

Edge Computing for Autonomous Vehicles

Matthew N. O. Sadiku¹, Paul A. Adekunle², Janet O. Sadiku³

¹Roy G. Perry College of Engineering, Prairie View A&M University, Prairie View, TX, USA

²International Institute of Professional Security, Lagos, Nigeria

³Juliana King University, Houston, TX, USA

ABSTRACT

The transition toward fully autonomous vehicles (AVs) represents one of the most ambitious engineering endeavors of the 21st century. Promising to revolutionize transportation by improving safety, reducing traffic congestion, and enhancing mobility efficiency, self-driving cars rely on a complex network of sensors, artificial intelligence (AI), and continuous data processing. However, the successful operation of these vehicles depends on their ability to process vast amounts of data in real time. Traditional cloud computing architectures, which rely on transmitting data to centralized, remote servers for processing and analysis, struggle to meet the stringent requirements of autonomous driving. To overcome these limitations, edge computing has emerged as a foundational technology for autonomous vehicles. Edge computing is a distributed computing paradigm that brings computation and data storage closer to the source of data generation—the “edge” of the network. Edge computing drastically reduces latency, optimizes bandwidth usage, enhances data privacy, and ensures operational resilience even in areas with poor network connectivity. This paper explores the critical role of edge computing in autonomous vehicles.

KEYWORDS: *edge computing, cloud computing, autonomous vehicles (AVs), automotive industry.*

INTRODUCTION

The evolution of autonomous vehicles (AVs) from science fiction to reality has brought forth unprecedented challenges in data processing and communication. The automotive industry is undergoing a profound transformation, shifting from mechanical transportation to intelligent, connected mobility. At the heart of this revolution is the autonomous vehicle (AV), a complex system that relies on artificial intelligence (AI), advanced sensors, and high-speed communication to navigate dynamic environments. Modern self-driving cars are generating terabytes of data per hour from an array of sensors including LiDAR, radar, and cameras. As AVs progress toward Level 4 and Level 5 autonomy, the volume of data they generate is staggering. Traditional cloud computing architectures, while powerful, introduce unacceptable latency for real-time, safety-critical decisions. Edge computing has emerged as the foundational architecture to bridge this gap, bringing computation directly to the vehicle and nearby infrastructure. It is a distributed computing paradigm that brings computation and data

storage closer to the location where it is needed. Figure 1 shows the symbol of edge computing [1].

Edge computing in autonomous vehicles processes data on the vehicle and not via a cloud server. In edge computing, data is processed close to the source of data generation, such as on local devices or nearby edge servers, rather than being sent to a distant cloud or centralized data center. The use of edge computing in autonomous vehicles is important to make decisions in real-time. In this distributed architecture, the cloud is not replaced but repurposed. In the context of autonomous vehicles, this architecture is multi-layered, comprising in-vehicle edge nodes, roadside edge infrastructure, and the traditional cloud. The deployment of 5G networks is inextricably linked to the success of edge computing in AVs [2].

CONCEPT OF EDGE COMPUTING

The history of edge computing traces back to the introduction of content delivery networks in the 1990s. The concept was straightforward: place servers close to end-user locations for faster cached

How to cite this paper: Matthew N. O. Sadiku | Paul A. Adekunle | Janet O. Sadiku "Edge Computing for Autonomous Vehicles" Published in International

Journal of Trend in Scientific Research and Development (ijtsrd), ISSN: 2456-6470, Volume-10 | Issue-4, August 2026, pp.55-65,

URL: www.ijtsrd.com/papers/ijtsrd141905.pdf



IJTSRD141905

Copyright © 2026 by author (s) and International Journal of Trend in Scientific Research and Development Journal. This is an Open Access article distributed under the terms of the Creative Commons Attribution License (CC BY 4.0) (<http://creativecommons.org/licenses/by/4.0>)



image and video transmission. Edge computing was created jointly by Microsoft and their academic collaborators. Today, more and more services are pushed from the cloud to the edge of the network. Since data is increasingly being produced at the edge of the network, it would be efficient to process the data there. Keeping this data closer to its users (at the edge) eliminates many of the problems inherent with the public cloud model. Figure 2 shows how edge computing works [3]. Edge computing allows data from IoT devices to be analyzed at the edge before being sent to the cloud.

Edge computing (EC) or edge cloud is a computing paradigm where substantial compute and storage resources are placed at the edge of the Internet, in close physical proximity to mobile devices, sensors, end users, and IoT devices. It refers to bringing the flexibility and openness of cloud-native infrastructure to that local infrastructure. The idea of “edge” is to do processing near the data source. The terms “cloudlets,” “micro data centers,” and “fog” are used to refer to these small, edge-located computing nodes or data centers. A *cloudlet* is a cluster of computers well connected to the Internet and can be treated as “data center in a box.” The main objective of a cloudlet is to extend the remote datacenter cloud services in close proximity to the end users. Physical proximity is the essence of edge computing since it improves latency, bandwidth, trust, and survivability. While the cloud has revolutionized the way we deal with data, the next wave of that revolution will happen at the edge [4].

A standard edge computing framework consists of three distinct levels [5]:

- The cloud: Which manages the overall data storage and processing.
- The edge: Tasked with near-instantaneous data handling.
- The device: Responsible for initial detection and basic data processing.

These three levels are related in the edge computing architecture shown in Figure 3 [6].

Edge computing covers a wide range of technologies such as wireless sensor networks, distributed data storage, and augmented reality. This has made its way to becoming the core of the data center. The term “edge” refers to the computing devices that sit closer to the sources of data, where the digital world meets the real world. These edge devices typically reside away from the centralized computing available in the cloud and are being created with increasingly compute capabilities. Typical edge devices are smartphones, tablets, sensors, wearables, routers,

switches, integrated access devices, multiplexers, smart TV, modern cars, and a variety of MAN/WAN access devices. Edge computing enables analytics and data gathering to occur at the source of the data.

Edge computing covers a spectrum of technologies such as cloudlets, fog computing, and mobile edge computing. A combination of edge and cloud computing is referred to as *fog computing* because it combines centralized and distributed computing resources into a single architecture. (Edge computing is a relatively new concept that should not be confused with fog computing.) It is practically unsafe and unnecessary to send such a large amount of data to the cloud. A comparison between cloud computing, edge computing is illustrated in Figure 4 [7]. Compared to cloud computing, fog computing and edge computing have the following five advantages [8]: (1) greater data transmission speed, (2) less dependence on limited bandwidths, (3) greater privacy and security, (4) greater control over data generated in foreign countries where laws may limit use or permit unwanted governmental access, and (5) lower costs because more sensor-derived data are used locally, and less data are transmitted remotely. Figure 5 shows edge computing [9], while Figure 6 shows how edge data is processed [3].

WHAT ARE AUTOMOUS VEHICLES?

Autonomous vehicles constitute one of the most spectacular recent developments of AI. As opposed to human-driven vehicles, autonomous vehicles essentially refer to self-driving vehicles. They are smart vehicles that are able to perceive their environment and to move on accordingly without human intervention. They operate with the capability to have automatic motions and navigate themselves depending on the environments and scheduled tasks. Figure 7 shows the architecture of autonomous car [10].

Autonomous vehicle or driverless car is an ambitious project which requires the fusion of many technologies like electronics, communications, mechatronics, software engineering, artificial intelligence, GPS, and industrial IoT. It is a vehicle that uses a combination of sensors, cameras, radar, and artificial intelligence (AI) to travel between destinations without a human operator. It is designed to be able to detect objects on the road, maneuver through the traffic without human intervention, and get to the destination safely. It is fitted with AI-based functional systems such as voice and speech recognition, gesture controls, eye tracking, and other driving monitoring systems. Several companies have announced their plan to get involved in autonomous or driverless and electric vehicle technology

Connected and autonomous vehicles (AVs) are now becoming a cornerstone of the increasingly connected world. They are receiving a lot of attention from manufacturers, service providers, governments, universities, consumers, and other stakeholders. The main goal of autonomous vehicles is to build a self-driving system that can perceive the road better than the best human driver. They are incredible innovation that will likely transform transportation, especially in urban environments, in the near future. Although autonomous vehicles can improve performance and safety, there are a myriad of serious technology, regulatory, and security challenges to consider in preparation for full vehicle autonomy.

Autonomous vehicles combine artificial intelligence (AI) and robotics. They are regarded as a promising answer to traffic jams, accidents, and environmental pollution. They will constitute the backbone of future next-generation intelligent transportation systems (ITS) providing travel comfort and road safety along with a number of value-added services. They are used in search and rescue, urban reconnaissance, mine detonation, supply convoys, etc. [11]. They can help save lives on the battlefield.

Autonomous vehicle (AV) is also described as “driverless,” “robotic,” or “self-driving.” AV is regarded as a multidisciplinary technology. The enabling technologies in support of connected autonomous vehicles include camera, GPS & GNSS, and sensors, radar, LiDAR (Light Detection and Ranging), and Internet of things. The race to develop autonomous vehicles has heated up with many major automotive manufacturers such as Tesla, Audi, General Motors, Mercedes Benz, Uber, Google, and Amazon [12]. Figure 8 shows the main set of sensors present in an AV [13], while Figure 9 shows some autonomous vehicles [14].

EDGE COMPUTING IN AUTONOMOUS VEHICLES

The transition from human-driven automobiles to fully autonomous vehicles (AVs) represents a paradigm shift in the automotive industry. The integration of autonomous vehicles (AVs) into modern transportation systems promises to revolutionize mobility, enhance road safety, and reduce traffic congestion. Historically, connected devices have relied on cloud computing for heavy processing tasks. However, in the context of autonomous driving, sending raw sensor data to a remote cloud server, waiting for analysis, and receiving a control command introduces a round-trip latency of 100 to 500 milliseconds. The sheer volume of data generated by self-driving cars, coupled with the uncompromising need for split-second, life-saving

decisions, renders traditional cloud computing insufficient for real-time operations. For a vehicle traveling at highway speeds, a fraction of a second can mean the difference between a safe maneuver and a catastrophic collision. This reality has led to the adoption of edge computing—a distributed paradigm that processes data at or near the source of generation. By bringing computation to the network edge, the automotive industry can overcome the latency, bandwidth, and reliability limitations inherent in cloud-centric architectures. Edge computing ensures that vehicles can make split-second, life-saving decisions independently, while also safeguarding passenger privacy and enabling advanced V2X communication. This architecture drastically reduces latency to under 10 milliseconds, enabling the ultra-fast perception-action loops required for autonomous driving [2]. Figure 10 is a representation of edge computing in autonomous vehicles [15].

APPLICATIONS OF EDGE COMPUTING IN AUTONOMOUS VEHICLES

The integration of edge computing facilitates several mission-critical applications that define autonomous driving capabilities. Figure 11 displays some use cases of edge computing in automotive [16]. Common applications of edge computing in AVs include the following [2,16]:

- *Real-time Perception:* Autonomous vehicles must accurately perceive their environment to navigate safely. Edge AI algorithms process data from cameras, LiDAR, and radar locally to detect pedestrians, cyclists, other vehicles, and road signs. Because different sensors have unique strengths—cameras provide high resolution and color, LiDAR offers precise depth and distance, and radar works well in adverse weather—sensor fusion is required to create a cohesive, 3D model of the environment. Performing sensor fusion at the edge ensures this environmental model is updated instantaneously, enabling the vehicle to react to sudden changes.
- *Vehicle-to-Everything (V2X) Communication:* Edge computing powers V2X connectivity, allowing vehicles to share critical information with each other and the infrastructure. V2X allows autonomous vehicles to communicate in real time with other vehicles (V2V), roadside infrastructure (V2I), and pedestrians (V2P). For example, a vehicle detecting black ice can instantly alert following vehicles via V2V communication, or a smart traffic light can broadcast its signal phase timing via V2I. This collaborative perception significantly enhances safety. According to the European

Telecommunications Standards Institute (ETSI), standardized edge computing is crucial for ensuring the interoperability and low latency required for effective V2X services. Figure 12 shows types of vehicle-to-everything (V2X) communications like V2V, V2I, V2P/V2H, V2S, V2R [13].

- *High-Definition (HD) Mapping:* Autonomous vehicles rely on high-definition (HD) maps, which contain centimeter-level precision regarding lane markings, road boundaries, and static obstacles. Maintaining these maps requires constant updates. Edge computing allows vehicles to compare their real-time sensor data against the stored HD map. If a vehicle detects a discrepancy-such as a new construction zone-it can process this anomaly at the edge and upload a lightweight update to the cloud or share it with nearby vehicles via MEC, rather than transmitting massive raw video files.
- *Predictive Maintenance:* Edge Computing can also be used to improve vehicle maintenance. By continuously monitoring telemetry data and component performance locally, edge computing enables predictive maintenance. AI models running on the vehicle's edge can detect subtle anomalies in engine vibration, battery health, or sensor calibration before a failure occurs. This local analysis provides instant alerts to the driver or fleet operator, improving reliability and reducing downtime without needing to stream all diagnostic data to the cloud.
- *Fleet Management:* Edge Computing can also improve fleet management by providing real-time information on vehicle location, fuel consumption, and driver behavior. This data can be used to optimize routes, reduce fuel consumption, and improve driver safety.

BENEFITS

The integration of edge computing into AVs provides several transformative advantages that make autonomous driving viable on a large scale. Edge computing offers a significant reduction in latency, bandwidth usage, improved safety, reduced downtime, enhanced driving experience, and data storage requirements, allowing autonomous cars to operate with greater efficiency and cost-effectiveness. Other benefits of edge computing in AVs include the following [2,17,18]:

- *Latency Reduction:* The primary and most critical benefit of edge computing in autonomous vehicles is the dramatic reduction in latency. Low latency is at the core of autonomous driving

safety. Cloud computing typically introduces a round-trip latency of 100 to 500 milliseconds, primarily due to the physical distance data must travel and network routing delays. This delay is unacceptably high for emergency situations. By eliminating the round-trip delay to remote servers, edge computing reduces response times to under 10 milliseconds, enabling the real-time reactions required for safe driving.

- *Bandwidth Efficiency:* The sheer volume of data generated by AV sensors poses a significant challenge for network infrastructure. Transmitting up to 5 terabytes of raw data per vehicle per hour to the cloud would overwhelm existing cellular networks, even with the advent of 5G. Such transmission is economically and technically unfeasible. Edge computing provides a highly efficient solution to this bandwidth bottleneck. Instead of transmitting raw data, the onboard edge computing systems process, filter, and analyze the data locally. It acts as an intelligent filter, processing raw data locally and only transmitting essential insights or critical event logs to the cloud.
- *Operational Resilience:* Autonomous vehicles must operate safely in areas with poor or no cellular coverage, such as tunnels or rural roads. A vehicle cannot simply cease functioning or compromise passenger safety if it enters a tunnel, travels through a rural area with poor cellular coverage, or encounters a network outage. Edge computing ensures that critical perception and control functions remain fully operational regardless of network availability.
- *Enhanced Data Privacy:* The data collected by autonomous vehicles is highly sensitive. It includes precise location histories, behavioral patterns, and potentially biometric inputs from passengers. Transmitting this continuous stream of personal data to centralized cloud servers increases the risk of interception, unauthorized access, and mass data breaches. Processing sensitive data-such as passenger locations and behavioral patterns-locally reduces the risk of interception during transmission.
- *Real-time Decision-making:* The advancement of autonomous driving systems has raised the importance of real-time decision-making abilities. The dream of truly autonomous vehicles (AVs) navigating our roads hinges on an astonishing ability of making life-or-death decisions in the blink of an eye. The successful deployment of AVs depends on the ability to process vast amounts of data in real-time, ensuring swift

decision-making. The core of real-time decision-making is promptly analyzing a continuous data flow to enable immediate reactions to unexpected situations, possible dangers, and uncertain road conditions. The self-driving cars make thousands of decisions every second, but they cannot afford to wait for the cloud to respond. Edge computing for autonomous vehicles enables processing to occur directly on the vehicle, enabling quick, safe, intelligent driving. Edge computing stores data locally to make in-the-moment decisions, which reduces latency and increases safety. In automotive edge computing, sensors collect raw data, and automotive edge computing processes it within the car. This collaboration allows for faster decisions, reduces risk, and improves driving safety, especially in difficult, unpredictable situations.

- *Market Growth:* The automotive industry is undergoing a major digital transformation. Cars are not just mechanical machines, but intelligent, software-driven machines. As this shift becomes more intense, edge computing in the automotive market becomes a driving technology for safety, automation, and connectivity. The automotive edge computing market is projected to grow rapidly, with estimates suggesting it will reach \$39 billion by 2032. Scaling this infrastructure requires massive investments in 5G deployment, roadside MEC servers, and intelligent traffic systems.
- *Environmental Benefits:* Edge processing reduces unnecessary data transmission to distant servers, which saves significant energy. Cars only send essential information to the cloud instead of streaming everything they see. This efficiency helps make autonomous vehicles more environmentally friendly.

CHALLENGES

In spite of its clear benefits, the deployment of edge computing in autonomous vehicles faces several significant technical and regulatory hurdles. Providing sufficient computing power within the constrained thermal and power envelopes of a vehicle is a major engineering hurdle. Other challenges of edge computing in AVs include the following [2,17,19]:

- *Hardware Costs:* Equipping autonomous vehicles with high-performance edge processors significantly raises manufacturing costs. These specialized chips, capable of handling complex AI workloads, add to the overall price of vehicles. For mass-market adoption, reducing hardware expenses will be critical.
- *Thermal Management:* The intense computational workloads generated by edge AI processors produce substantial heat. In the confined space of a vehicle, which is subject to extreme environmental temperatures ranging from freezing cold to scorching heat, effective thermal management is paramount. Failure to dissipate heat adequately can lead to thermal throttling, where processors reduce their speed to prevent damage, thereby compromising the real-time processing capabilities essential for safe driving. Ensuring 100% AI reliability in harsh conditions requires innovative cooling solutions and thermal-aware optimization frameworks. Automotive makers should design lightweight computing devices that will not overheat when running real-time AI workloads. The excess heat can be fatal, a threat to hardware, or even a security risk.
- *Heterogeneous Integration:* Different autonomous driving tasks are best suited for different types of processors. For example, GPUs excel at parallel processing for image recognition, Central Processing Units (CPUs) are efficient for logical control tasks, and Digital Signal Processors (DSPs) are optimal for feature extraction. Integrating these diverse hardware components into a cohesive, heterogeneous computing architecture that can efficiently schedule and allocate workloads in real time is a complex systems engineering problem.
- *V2X Communication Latency:* V2X technology enables vehicles to communicate with each other (V2V), roadside infrastructure (V2I), and pedestrians (V2P). This cooperative driving expands a vehicle's situational awareness beyond its line of sight, such as receiving advance warnings of an accident around a blind corner. However, V2X requires ultra-low latency and highly reliable wireless networks, typically facilitated by 5G. Maintaining consistent, robust connectivity in challenging environments like tunnels, rural areas, or densely populated urban canyons with signal interference poses a significant challenge.
- *Bandwidth Congestion:* Even with localized edge processing, AVs must transmit critical metadata, anomaly reports, and system health metrics to the cloud, while receiving updated AI models and traffic data. When millions of AVs are deployed, the aggregate data transmission could overwhelm existing cellular networks, leading to bandwidth congestion and packet loss. Intelligent data triage-determining exactly what data must be sent to the cloud and what can be discarded-is necessary to

mitigate this issue, but designing foolproof filtering algorithms is difficult.

- **Safety:** Safety is the most important requirement for autonomous vehicles; hence, the ultimate challenge of designing an edge computing ecosystem for autonomous vehicles is to deliver enough computing power, redundancy, and security so as to guarantee the safety of autonomous vehicles. Specifically, autonomous driving systems are extremely complex. These complexities impose numerous challenges for the design of autonomous driving edge computing systems. Thus, it is imperative to deliver sufficient computing power with reasonable energy consumption, to guarantee the safety of autonomous vehicles, even at high speed.
- **Cybersecurity:** Cybersecurity risk increases as vehicles become connected devices. As AVs become heavily reliant on edge computing and interconnected networks, their attack surface expands exponentially, making cybersecurity a paramount concern. As AVs become highly connected edge nodes, they become targets for cyberattacks. Securing the edge involves protecting sensor data, communication links, and onboard operating systems. However, these models are highly vulnerable to adversarial attacks, where subtle, often imperceptible perturbations are introduced to the input data (e.g., placing specific stickers on a stop sign) to deceive the AI into making incorrect classifications.
- **Hardware Constraints:** Automotive edge computing must operate under strict Size, Weight, and Power (SWaP) constraints. High-performance GPUs consume significant energy and generate heat, which can degrade a vehicle's range, especially in electric vehicles (EVs). Developing power-efficient, domain-specific AI accelerators (ASICs) is a critical area of ongoing research.
- **Regulatory Compliance:** The deployment of edge computing infrastructure, particularly roadside units, involves complex regulatory and jurisdictional challenges. Determining who owns, maintains, and secures the edge infrastructure—whether it be local governments, telecom operators, or private consortia—remains unresolved in many regions. Regulatory bodies will increasingly focus on defining the functional safety requirements for in-vehicle edge systems, ensuring that AI-driven decisions are transparent, explainable, and resilient to failure.

- **Standardization:** The successful deployment of edge computing in autonomous driving requires seamless collaboration across a vast ecosystem of automotive manufacturers, telecommunications providers, hardware vendors, and software developers. Universal adoption of standards is still lacking. Organizations like the European Telecommunications Standards Institute (ETSI) and IEEE are actively working on standardizing multi-access edge computing (MEC) and cybersecurity protocols to ensure interoperability and safety across different manufacturers. As edge computing architectures mature, the industry will require standardized protocols for V2X communication, data sharing, and safety certification.

FUTURE OF EDGE COMPUTING IN AUTONOMOUS VEHICLES

The transition from human-driven cars to fully autonomous vehicles represents one of the most profound technological shifts of the 21st century. The future of transportation relies on the ability to compute at the speed of reality, and edge computing is the engine driving that capability. The future of autonomous driving relies heavily on the evolution of smart cities. As urban infrastructure becomes more intelligent, the distinction between the vehicle's edge and the city's edge will blur. EV charging stations, streetlights, and toll booths will act as distributed edge nodes, processing traffic flow data and providing predictive analytics to passing AV fleets. This integration will be essential for managing urban congestion and enabling large-scale robotaxi deployments.

The future of edge computing in autonomous vehicles is characterized by increased decentralization, smarter infrastructure integration, and advanced machine learning techniques. It holds immense potential for revolutionizing the transportation industry. Hybrid architectures are the future. The rollout of 5G networks will synergize with edge computing. As technologies like 5G, advanced AI accelerators, and V2X communication continue to evolve, the integration of edge computing will be the driving force that finally brings fully autonomous transportation from concept to reality. To continuously improve AI models without compromising data privacy or overwhelming network bandwidth, the industry is moving toward federated learning. In this paradigm, raw sensor data never leaves the vehicle. Instead, the vehicle uses its local data to train a localized version of the AI model [2].

CONCLUSION

Edge computing is not merely an enhancement to autonomous vehicles; it is a fundamental necessity for their safe and scalable deployment. It is undeniably the linchpin that will enable the transition to fully autonomous, Level 4 and Level 5 vehicles. By decentralizing computation and bringing intelligence to the network's edge, AVs can achieve the real-time responsiveness, reliability, and bandwidth efficiency necessary for safe operation in complex environments. By processing massive volumes of sensor data locally, edge architectures provide the ultra-low latency and deterministic reliability required for safe, real-time driving decisions. However, the path to widespread commercialization is obstructed by significant challenges. Engineers must develop hardware that balances extreme computational power with stringent energy and thermal limits. As the technology matures, edge computing will become the backbone of safer, smarter, and more reliable autonomous mobility. More information on edge computing in autonomous vehicles can be found in the book in [20].

REFERENCES

- [1] T. West, "Using edge computing for real-time marketing insights," April 2024, <https://www.vdigitalservices.com/utilizing-edge-computing-for-real-time-marketing-insights/>
- [2] <https://manus.im>
- [3] "Harnessing edge computing for business growth," <https://www.vertiv.com/en-us/solutions/vertiv-guide-to-edge-computing/>
- [4] M. N. O. Sadiku, A. M. Oteniya, and S. M. Musa, "Edge computing," *International Journal of Trend in Research and Development*, vol. 5, no. 4, July – Aug. 2018, pp. 553-554.
- [5] A. Harper, "How is edge computing enhancing telecom speeds for end users?" <https://www.jevera.software/post/how-is-edge-computing-enhancing-telecom-speeds-for-end-users>
- [6] J. Gonzalez et al., "Edge computing architecture and use cases," <https://lfdge.org/edge-computing-architecture-and-use-cases/>
- [7] "Edge computing needs edge AI," August 2018, <https://www.imagimob.com/blog/edge-computing-needs-edge-ai#>
- [8] D. C. Klonoff, "Fog computing and edge computing architectures for processing data from diabetes devices connected to the medical Internet of things," *Journal of Diabetes Science and Technology*, vol. 11, no. 4, 2017, pp. 647 – 652.
- [9] "Edge computing definition and use cases," March 2020, <https://jelvix.com/blog/what-is-edge-computing>
- [10] T. Raviteja and R. Vedaraj, "An introduction of autonomous vehicles and a brief survey," *Journal of Critical Reviews*, vol. 7, no. 13, 2020, pp. 196-202.
- [11] J. Connelly et al., "Current challenges in autonomous vehicle development," *Proceedings of SPIE*, May 2006.
- [12] M. N. O. Sadiku, S. M. Musa, and A. Ajayi-Majebi, "Artificial intelligence in autonomous vehicles," *International Journal of Trend in Scientific Research and Development*, vol. 5, no. 2, Jan.-Feb. 2021, pp. 715-720.
- [13] A. Biswas and H. Wang, "Autonomous vehicles enabled by the integration of IoT, edge intelligence, 5G, and Blockchain," *Sensors*, vol. 23, no. 4, February 2023.
- [14] "Edge computing use cases including real-time analytics, IoT device management, and low-latency applications such as autonomous vehicles and augmented reality," <https://stock.adobe.com/images/edge-computing-use-cases-including-real-time-analytics-iot-device-management-and-low-latency-applications-such-as-autonomous-vehicles-and-augmented-reality/748760970>
- [15] "Teknologi edge computing di autonomous vehicles," October 2024, <https://teknovortexel.com/2024/10/27/teknologi-edge-computing-di-autonomous-vehicles/>
- [16] S. Bhattacharyya, "The future of autonomous cars with edge computing," July 2023, https://www.analyticssteps.com/blogs/future-autonomous-cars-edge-computing#google_vignette
- [17] K. Kumar, "Edge computing in autonomous vehicles: Tech explained," February 2026, <https://theninehertz.com/blog/edge-computing-in-autonomous-vehicles>
- [18] Sushree, "Edge computing in autonomous vehicles: What powers it in 2026," September 2025, <https://kanerika.com/blogs/edge-computing-in-autonomous-vehicles/>

- [19] S. Liu et al., "Edge computing for autonomous driving: Opportunities and challenges," *Proceedings of the IEEE*, vol. 107, no. 8, August 2019, pp. 1697-1716.

- [20] R. Gutierrez, *Edge AI and Autonomous Systems Explained : A Beginner's Guide to Smart Machines, Real-Time Computing, and the Future of Intelligence*. Kindle Edition, 2026.



Figure 1 The symbol of edge computing [2].

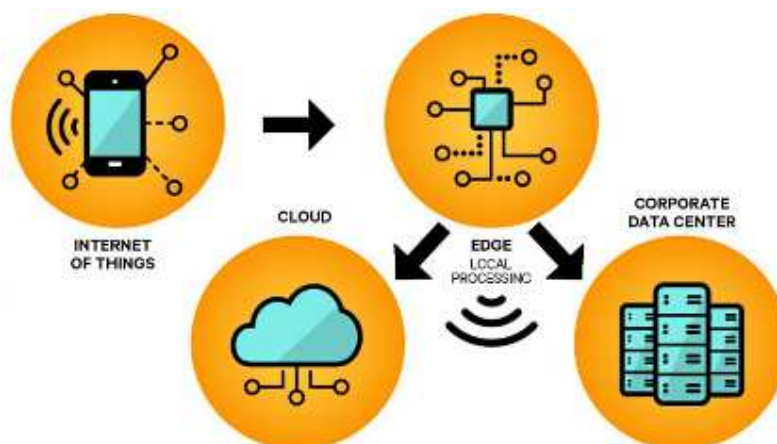


Figure 2 How edge computing works [3].

Edge Computing Architecture

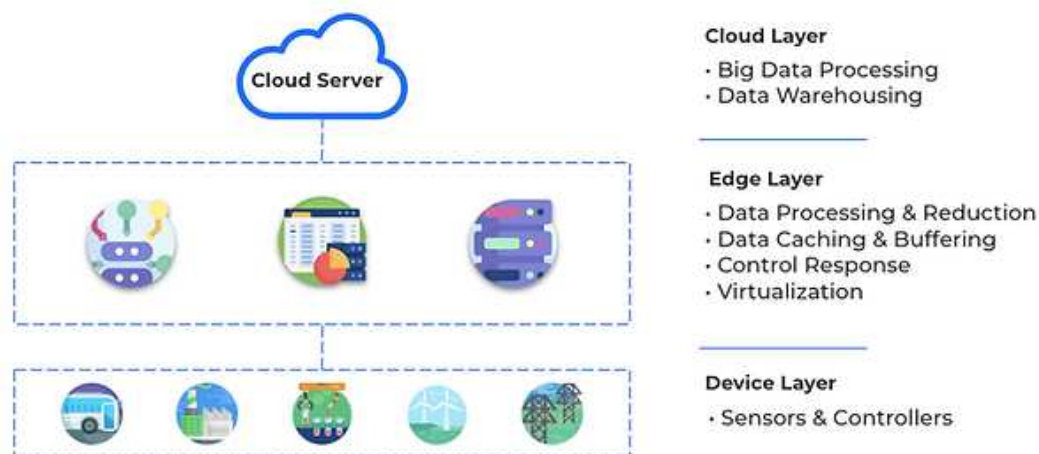


Figure 3 Edge computing architecture [6].

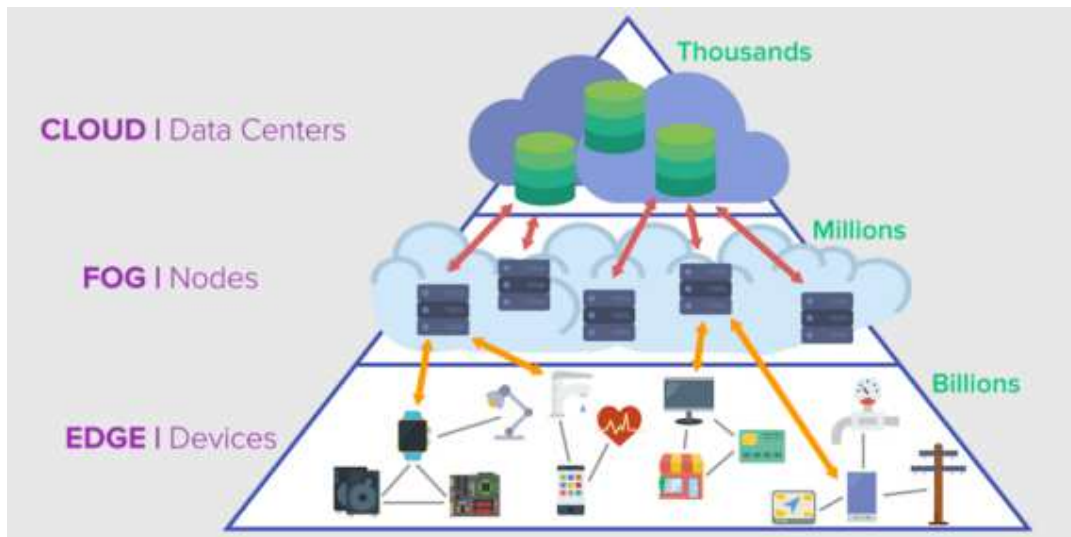


Figure 4 The relationship between cloud computing, edge computing, and fog computing [7].

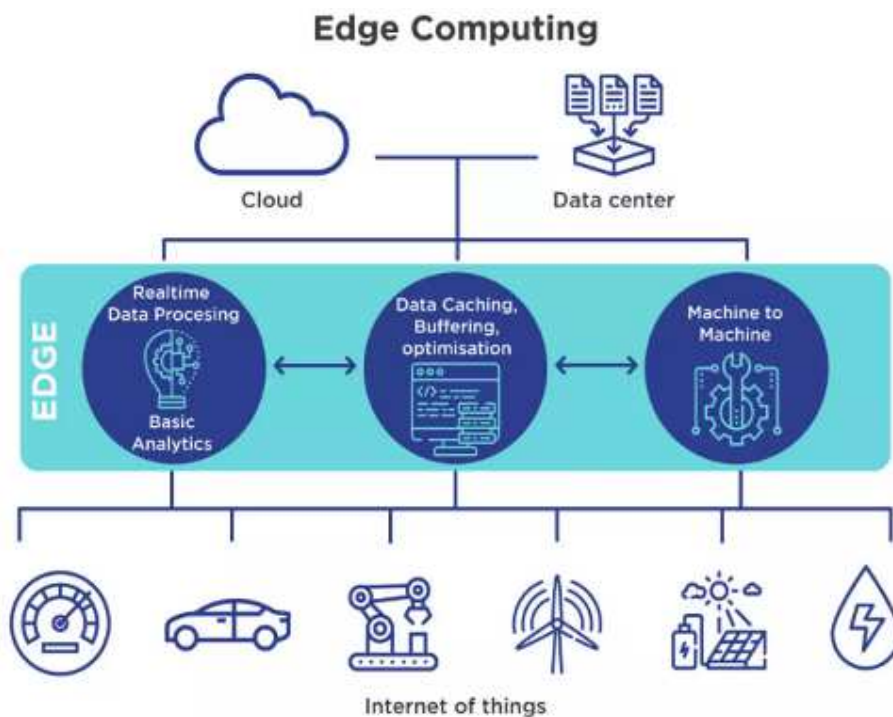


Figure 5 Edge computing [9].

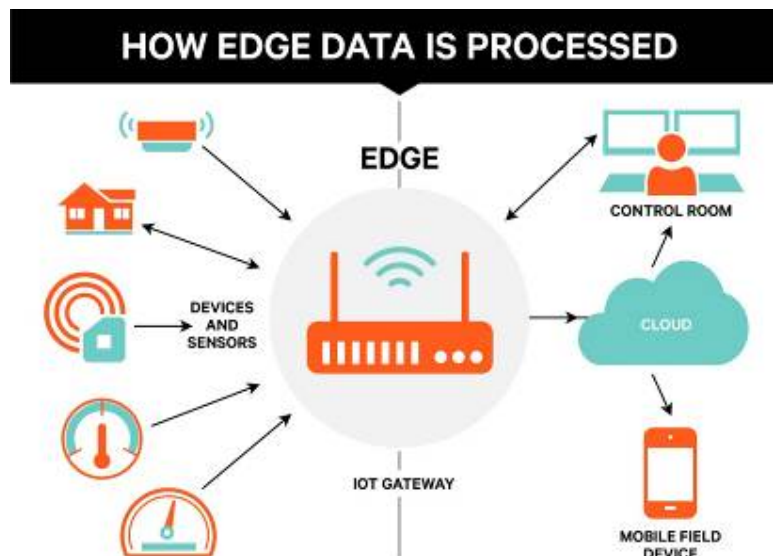


Figure 6 How edge data is processed [3].

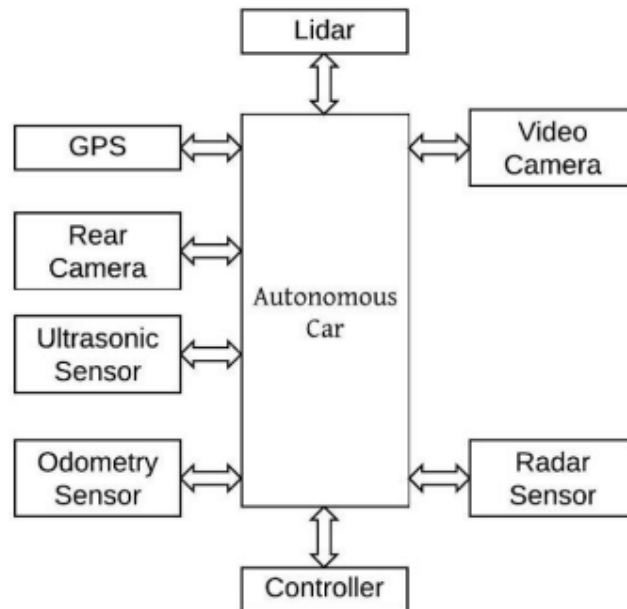


Figure 7 Architecture of Autonomous Car [10].



Figure 8 The main set of sensors present in an AV [13].



Figure 9 Some autonomous vehicles [14].



Figure 10 A representation of edge computing in autonomous vehicles [15].



Figure 11 Some use cases of edge computing in automotive [16].

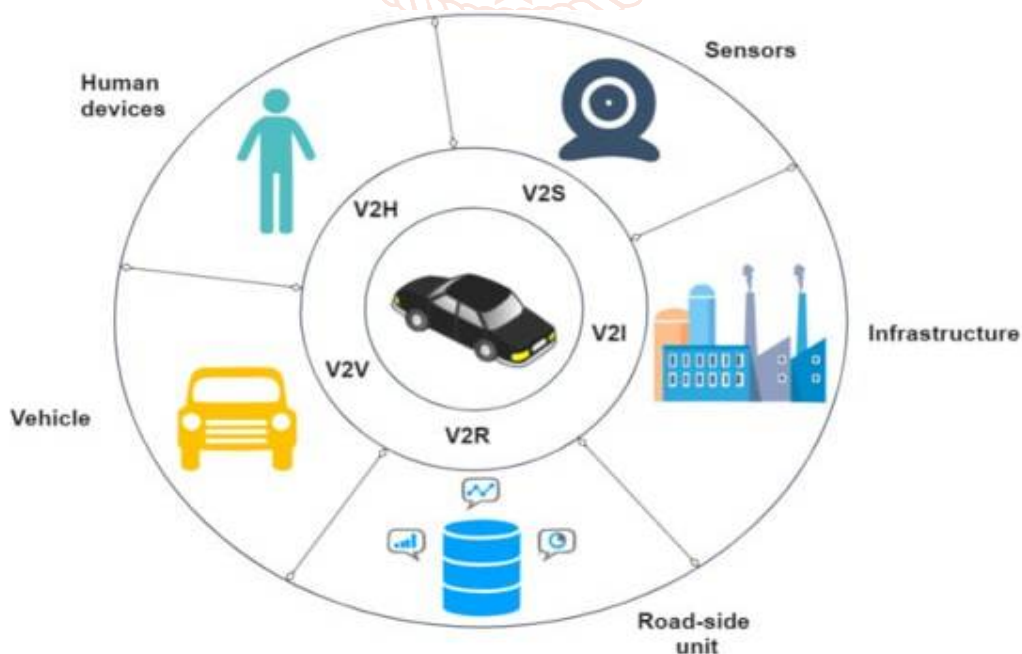


Figure 12 Types of vehicle-to-everything (V2X) communications [13].