

# AI-Powered 5G Networks: A Roadmap for The Future of Wireless Communications

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## Abstract

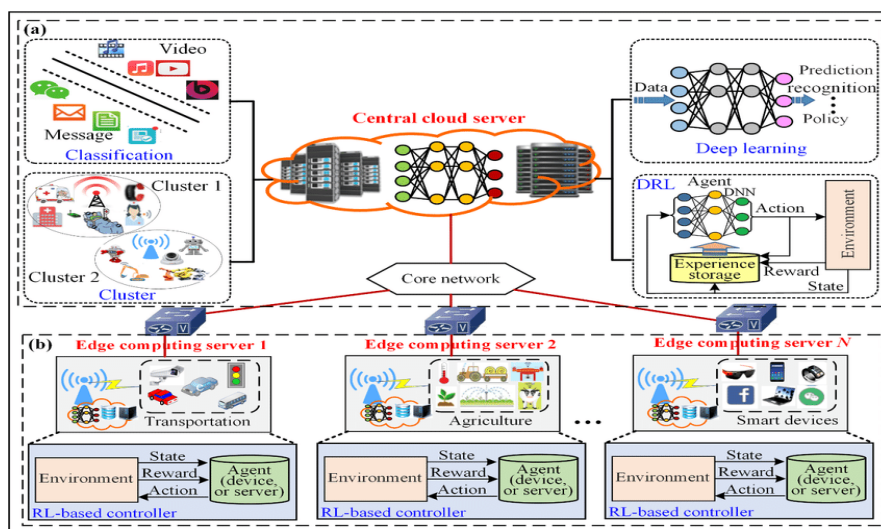
Artificial intelligence (AI) is a rapidly evolving technology that has the potential to revolutionize 5G networks. AI could be employed to improve the performance, security, and user experience of 5G networks in a variety of ways. For example, AI could be employed to improve network performance, security, and personalize the user experience. However, there are several challenges that need to be addressed to successfully implement AI in 5G networks. Some of the most significant challenges include Data privacy and security, complexity, latency, cost, skills shortage. Despite these challenges, the potential benefits of AI in 5G networks are significant. AI could be employed to improve the performance, security, and user experience of 5G networks, and it is likely that AI will play an increasingly significant role in the development and deployment of 5G networks in the future. This study spots the lights on the benefits of implementing AI Techniques in 5G Networks. It discusses different AI Techniques that can be merged in 5G networks, applications, key requirements for implementing AI in 5G networks, available Simulation Tools to help researchers in this field. This will assist researchers in choosing the optimal tool for their work while incorporating additional features. Finally, this study discussed in details challenges and future research directions for implementing AI techniques in 5G networks.

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**Keywords:** *AI, 5G, Wireless, Networks, IoT, Optimization, Security.*

# 1. INTRODUCTON

5G is the fifth generation of cellular network technology, and it promises to deliver much faster speeds, lower latency, and greater capacity than previous generations [1]. AI is playing a key role in the development of 5G networks, as it could be employed to improve a variety of aspects of network performance, including [2-6]: (i) Network optimization: AI could be employed to analyze vast amounts of data generated by the network to identify patterns and trends. This information can then be used to optimize the network for different traffic types and user densities. For example, AI could be employed to predict traffic patterns and adjust the network accordingly, which can help to reduce congestion and improve performance. (ii) Network security: AI could be employed to identify and mitigate security threats to 5G networks. AI could be employed, for instance, to identify and stop cyberattacks or to examine network traffic for suspicious trends. (iii) Network automation: AI can also be used in automating a variety of network management tasks, such as network provisioning, fault detection, and performance optimization. This can help to reduce the workload on network operators and improve the overall efficiency of the network. (iv) User experience: AI could be employed to personalize the user experience on 5G networks. For example, AI could be employed to recommend content to users, or to optimize network resources for different applications. Figure 1 [9] presents the architecture of AI-Empowered MEC. This can help to improve the overall quality of service for 5G users. Further, there are some specific examples [7-8] of how AI is being used in 5G networks today. First, AT&T is using AI to improve the efficiency of its 5G network. The company is using AI to automate network provisioning, which has helped to reduce the time it takes to deploy new 5G services. AT&T is also using AI to predict traffic patterns and adjust the network accordingly, which has helped to improve performance and reduce congestion. Second, Verizon is using AI to improve the security of its 5G network. The company is using AI to analyze network traffic for malicious patterns, and to detect and respond to cyber-attacks. Verizon is also using AI to automate security tasks, such as vulnerability scanning and patch management. Third, Qualcomm is using AI to personalize the user experience on 5G networks. The company is developing AI-powered features that will recommend content to users, optimize network resources for different purposes, and enhance the total quality of service. These are just a little example of how AI is being used in 5G networks today. As 5G continues to evolve, AI is likely to play an even greater role in improving the performance, security, and user experience of 5G networks.



**Fig.1.** Architecture of AI Empowered MEC.

According to a recent study by Grand View Research [10], the global market for AI in 5G networks is expected to grow from \$3.3 billion in 2021 to \$24.4 billion by 2028. This growth is being driven by the increasing demand for high-speed, reliable, and secure wireless connectivity. AI is being used to improve the performance, security, and user experience of 5G networks in a variety of ways. Here is a summary some statistics for published papers in implementing AI techniques in 5G networks based on Scopus and science direct databases:

- In 2022, there were over 1,000 papers published on the topic of AI in 5G networks.
- The number of papers published on this topic has been growing exponentially in recent years.
- Most of these papers are published in top academic journals, such as IEEE Transactions on Wireless Communications, IEEE Journal on Selected Areas in Communications, and ACM Transactions on Sensor Networks.
- The papers cover a wide range of topics, including network slicing, security, user experience, and network optimization.
- The papers are authored by researchers from all over the world, including China, the United States, Europe, and South Korea.

These statistics show that the research on AI in 5G networks is rapidly growing. This is a promising sign for the future of 5G networks, as AI has the promise to transform the way we use wireless connectivity. This paper has three-fold contributions: First, casting light on the power of merging AI techniques in the 5G network. Second, well understanding the different applications of an AI-empowered 5G network, its key requirements, simulation environments and presenting a detailed study about it. Third, it introduces challenges and future research directions in this field.

Finally, this paper is structured as follows: Section 2 delivers an overview of AI Techniques for 5G Networks followed by Applications of AI in 5G Networks in section 3. In Section 4, a Literature Review has been presented followed by Key Requirements for implementing AI in 5G networks and simulation tools in section 5 and 6, respectively. Section 7 discusses Challenges and Future Research Directions. Finally, the paper concludes with suggestions for where the field could go.

## **2. AI TECHNIQUES FOR 5G NETWORKS**

There are a variety of AI techniques that could be employed to improve 5G networks. Some of the most promising techniques include: (i) Machine learning: Machine learning could be employed to train models that can predict traffic patterns, identify security threats, and optimize network resources. (ii) Deep learning: Deep learning is a type of machine learning that could be employed to train models that can learn complex patterns from data. Deep learning is particularly well-suited for tasks such as image recognition and natural language processing, which are both important for 5G networks. (iii) Reinforcement learning: Reinforcement learning is a type of machine learning that could be employed to train agents that can learn to make decisions in an environment. Reinforcement learning is particularly well-suited for tasks such as network resource allocation and self-organizing networks. Usually, AI-based algorithms that are used to increase 5G networks efficiency have the following structure:

1. Collect data: The algorithm first collects data on network traffic patterns, such as the number of consumers, the type of traffic, and the time of day.
2. Train the model: The algorithm then trains a machine learning model on the collected data.

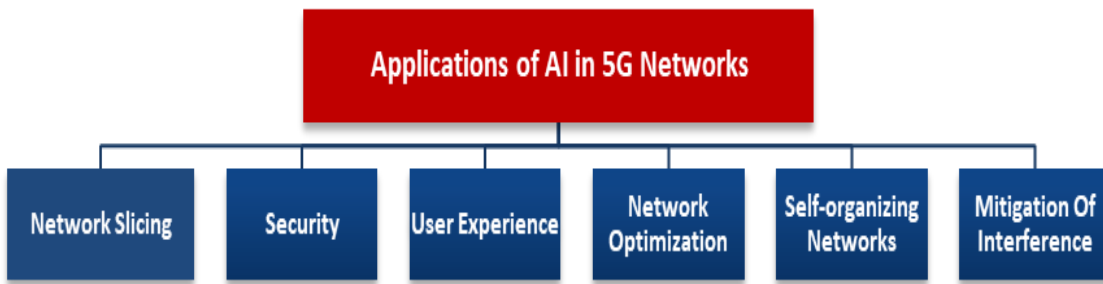
The model learns to predict traffic patterns and optimize network resources accordingly.

3. Deploy the model: The model is then deployed in the network. The model is used to make real-time decisions about network resource allocation, such as which base stations to use and how much bandwidth to allocate to each user.

Such algorithms can help to improve the efficiency of 5G networks in several ways. It can help to: (i) Reduce congestion: By predicting traffic patterns, the algorithm can aid to decrease congestion in the network. This can improve performance for all users. (ii) Optimize network resources: The algorithm can optimize network resources by allocating them more efficiently. This could assist in enhancing performance and reducing costs. (iii) Improve security: The algorithm can facilitate improving security by identifying and blocking malicious traffic. However, these kinds of algorithms do have some drawbacks. One drawback is that it requires a lot of data to train the model. This data could be expensive and difficult to collect. Another drawback is that the algorithm can be computationally intensive. This can make it difficult to deploy the algorithm in real time. Overall, this algorithm is a promising approach to improving the efficiency of 5G networks. However, it is vital to consider the drawbacks of the algorithm before deploying it in a production environment.

### **3. APPLICATIONS OF AI IN 5G NETWORKS**

AI has the power to revolutionize 5G networks in a many ways. Here are various of the most promising applications [11-15] of AI in 5G networks in Fig.2. (i) Network slicing: Network slicing is a 5G feature that allows for the creation of virtual networks within a physical network. This can be utilized to create dedicated networks for specific applications, such as AR/VR or industrial automation. AI can be employed to automate the procedure of network slicing, and to optimize the performance of virtual networks. (ii) Security: AI could be employed to improve the security of 5G networks by detecting and responding to cyber-attacks. For example, AI can be employed to analyze different networks traffic for malicious patterns, or to find and block unauthorized access to network resources. (iii) User experience: AI can be exploited to improve the user experience on 5G networks by personalizing content and services, and by optimizing network resources for different applications. For example, AI could be employed to recommend content to users, or to adjust the network bandwidth for diverse types of traffic. (iv) Network optimization: AI is employed in optimizing network performance via the prediction of traffic patterns and allocating resources. This can aid to decrease congestion and advance performance for all users. (v) Self-organizing networks: AI can be employed to create self-organizing networks that can adapt to variations in traffic patterns and network conditions. This helps to enhance the reliability and resilience of 5G networks. (vi) Mitigation of interference: AI could be employed to mitigate interference between different users and devices on the network. This can help to improve performance and reliability for all users. These are just a few of the many potential applications of AI in 5G networks. As AI technology continues to develop, it is likely that new and innovative applications of AI will be discovered



**Fig.2.** Applications of AI in 5G Networks.

#### **4. LITERATURE REVIEW**

Many researchers around the world have been published on the topic of implementing AI techniques in 5G networks. Zhang et al. in (2022) presented the AI Techniques for Improving the Performance and Security of 5G Networks by surveying the latest AI techniques that could be employed to improve the performance and security of 5G networks. The paper discusses the use of AI for tasks such as network slicing, security, user experience, and network optimization. Further, the Role of AI in 5G Networks as a Survey of Techniques and Applications by Wang et al. [18] in (2022) provides a comprehensive overview of the role of AI in 5G networks. The paper discusses the different AI techniques that could be employed in 5G networks, as well as the challenges and opportunities for using AI in 5G networks.

AI-Powered 5G Networks: The Future of Wireless Communications by Chen et al. [18] who argues that AI is essential for the future of 5G networks. The paper discusses the ways in which AI could be employed to improve the performance, security, and user experience of 5G networks. Further, Li et al. [19] in (2022) presented his study in AI in 5G Networks discussing the challenges and opportunities for implementing AI in 5G networks. The paper discusses the technical challenges of implementing AI in 5G networks, as well as the regulatory challenges and the social implications of using AI in 5G networks.

Moreover, " A Roadmap Toward Prospects for IoT Enabled 5G Networks " by Singh et al. [20] (2023) provides a roadmap for the future of AI in 5G networks. The paper discusses the key research areas that need to be addressed to fully realize the potential of AI in 5G networks. These are just a few of the many papers that have been published on the topic of implementing AI techniques in 5G networks. These papers provide a good overview of the latest research in this area, and they highlight the potential of AI to revolutionize 5G networks.

The evolution of wireless communication networks towards 5G has been a topic of extensive research and discussion in recent years. Wang et al. (2014) [21] proposed a cellular architecture for 5G systems that separates indoor and outdoor scenarios, highlighting technologies such as massive MIMO, energy-efficient communications, cognitive radio networks, and visible light communications. Yang et al. (2015) [22] focused on physical layer security to safeguard data confidentiality by exploiting the intrinsic randomness of the communications medium. The use of disruptive technologies in 5G networks can enhance security measures. Akyildiz et al. (2016) [23] identified ten key enabling technologies for 5G, including network ultra-densification, big data, and mobile cloud computing, and new radio access techniques. These technologies are expected to bring a

renaissance to wireless communication networks. Security is a crucial aspect of 5G networks, as wireless transmissions are inherently vulnerable to security breaches. The future of commercial wireless systems is also being shaped by advancements in millimeter wave technology, as discussed by Heath (2016). [24] The propagation differences, hardware constraints, and computational challenges associated with high data rates have reinvigorated research in wireless communications. Agiwal et al. (2016) [25] conducted a comprehensive survey of next-generation 5G wireless networks, highlighting architectural changes in the radio access network design, smart antennas, and cloud and heterogeneous RAN.

Looking beyond 5G, Rodriguez et al. (2017) [26] focused on reducing energy consumption in next-generation mobile small cells through secure network coding. The deployment of new disruptive technologies, such as femtocell-type cells, aims to increase capacity, support more users, and lower the cost per bit. Dore et al. (2020) [27] discussed the technology roadmap for beyond 5G wireless connectivity in the D-band, considering the trade-offs between scenario requirements and current silicon technology limits.

Finally, as discussions on the future of wireless communications continue, Letaief et al. (2019) [28-29] and Bojkovic et al. (2021) [30] explored the potential technologies for 6G networks, emphasizing the role of AI in enabling mobile applications and network optimization. The roadmap to 6G envisions an unprecedented transformation in wireless cellular systems, driven by AI-powered methodologies and technologies.

**Table 1.** Summary of Related Work [31-44]

| Reference               | Year | Focus  | Area  |
|-------------------------|------|--|---|
| Mao, et al. [31]        | 2018 | Security and privacy protection                                      | 6G networks   |
| Tong, et al. [32]       | 2019 | V2X  | Integration between AI and FC technologies                        |
| Gacunin and Wagner [33] | 2019 | Customer experience management (CEM) in 5G cellular networks         | -   |
| Ma, et al. [34]         | 2020 | AV   | Related and emerging technologies                                 |
| Sheraz, et al. [35]     | 2020 | AI-based caching techniques using ML algorithms                      | V2X   |
| Rihan, et al. [36]      | 2020 | AI and FC technologies in V2X networks                               | Resource management   |
| Lin and Zhao [37]       | 2020 | AI-based resource management   | Future wireless networks  |
| Chen, et al. [38]       | 2020 | Static and mobile sensor networks                                    | Path selection  |
| Sun, et al. [39]        | 2020 | Privacy protection   | 4G networks   |
| Wu et al. [40]          | 2020 | Security protection  | 3G networks   |
| Shafin [41]             | 2020 | Road map to enable AI-enabled cellular networks                      | Performance, complexity, and robustness                           |
| Guo [42]                | 2020 | Explainable AI for human-machine interfacing and targeted healthcare | Neural networks-based machine learning                            |
| Khan [43]               | 2022 | UAVs within the context of 6G mobile networks                        | Swarms of CAVs operating within the context of 5G mobile networks |
| Zhu, et al. [44]        | 2023 | Wireless networking  | Intelligence aspects, security and privacy, and energy efficiency |

Based on literature on Table 1, next-generation wireless networks aim for significant improvements in connectivity, data capacity, and speed. Denser base stations with wider bandwidths are crucial for achieving these goals, but pose challenges. While wireless communication is the foundation for systems like IoT, robotics, and self-driving cars, limitations like ultra-low latency and big data handling create hurdles.

AI emerges as a powerful solution due to its adaptability and seamless integration within these systems. AI's success in tackling complex problems across various domains, from military defense to natural language processing, is a testament to its capabilities. Traditional methods struggle with the growing complexity of network performance and environments. Designing and optimizing wireless communication systems becomes increasingly challenging, demanding advanced methods and algorithms. AI, which learns from its surroundings and vast datasets, is a proven and robust method for handling such complexity.

Researchers believe AI can address the unprecedented data traffic and user demands of future networks. AI algorithms can dynamically adjust network protocols and resource management to optimize performance. This tackles current limitations and paves the way for improved performance in 5G and 6G networks, impacting various aspects of our lives.

AI's practical applications in wireless communication are diverse, encompassing resource management, signal processing, and channel modeling. These applications play a vital role in optimizing physical layer design, network management, and resource allocation. As referenced in [9] (Figure 2), AI advancements extend beyond 5G, impacting physical layer research, network optimization, channel measurements, algorithm development, and standardization. The combination of AI and machine learning (ML) has the potential to revolutionize communication systems beyond 5G. Their ability to adapt, learn from data, and make real-time decisions empowers them to address complex and evolving challenges. Ultimately, this leads to more robust, efficient, and user-centric communication systems.

The key strengths of AI methods lie in their ability to:

- Predict and detect network performance for better scheduling.
- Model systems more accurately than conventional methods, leading to intelligent approaches.
- Open doors for constructing updated models based on evolving traffic patterns.

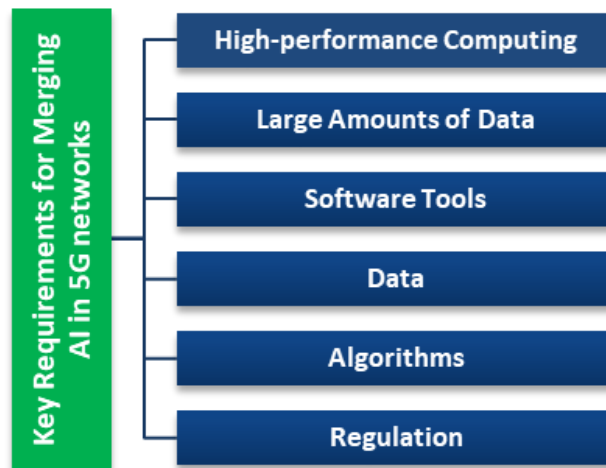
These methods are generally used to optimize design specifications and find optimal solutions quickly, similar to their applications in manufacturing for safer processes, cost reduction, and revenue growth.

## **5. KEY REQUIREMENTS FOR IMPLEMENTING AI IN 5G NETWORKS**

The hardware and software requirements for implementing AI in 5G networks vary depending on the specific application. However, some general requirements as shown in Fig.3 include: (i) High-performance computing: AI-based algorithms can be computationally intensive, so it is important to have high-performance computing resources available to train and deploy AI models. This can include dedicated hardware accelerators, such as GPUs and FPGAs, as well as high-performance CPUs. (ii) Large amounts of data: AI models require copious amounts of data to train. This data can come from a variety of sources, such as network traffic logs, sensor data, and user behavior data. AI-

based algorithms require copious amounts of data to train and improve their performance. (iii) Software tools: A variety of software tools are available to help with the development and deployment of AI in 5G networks. These tools can help with tasks such as data collection, data preparation, model training, and model deployment. (iv) Algorithms: There are a variety of AI-based algorithms that could be employed in 5G networks. The specific algorithm that is used will depend on the specific application. For example, machine learning algorithms could be employed for tasks such as network slicing and security, while deep learning algorithms could be employed for tasks such as image recognition and natural language processing. (v) Expertise: Implementing AI in 5G networks requires a deep understanding of both AI and networking. It is important to have a team of experts who can work together to design, develop, and deploy AI solutions for 5G networks. (vi) Regulation: There are several regulations that need to be considered when implementing AI in 5G networks. These regulations vary from country to country, so it is important to understand the specific regulations that apply to your network.

In addition to these general requirements, there are also some specific hardware and software requirements for certain AI applications in 5G networks. For example, applications that require real-time inference, such as network security and traffic management, will need to have hardware accelerators that can support low latency. Applications that require copious amounts of data storage, such as machine learning for recommendation systems, will need to have high-capacity storage systems. The hardware and software requirements for implementing AI in 5G networks are constantly evolving as innovative technologies become available. It is important to stay up to date on the latest trends in AI and 5G to ensure that your networks are using the most efficient and effective technologies. However, these are just a few examples of the many ways that AI could be employed in 5G networks. As 5G continues to evolve, AI is likely to play an even greater role in improving the performance, security, and user experience of 5G networks.



**Fig.3.** Applications of AI in 5G Networks.

## 6. SIMULATION TOOLS

This section provides a high-level overview of simulators that could be employed to model 5G networks with AI-based algorithms. There are several simulation tools available for AI and 5G. Some of the most popular tools include:



### **6.1. NS-3**

NS-3 is a free and open-source network simulator that could be employed to simulate a variety of network topologies and protocols. It is a popular choice for simulating 5G networks, as it supports a wide range of 5G features, such as millimeter wave (mmWave) communications and network slicing.

### **6.2. OMNeT++**

OMNeT++ is another popular open-source network simulator that is like NS-3. It supports a wide range of network protocols and topologies, and it is also a desirable choice for simulating 5G networks.

### **6.3. MATLAB**

MATLAB is a commercial software package that could be employed for a variety of tasks, including simulation. It has several built-in functions for simulating 5G networks, and it can also be used to develop custom simulation models.

### **6.4. Python**

Python is a general-purpose programming language that can also be used for simulation. There are several Python libraries available for simulating 5G networks, such as ns-3-python and omnetpp-python.

These are just a few of the many simulation tools that are available for AI and 5G. The best tool for a particular project will depend on the specific requirements of the project. In addition to these general-purpose simulation tools, there are also several specialized simulation tools that are designed for specific aspects of AI and 5G. For example, there are tools for simulating machine learning algorithms, tools for simulating wireless networks, and tools for simulating network security. The use of simulation tools is essential for the development and deployment of AI and 5G networks. By simulating different network topologies and protocols, engineers can assess and evaluate modern technologies before they are deployed in the real world. This helps to ensure that the networks are reliable and efficient, and that they can meet the needs of users.

## **7. CHALLENGES AND FUTURE RESEARCH DIRECTIONS**

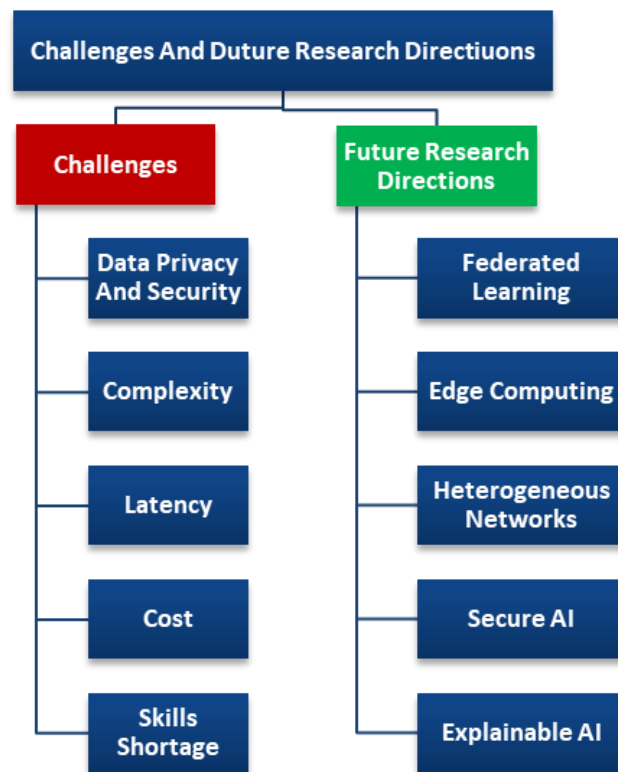
There are several challenges [45-50] for implementing AI in 5G networks. (i) Data privacy and security: AI models require copious amounts of data to train. This data can be sensitive, such as user location data or financial data. It is important to protect this data from unauthorized access, and to ensure that it is used in a responsible manner. (ii) Complexity: AI-based algorithms can be complex and difficult to understand. This can make it difficult to debug and troubleshoot AI-powered systems. It is focal to have a good understanding of how AI works to deploy it successfully in 5G networks. (iii) Latency: AI-based algorithms can be computationally intensive, which can lead to latency issues in 5G networks. This is especially true for applications that require real-time inference, such as network security and traffic management. It is critical to choose AI-based algorithms that are designed for real-time performance, and to ensure that the network has the necessary resources to support them. (iv) Cost: The cost of implementing AI in 5G networks can be high, due to the need for

high-performance computing resources and specialized software. It is vital to carefully consider the cost of AI before deploying it in 5G networks. (v) Skills shortage: There is a deficiency in skilled engineers with the knowledge and experience to implement AI in 5G networks. This can make it difficult to find and hire the talent needed to develop and deploy AI-powered solutions. It is important to invest in training and development programs to help engineers learn the expertise they need to work with AI in 5G networks.

Despite these challenges, there are several opportunities for AI in 5G networks. AI could be employed to improve the performance, security, and user experience of 5G networks. As 5G continues to evolve, AI is likely to play an even greater role in the development and deployment of 5G networks [51]. These are just some of the challenges that need to be addressed to successfully implement AI in 5G networks. However, the capability benefits of AI are significant, and it is likely that AI will play an increasingly significant role in the development and deployment of 5G networks in the future. There are some future research directions [52-58] for implementing AI in 5G networks:

- Edge computing: This distributed computing paradigm moves storage and computation closer to the point of usage. Because less data needs to be sent to the cloud, this can be leveraged to improve the latency of AI applications on 5G networks [56].
- Federated learning: Federated learning is a ML technique which permits devices to train machine learning models without sharing their data with a central server. This could be employed to enhance the privacy of AI models in 5G networks, as it prevents the network operator from collecting and storing user data [57].
- Heterogeneous networks: These networks are made up of many diverse kinds of devices, including computers, cellphones, and sensors. Because AI can learn the behaviors of many devices and adjust the network accordingly, it may be used to manage and improve heterogeneous networks [58].
- Explainable AI: The goal of this field of study is to create AI-based algorithms that can provide an explanation for their choices. This holds significance for 5G networks, as it enables network operators to discern potential biases and comprehend the decision-making process of AI models [59].
- Secure AI: Research in this area aims to create AI-based algorithms that are impervious to cyberattacks. This is significant for 5G networks since the network will be managed and controlled by AI models [60-62].

Finally, these mentioned points are just some of the future research directions for implementing AI in 5G networks. Since AI technology continues to develop, it is likely that new and innovative ways to use AI in 5G networks will be discovered. Figure 4 summarizes Challenges and Future Research Directions of AI in 5G Networks.



**Fig.4.** Challenges and Future Research Directions of employing AI in 5G-Networks.

## 5. CONCLUSIONS

AI is a powerful tool that could be employed to improve a variety of aspects of 5G networks. By using AI, network operators can improve performance, security, and the user experience. As 5G continues to evolve, AI is likely to play an even greater role in the development and deployment of 5G networks. However, there are a few challenges that need to be addressed to successfully implement AI in 5G networks. Some of the most significant challenges include data privacy and security, complexity, latency, cost, and skills shortage. the challenges and future research directions for integrating AI into 5G networks. AI faces many challenges to be employed in 5G networks include data privacy and security, complexity, cost, lack of skilled personnel, secure AI, edge computing, managing heterogeneous networks, and ensuring explainable AI. There are also corresponding research areas to address these challenges. Federated learning is a possible solution to data privacy and security concerns. Complexity can be reduced through edge computing. Research into heterogeneous networks can improve how different networks work together. Finally, there is a need for continued development in secure AI and ensuring AI models are understandable (explainable AI).Despite these challenges, there are a few opportunities for AI in 5G networks. AI could be employed to improve the performance, security, and user experience of 5G networks. As 5G continues to evolve, AI is likely to play an increasingly significant role in the development and deployment of 5G networks. Moreover, there are some concluding remarks about implementing AI in 5G networks. First, AI has the potential to revolutionize 5G networks, but it is important to address the challenges before it can be widely adopted. Second, Federated learning, edge computing, heterogeneous networks, secure AI, and explainable AI are some of the future research directions for implementing AI in 5G networks. Finally, as AI technology continues to develop, it is likely that new and innovative ways to use AI in 5G networks will be discovered.

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