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### A COMPREHENSIVE REVIEW AND CRITICAL ANALYSIS OF METHODOLOGIES, LIMITATIONS, AND CHALLENGES IN IOT-5G INTEGRATION

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#### ABSTRACT

The integration of Internet of Things (IoT) and 5G technologies promises to revolutionize various industries by enabling ultra-reliable, low-latency, and high-throughput communication for an unprecedented number of connected devices. However, this convergence presents several challenges, including scalability, security, energy efficiency, and seamless interoperability. This paper provides a comprehensive review and critical analysis of existing methodologies for IoT-5G integration, highlighting key design frameworks, protocols, and architectural models. We explore various approaches to ensure efficient data transmission, resource management, and network optimization, emphasizing their performance metrics and limitations. Furthermore, we examine the primary challenges faced during integration, such as interference management, quality of service (QoS) assurance, security vulnerabilities, and the handling of heterogeneous devices and networks. Aim to identify research gaps and suggest potential solutions that could address the limitations and enhance the reliability and scalability of IoT-5G systems in real-world applications. However, with the advent growth in the number of IoT devices the signaling overhead in the IoT systems will grow exponentially. A novel queuing model based scheduling scheme is also proposed to overcome the limitations of the existing scheduling schemes and also to include the proposed protocol stack.

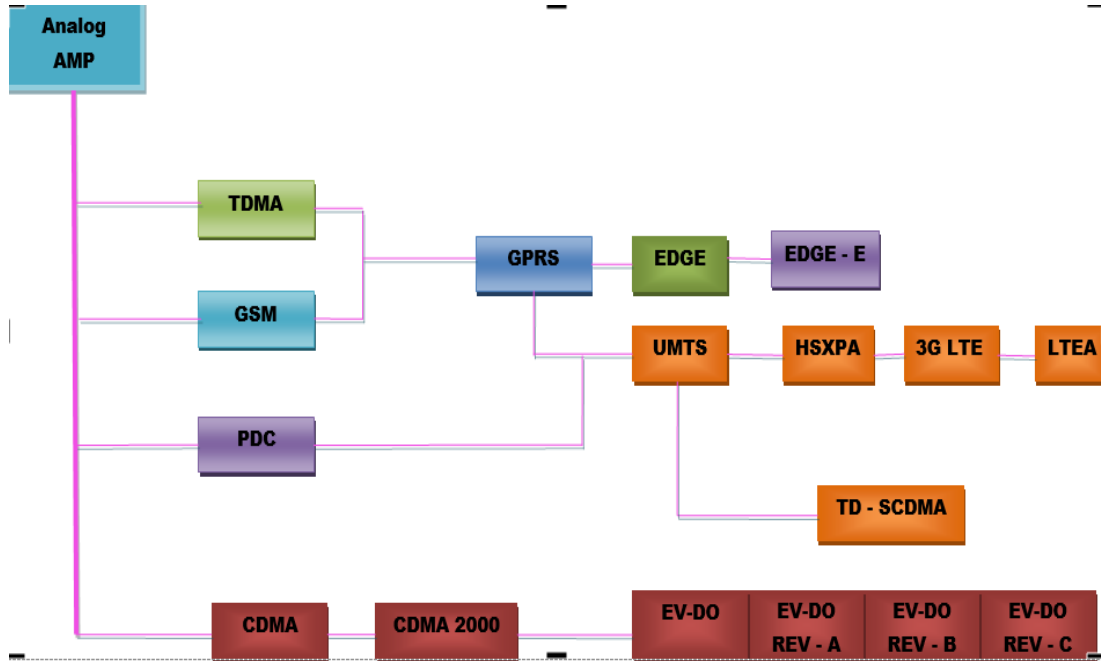
#### Introduction

The integration of the Internet of Things (IoT) with 5G technology represents a significant leap forward in the evolution of connected systems, offering unprecedented opportunities for data transmission, connectivity, and automation across a wide range of applications. As IoT devices proliferate, the need for a high-speed, low-latency, and highly reliable network becomes essential, making 5G an ideal enabler. This review aims to provide a comprehensive analysis of the methodologies employed in the IoT-5G integration, highlighting the key technological advancements such as network slicing, edge computing, and massive machine-type

communication. In addition, it critically examines the limitations and challenges faced during the convergence of these two technologies, including concerns related to scalability, security, energy consumption, and the need for efficient resource management. By delving into both the opportunities and obstacles associated with this integration, this paper seeks to provide a holistic understanding of the current landscape and offer insights into the path forward for IoT-5G convergence in various sectors [1].

Analog interface technology was used over circuit switched networks and supported voice-only capabilities in these first generation systems. Only very limited features were available for 1G cellular system and they suffered poor voice quality with limited radio coverage yet a new era of personal communication began with these systems. The second generation (2G) of wireless cellular systems was introduced in the early 1990's with the increased requirement for cellular services and need for enhanced quality and more features. The 2G systems characterized digital air interface and mainly voice-centric technology utilizing circuit-switched network however it provided higher bandwidth, better voice quality and limited data services. The 2G systems achieved remarkable popularity and were successfully set out across the globe. Increased bitrates were introduced later by the enhanced second-generation [2].

The 2.5G systems were developed to support data rates from 57.6 Kbps to 171.2 Kbps. The grand success of these systems and the express growth of internet resuIoT-5Gd in the need for systems with higher capacity, better Quality of Service (QoS) support than 2G and 2.5G wireless systems. This solidly assured the evolution of 3G wireless cellular systems. The third generation system guaranteed wide-area coverage at 384 Kbps and local area coverage up to 2 Mbps, which qualified it to support new data services and enhance those supported by current 2G and 2.5G systems. The 3rd Generation Partnership Project (3GPP) (Karenina et al 2005) developed the Universal Mobile Telecommunications System (UMTS), one of the most famous 3G systems. UMTS assured a transmission rate of up to 2 Mbps, which could support new data services and enhance those supported by current 2G systems.



**Figure 1: Evolution of Wireless Standard**

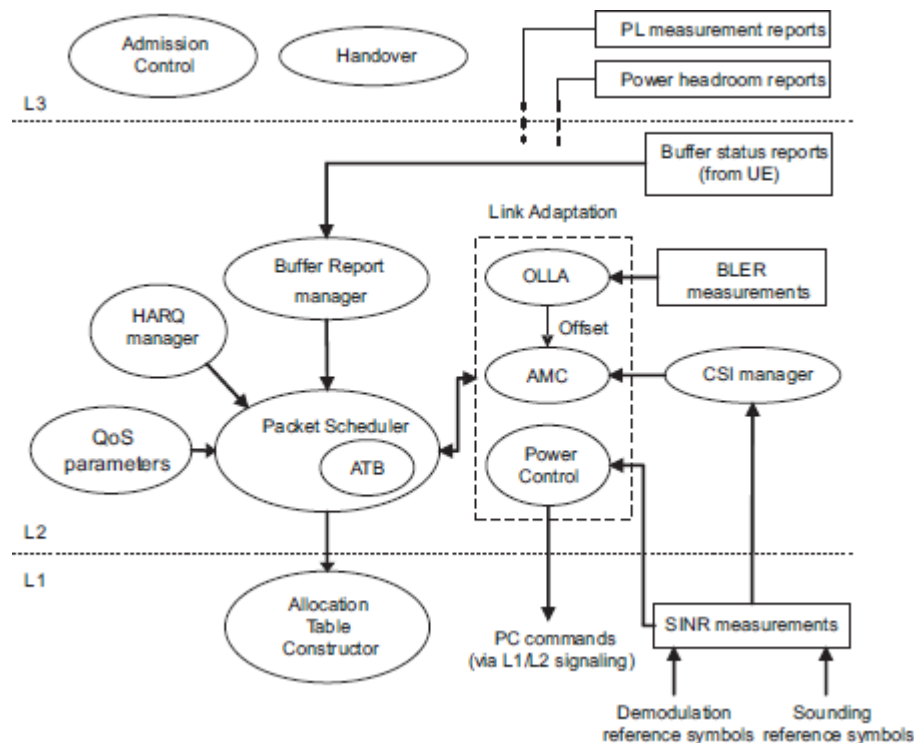
Prophesies for emerging mobile wireless markets, however, anticipated that services like multimedia on demand will compress the bandwidth. The need for data rates beyond what is offered by current 3G wireless systems was prompted by this. Broadband Wireless Access Systems (BWASs) have been developed to enhance the support for such high data rates. For instance, the High Speed Downlink Packet Access (HSDPA) (3GPP TS 25.308 2003), a 3.5G BWAS was standardized by 3GPP as an annex to the existing 3G UMTS. Data rate up to 14.4 Mbps was supported by HSDPA hypothetically which was 7 times larger than that of UMTS. IEEE 802.16 group (IEEE 802.16-2004), (IEEE 802.16e 2005) standardized the Worldwide Interoperability for Microwave Access (WiMAX), another BWAS which supported up to 70 Mbps.

The customization of versions of both 3GPP and 3GPP2 beyond 3G systems was resulted by the development of IEEE 802.16 standards [1]. The proposed systems are all based on the OFDMA technology; however the network architecture is akin to that of the IEEE 802.16 standard. Evolved universal terrestrial radio access (evolved-UTRA) (3GPP TSG RAN TR 25.912 v7.2.0), the beyond 3G system in 3GPP is also widely known as Long-Term Evolution (LTE) whereas 3GPP2's version is referred as ultra mobile broadband (UMB) (3GPP2 TSG

C.S0084-001-0 v2.0). Mobile WiMAX, IOT-5G and UMB, all these three beyond 3G systems meet IMT-2000 requirements and therefore make a part of IMT-2000 family of standards.

As an aIoT-5G native for dedicated data transmissions, shared channels facilitated the support for higher number of mobile users, consequently reducing the per-bit transmission costs and improving utilization of the system. Their varying channel quality conditions, diverse QoS requirements and anticipated high traffic demands yet remained to be a challenge in sharing such channels. Hence to fulfill the QoS requirements of existing and new multimedia services and to maximize the system throughput simultaneously, more cautious and efficient bandwidth management and resource sharing techniques is required by the IOT-5G [2].

New mobile stations (MS) were granted access to meet the Quality of Service (QoS) requirements of incoming calls only when sufficient bandwidth was available in the network. The scheduling procedure in this system was referred to as Class-level QoS provisioning, while the process of resolving contention for shared network resources was known as the packet scheduling procedure. This procedure involved allocating bandwidth among users and determining the transmission order. The selection of scheduling algorithms was based on the types of users in the network and their respective QoS requirements. To ensure each user's QoS needs were met, the number of transmission time frames assigned to each traffic class was carefully determined.



**Figure 2: Interaction between various layer**

By the introduction of Internet of Things (or IoT) and its impact on everything from the way we shop during travel to the way manufacturers keep track of inventory, it had a huge impact in the society and soon it necessitates the creation of a dedicated infrastructure. GSMA predicts that the Internet of Things (IoT) traffic will grow exponentially in the years to come. An estimated 24 billion interconnected devices will come to existence by 2020 [1]. Further, IoT devices are expected to generate Machine Type Traffic which will be quite different from the Human Type Traffic [2]. Machine Type traffic is usually small in data size and the devices are mostly inactive and they actually come into existence or transmit only when there is data to transmit. These patterns will create a new culture of data traffic in the future wherein a large amount of traffic will be generated by the control plane which in turn will cause huge congestion in evolved node B (eNB) and to mobility management entity (MME) [3].

## **MOTIVATIONS AND OBJECTIVES**

The channels quality does not remain the same for all the MS. Those MS located close to Base Station enjoy better channel quality conditions and on other hand, those located at farther from BS suffer a poor channel qualities which remain a generic feature in wireless communication. However, it is not the fault of MS being located far away from BS. Other factors namely, interference from other users, obstacles, etc also influences the channel quality of the MS. The typical function of the bandwidth management schemes is to estimate the channel quality conditions and provide resources..

Network congestion is another major challenge faced by the bandwidth provisioning mechanisms as IOT-5G systems are designed basically to handle multimedia traffic which makes the network congested easily which is a serious issue to consider. Simultaneously we have to handle different users with different bandwidth requirements and different services which complicates the task of the scheduler even more. Revenue loss is the last issue of bandwidth management in IOT-5G due to serving low- revenue-generating users. Different levels of revenue losses are faced by Network operators from serving users based on their channel quality conditions, the amount of the buffered data they have at the base station and their willingness to pay for different services.

### **A Simplified Protocol Stack for improved IoT support in IOT-5G**

A novel protocol architecture, namely, Reduced Control Plane Protocol(RRCP) architecture suite to effectively reduce the number of signaling messages or reduce the duplication of signaling messages when an User Equipment (UE) tries to join the IOT-5G network from its sleep mode. We evaluate the performance of the proposed scheme with conventional scheme in terms of many metrics which include, number of channels allocated, delay and energy consumed and prove that the proposed architecture is much better than the existing conventional architecture.

## **Literature review**

Zhang et al. (2019) In their 2019 paper, Zhang et al. discussed the integration of 5G networks with IoT, emphasizing how 5G's ultra-low latency and high bandwidth capabilities are essential for enhancing IoT applications, particularly in smart cities and industrial automation [4]. They proposed a framework involving network slicing to ensure efficient resource allocation and management in heterogeneous IoT environments. However, they identified scalability as a significant challenge, highlighting the difficulty in managing large numbers of devices and maintaining consistent performance across varying network conditions.

Liu et al. (2020) provided a detailed overview of the methodologies for integrating IoT with 5G, focusing on edge computing as a means to process data closer to the source and reduce latency. Their research emphasized the critical role of edge nodes in managing real-time data streams from IoT devices, proposing a multi-tier architecture to enhance system performance. However, they also pointed out limitations related to the complexity of deploying edge computing infrastructures and the security risks inherent in distributing computing resources across diverse locations.

Chen et al. (2021) In their 2021 work, Chen et al. conducted an extensive review of the security challenges faced in IoT-5G integration. They identified a number of vulnerabilities, including data breaches, device spoofing, and denial-of-service attacks, that could potentially compromise the performance and reliability of the integrated network. They suggested that AI-driven security mechanisms and blockchain technology could offer solutions to these challenges, but acknowledged the difficulty in ensuring interoperability across different IoT devices and 5G network configurations.

Singh et al. (2021) focused on the challenges of ensuring energy efficiency in the IoT-5G ecosystem. They discussed energy consumption as one of the most pressing issues in large-scale IoT networks, where the density of connected devices could significantly impact power usage. Their paper proposed the use of advanced energy harvesting techniques and low-power wide-area networks (LPWANs) to address this challenge. Despite these efforts, they highlighted the difficulty of achieving sustainable energy solutions without compromising the performance of real-time applications.

Gupta et al. (2022) conducted a study on the role of AI and machine learning in optimizing IoT-5G integration. They explored how AI algorithms could be used for resource allocation, network management, and predictive analytics to improve the efficiency and reliability of the integrated system. However, the authors also pointed out that the complexity of implementing AI at scale, especially in diverse IoT environments, creates new barriers in terms of computational resources and training data requirements [5].

Wang et al. (2023) In their 2023 review, Wang et al. provided an updated analysis of the IoT-5G integration landscape, focusing on the regulatory and standardization challenges. They noted that the rapid development of both IoT and 5G technologies has outpaced the development of standardized protocols, leading to interoperability issues between devices and networks. Furthermore, the lack of a unified global regulatory framework complicates the deployment of IoT-5G solutions across different regions. Their work underscored the need for international collaboration to establish consistent standards that facilitate smoother integration.

Kumar et al. (2023) reviewed the advancements in network slicing and its impact on IoT-5G integration. They argued that network slicing allows for better resource management and customized service delivery for different types of IoT applications. However, the authors acknowledged that challenges remain in terms of effectively managing slice allocation, ensuring fairness, and preventing resource over-provisioning. They called for more research into dynamic slice management techniques to address these concerns.

Al-Fuqaha et al. (2024) provided a comprehensive analysis of the potential of 5G for enabling smart healthcare applications. They discussed how 5G networks could facilitate real-time monitoring of patients through IoT devices, offering faster transmission of medical data and improving overall healthcare delivery. However, they raised concerns about data privacy, security, and the robustness of the system in case of network failures or overloads, emphasizing the need for better disaster recovery strategies.

Zhang and Li (2024) reviewed recent advancements in IoT-5G integration methodologies with a focus on machine-type communications (MTC). Their study highlighted the potential of 5G to support massive MTC applications, such as smart agriculture and smart cities. They identified challenges such as network congestion, data interference, and the strain on 5G resources caused



by a high density of connected devices [6]. They proposed the use of AI-based dynamic network resource management techniques to address these challenges.

## **IOT-5G-A**

For the sake of conceptual clarity, we first introduce the IOT-5G-A systems in brief in order to understand the factors influencing the scheduling of the system [3]. Figure 2 Illustrates the IOT-5G-A advanced architecture with the network elements. The network elements include the MME, SGW PGW and eNodeB.

### **SGW-Serving Gateway**

The SGW is a single serving gateway that acts as a local anchor in order to facilitate handover between two eNodeB devices. It also takes care of other operations like packet routing, packet forwarding and buffering the downlink packets. SGW is also responsible for downlink packet marking.

### **PGW-PDN Gateway**

PGW is responsible for all the IP packet-based operations. This includes deep packet inspection, UE IP address allocation, Transport level packet marking in uplink and downlink, accounting etc.

### **eNodeB**

An eNodeB is an element of IOT-5G Radio Access Network that's responsible for allocating the radio resources. It is the key node where all the scheduling happens [7].

### **Carrier Aggregation**

Carrier Aggregation (CA) is a technique which is used by network service providers to combine a number of separate IOT-5G RF carriers. CA enables the users to increase the peak user data rates and overall capacity of their networks and to exploit fragmented spectrum allocations. CA may be employed on both on FDD and TDD based IOT-5G systems. It accomplishes this by aggregating the two or more component carriers up to a maximum of five component carriers. CoMP aims at improving network performance at cell edges.

The strategy is usually of two types, namely, joint transmission and dynamic point transmission. In both the cases the transmitter and receiver communicate effectively to improve the quality for downlink and uplink respectively but in joint transmission all transmitters and receivers

transmits/receives all sub frames. In dynamic point scheduling data is transmitted/ received only from one transmitter/receiver point at a time. This might cause considerable delay and additional resources must be provided for signaling. They can be located close to each other or they can be at distinctive positions. FDD and TDD. The number of carriers aggregated may differ in UL and DL provided the number of DL carriers is always more than uplink carriers. The aggregation can be done on both inter-band and intra-band frequencies.

### **Parameters impacting scheduling**

MAC scheduler on the eNodeB is responsible for effectively allocating the radio resources to all the physical devices connected to the network. The scheduler takes into account various parameters before it takes this decision. Changes in the values of the parameters are constantly monitored and the feedback is fed to the scheduler as variations in values of these parameters can affect the scheduling process significantly. Some of the physical layer parameters that are constantly monitored include Channel Conditions, retransmission, and number of Physical Resource blocks (PRB) [9] allocated for a particular UE and the transfer buffer status.

### **Channel State Information (CSI)**

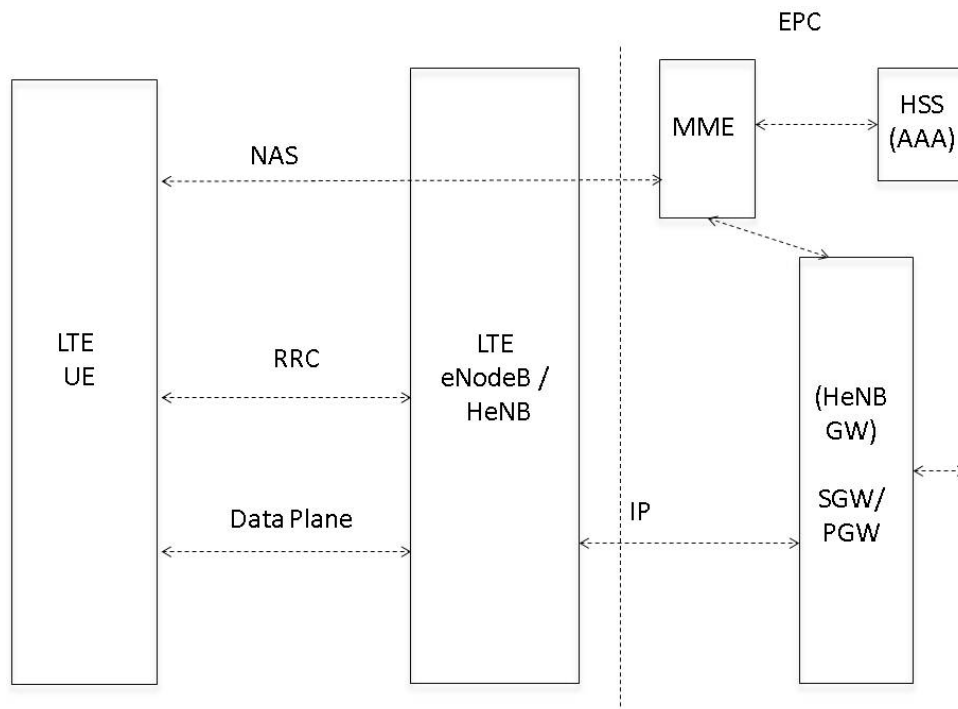
The CSI of the IOT-5G-A is defined by three parameters, namely, CQI, PMI and RI. The UE shall measure all the three parameters and regularly update the same to the eNodeB. The UE then modifies the necessary scheduling schemes and transmits the signals to suit the conditions [10].

The channel quality directly impacts two factors. First, it provides the status of the current Communication Channel Quality and second the quality of the channel directly translates into throughput. The typical CQI value ranges from 0 ~ 30. 30 indicates the best channel quality and 0, 1 indicates the poorest channel quality. Depending on which value the UE reports, the network transmits data with different transport block sizes. If the network gets a high CQI value from the UE, it transmits the data with a larger transport block size and vice versa. The UE carries out the CQI based on signal-to-noise ratio (SNR), signal-to-interference plus noise ratio (SINR) and signal-to-noise plus distortion ratio. The channel quality also defines the modulation scheme of the system.

### **Methodology**

An IOT-5G system is an IP enabled architecture which has many features, including, high security and spectral efficiency which makes it very suitable for IoT services [11]. Internet of Things systems play a vital role in defining the future of various Machine-to-Machine (M2M)

commitments such as energy, transportation, predictive maintenance, logistics, medicine and smart home systems and hence various potential systems may require the help of IOT-5G systems or dedicated IoT frameworks for their maintenance and survival. Other potential applications could include connected vehicles, smart grids, smart cities and seamless integration of wireless sensor networks. In all these cases it becomes imperative for the devices to reduce its computations in order to reduce the load on the cloud applications and thereby improve its computational capability. Even though the IOT-5G systems do have enough spectrum, the number of IoT devices in each cell expected to grow to more than 1,00,000 by 2020 which will create a huge bottleneck for cloud offloading which will limit the performance of these systems.



**Figure 3: Conventional IOT-5G Architecture**

Whenever an IoT device attempts to join the IoT-5G network, it must undergo a mutual authentication procedure known as Evolved Packet System Authentication and Key Agreement (EPS-AKA). The standard EPS-AKA process is illustrated in Figure 3. This procedure follows a shared key approach, where the shared key 'K' is securely stored in both the Universal Subscriber Identity Module (USIM) of the IoT-5G User Equipment (UE) and the Authentication Centre (AuC) within the Home Subscriber Server (HSS). Upon powering on, the IoT device initiates basic network joining procedures, after which the EPS-AKA process is executed to complete the authentication and network access.

As multiple IoT devices attempt to connect to the IoT-5G network, the control plane experiences increased overhead. Additionally, some IoT devices may remain inactive for extended periods, further impacting network resource management.

**Table 1: List of Keys associated with IOT-5G EPS-AKA PROCESS**

Key(s)	UE	EPCandIOT-5GeNodeB
K(shared key)	USIM	AuC/HSS
Apair of keys (CK, IK)	UE	AuC/HSS
K <sub>ASME</sub>	UE	HSS/MME
K <sub>NASenc</sub> , K <sub>NASint</sub>	UE	MME
K <sub>eNB</sub>	UE	(MME derives and provides to) eNodeB
K <sub>UPenc</sub> , K <sub>RRCint</sub> , K <sub>RRCenc</sub>	UE	(Computed at) eNodeB

The HSS then generates one or more Authentication Vectors (AVs) for the UE i.e., for the specified IMSI. The AV is generated by the HSS is developed using a shared key K, a random challenge RAND and the sequence number SQN associated with that IMSI and SN. Each AV consists of a key K<sub>ASME</sub>, a random challenge R<sub>AND</sub>, a network authentication token AUTN which is generated by the Home Subscriber Server and Expected Response. A Key Set Identifier, KSI<sub>ASME</sub>, is assigned by MME. It uniquely identifies K<sub>ASME</sub>. The authentication procedure between the UE and EPC is also based on the AV.

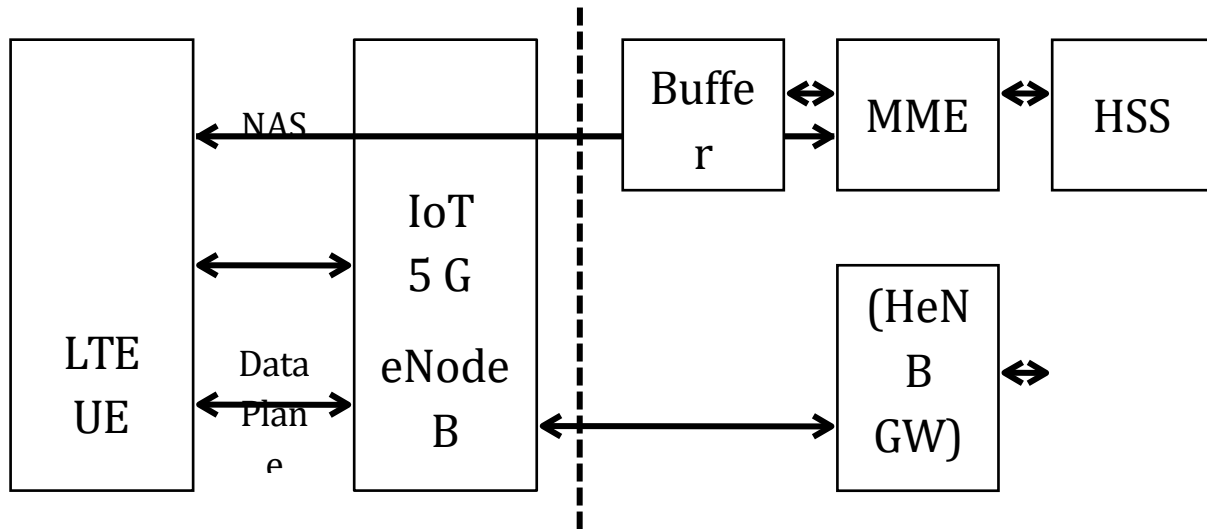
The Buffer should ideally be a trusted partner of the network service provider. Our proposed lightweight Attach procedure works in three broad stages. The proposed RCPP scheme does not hold good for new UE which try and join the network for the first time but the scheme works effectively when an UE returns from the SLEEP mode. The procedure may be divided into three parts. The first part is that the UE attempts a normal Attach procedure. In the second stage, some of the messages that were intended to the eNodeB will be intercepted by the buffer and the Attach procedure will be carried out between buffer and MME. This avoids duplication and avoiding multiple messages over the air-interface between UE and eNodeB. In the third stage, authentication between the UE and EPC is carried out. The context of discussion of the RCPP protocol is limited only up to the ATTACH procedure. Albeit the Buffer is able to emulate the

operations of the UE it is limitedly till the Attach is completed. The rest of the operations is completely controlled by the network service provider and the UE does not have rights to take control over it [11].

The buffer extracts the IMSI from the message. The buffer then pings the MME with the IMSI and  $KSI_{ASME}$  of the UE. This happens with the help of a new secure channel. The buffer requests the key,  $K_{ASME}$  and key  $K$ , for the IMSI from MME. Once both the keys are received by the buffer new values are assigned by the MME. This again happens over the same security channel that was mentioned earlier. The key  $K$ , may leave the attach procedure but it should be noted that the buffer is still under the control of the network service provider. From now on whenever, the same UE wants to join the network the AUTN message is generated by the buffer. A verification in terms of the validation of the past and the present AUTN messages id also carried by the buffer.

A modified attach message may be employed to do the same. This will induce some changes in the functions of the MME. Now the MME will start to its NAS security procedure. The MME will issue the NAS Security Mode Command and this message will traverse to the UE. IN the proposed scheme, the traversing message will be intercepted by the buffer and also updated on the details of the message [12].

In the generic IOT-5G architecture whenever an IoT device enters the sleep mode, its entire settings is removed from eNodeB. All the security keys are removed from the eNodeB as well. Where as in the proposed RRCP protocol the settings are not completely removed. The MME keeps part of the setting of the UE. The  $K_{ASME}$  key is stored in the buffer and if the gateways want to use the same keys the buffer does not have to fetch them from the MME again. The NAS counters are also stored in the buffer.



**Figure 4: Proposed RCPP Protocol Architecture**

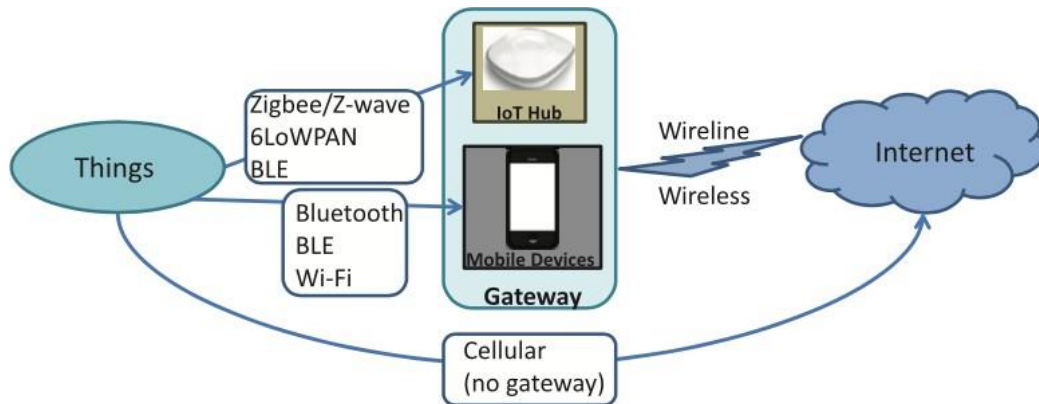
### Result analysis

Machine Type traffic is usually small in data size and the devices are mostly inactive and they actually come into existence or transmit only when there is data to transmit. These patterns will create a new culture of data traffic in the future wherein a large amount of traffic will be generated by the control plane which in turn will cause huge congestion in evolved node B (eNB) and to mobility management entity (MME) [59]. This in turn shall increase the cost and overhead of the network service providers because the signal traffic shall considerably increase [13].

In IoT-5G networks, there are two radio ON/OFF states: RRC\_Connected and RRC\_Idle. In conventional traffic patterns, data transmission occurs only in the RRC\_Connected state, leading to significant power savings. However, IoT traffic patterns differ significantly, as IoT-5G networks frequently switch between RRC\_Connected and RRC\_Idle states due to the nature of IoT devices, which generate sudden bursts of data. This frequent switching causes excessive signaling load on the network, making it essential to carefully manage the ON/OFF states in IoT-5G to suit IoT traffic efficiently.

If the scheduling algorithm operates exclusively in the RRC\_Connected mode, the scheduler can reduce the number of transitions between RRC\_Connected and RRC\_Idle states. However, this approach would lead to increased power consumption, rapidly draining the battery of IoT devices. Conversely, if the scheduling pointer is set to RRC\_Idle mode, IoT devices will require frequent channel setups to transmit their short, periodic bursts of data. This increases network latency and degrades performance, particularly for delay-sensitive applications such as medical health monitoring, surveillance, and vehicular communication.

We propose a Dynamic Channel-Aware / QoS-Unaware Uplink Scheduling (DCA) algorithm, which can be implemented in the eNodeB of the IoT-5G system. This algorithm is designed to enhance network performance, ensuring efficient handling of IoT traffic while maintaining an optimal balance between power consumption, bandwidth utilization, and latency.



**Figure 5: Network Architecture of IoT**

When the device are connected using a cellular network they consume more power. On the other hand, if they are connected to the internet using a gateway or what is in other words is called IoT hubs then the hubs are connected to a power source. Some of the popular IoT protocols that are used by the IoT devices include ZigBee, Blue tooth low Energy, Wi-Fi, Z-wave and sometimes even mobile gateways are used.

**Table 2: IoT Traffic Types**

Type	Description	Model	Applications
Periodic Update(PU)	Updates are sent by the devices at regular intervals	Periodic ON/OFF	Household devices like monitoring of water, gas etc
Event-Driven(ED)	Updates are sent as when an event happens	Poisson process	Alerts, health monitoring
Payload Exchange (PE)	Data is sent after device-triggered events	Interrupted Poisson process	Surveillance systems

Various Discontinuous Reception (DRX) selection algorithms combined with packet schedulers have been proposed in the literature to meet the QoS requirements of IoT-5G networks while reducing power consumption. However, most existing mechanisms primarily focus on power saving while overlooking the massive signaling load generated by IoT devices. Additionally, much of the research in the literature emphasizes downlink scheduling, despite the fact that IoT traffic is predominantly uplink-oriented.

**Table 3: Simulation parameters**

Parameters	Values
Bandwidth	10MHz
Total number of sub-carriers used per slot	600
Number of sub-carriers used per PRB	12
Number of available used per PRB	50
Subcarriers spacing	15khz
Slot Duration	0.5ms
TTI Duration	2 slots=1ms
Frame Duration	10ms



Cellular layout	1 cell
Cell radius	1500m
Height of the eNodeB antenna	32m
Pathloss model	Cost HATA231
Shadow fading	Log-normal standard deviation(8dB)
Fast fading	Rayleigh Distribution
Total eNodeB transmission power	46dBm
Thermal noise density	-174dBm/Hz
Number of users	10-250
Position of users	Random
MCSs of users	Selected using link adaption 1
The distribution of users for each traffic	6
Resources allocation decision	Per TTI
Adaptive threshold resource Utilization	GBR=80%
Simulation duration	1000 TTIs

The proposed scheme is able to demonstrate an increased throughput of 19.8 %, 35.5%, and 52% compared to the UARRA, ERMS and ARRA schemes, respectively, especially under heavy traffic conditions [15]. This is because the traffic is modelled as a Poisson Process. The scheduler takes the decision based on a tradeoff between three parameters, namely, power consumption delay and signal load and the optical static parameters depending on the traffic type is chosen to increase the throughput.

## **Conclusion**

In conclusion, the integration of Internet of Things (IoT) and 5G networks presents a transformative opportunity for various industries, enabling faster, more reliable communication and enhanced data processing capabilities. However, this integration is not without its challenges. Methodologies such as network slicing, edge computing, and enhanced communication protocols offer promising solutions, but issues such as interoperability, security, scalability, and energy efficiency remain significant hurdles. The complexity of managing diverse IoT devices within a 5G framework, coupled with the potential for network congestion

and latency, necessitates continued research and development. Moreover, the evolving standards and regulatory frameworks pose additional barriers to seamless implementation. While the potential benefits of IoT-5G convergence are immense, addressing these challenges through innovative solutions, robust security measures, and standardized approaches will be crucial to unlocking the full potential of this integration.

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