IoV. The automobile industry stands out as one of the fastest-growing sectors within the IoT landscape. Projections indicate that sales of connected cars will reach up to 81 million annually by 2025, with 80% of new cars featuring some form of connected drive technology. This rapid adoption of connected technologies underscores the transformative potential of IoV. The economic value generated by IoV is estimated to range between 210 and 740 billion dollars per year by 2025, highlighting its immense potential to revolutionize various industries and contribute significantly to the global economy [18].

## 5. References

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