

# Optimizing Resource Allocation in 5G Networks: A Network Slicing Approach with CNN-Based Femtocell Management

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## Abstract:

The swift advancement of 5G networks demands creative solutions to address the dynamic and diverse requirements of modern wireless communication. In this investigation, we delve into an innovative approach to improve resource allocation by integrating Convolutional Neural Network (CNN)-based femtocell management with network slicing techniques. The suggested system takes advantage of the intrinsic flexibility of network slicing to create isolated and tailored slices customized to meet specific application demands. The inclusion of CNN algorithms in femtocell management amplifies the adaptability and intelligence of the system. By conducting thorough simulations and real-world experiments, our study evaluates the performance enhancements derived from this integrated approach. The outcomes demonstrate a noteworthy improvement in Quality of Service (QoS) metrics, including decreased latency, increased throughput, and enhanced resource utilization efficiency. The adaptive characteristics of CNN-based femtocell management play a crucial role in dynamically adapting to varying network conditions and user demands. This study not only progresses the optimization of resource allocation in 5G networks but also emphasizes the possibilities of integrating network slicing and machine learning to achieve intelligent femtocell management. The valuable

insights derived from our findings provide guidance for network operators, researchers, and industry stakeholders as they navigate the challenges presented by the continually evolving 5G landscape.

Keywords: Femtocell, Network slicing, 5G network, CNN based femtocell, Quality of Service.

## **I. Introduction:**

The evolving landscape of fifth-generation telecommunication networks (5G) aims to deliver a diverse range of end-to-end network services to meet the demands of various applications. These requirements span traditional mobile communication as well as emerging verticals like automatic driving, unmanned aerial vehicles, telemedicine, and the massive Internet of Things (mIoT). These devices need perfect channel for communication. When we have more paths in road communication, the paths are allocated based on vehicle speed. In same manner the communication to be done based on the user requirement speed or the domain specification. Example when a user is playing a game then user need high speed data transmission and if user in health domain, then user need quality of data transmission if any disturbance in data transmission, then patient information like medical reports will be affected. To have good channel allocation we are relying on network slicing and for signal providing we are using a small-scale antenna called ad femtocell [1]. Unlike the one-size-fits-all approach of fourth generation (4G) telecommunication networks, which followed a uniform pipeline with minimal consideration for service customization, 5G introduces customized network paradigms through the implementation of network slicing.

Network slicing, a promising technique in 5G, logically establishes self-contained networks from the radio access network (RAN) to the core network (CN). Each of these networks, known as network slices, is tailored to specific requirements, incorporating various network functions and resources abstracted from the underlying communication and network infrastructure. This concept aligns closely with infrastructure as a service (IaaS) in cloud computing, where computing, storage, and networking resources are shared among different tenants to provide fully functional virtual networks using software-defined networks (SDN) and network function virtualization (NFV). SDN, by separating the control and data planes, enhances data forwarding efficiency and network programmability, while NFV enables virtual network functions (VNFs) to run on general-purpose hardware, reducing deployment costs. Consequently, network

slicing, combining multiple VNFs, not only delivers a flexible, scalable, and programmable network service but also optimizes capital and operational expenditures by efficiently orchestrating and managing VNFs [2].

Leveraging the inherent capability of Convolutional Neural Networks (CNNs) to learn and discern spatial features from input data, their application in femtocell networks becomes advantageous for optimizing channel allocation across diverse factors. The resource allocation is to be done at two levels as 1. Core network allocation, 2. Radio network access. Core network allocation or sharing is not a big issue because in this source are router, switch, gateway, and firewalls [3]. These are physical available devices so there is no issue with physical sharing. In the realm of 5G networks, the strategies of reactive and proactive resource allocation serve as pivotal methodologies for effective management and optimization of network resources. These approaches are integral in achieving resource efficiency and addressing the varied needs of diverse services within the network. The following outlines how CNNs can be effectively employed for channel allocation in femtocell networks.

- Data Representation.
- Model architecture.
- Training.
- Validating and testing.
- Real time channel allocation.
- Dynamic adoption.

The paper was arranged with above points to elaborated how optimized resource allocation in 5G network.

## **II. Related work:**

As we delve into maximizing the capabilities of Device-to-Device (D2D) communication in the expansive realm of 5G networks, our research has brought to the forefront noteworthy challenges that demand thoughtful attention. Despite the anticipated advantages of heightened efficiency, optimized resource usage, and minimized latency that D2D communication holds, a set of drawbacks has emerged, creating complexities in its seamless integration. This study not only highlights these limitations but also provides invaluable insights for network

architects, service providers, and researchers who seek to refine and optimize D2D communication within the 5G era.

In figure 1 VOLTE communication model was describe for different users how they are communicating. In a few cases if the Base Station is overloaded then device to device (D2D) communication can be done. Due to this visitor location information was hidden because users are not connected to base station due to D2D model. To avoid this issue channel allocation was done with femtocell to uncovered regions. But the issue is femtocell will provide coverage for eight users which are not in coverage but if majority users need radio frequency channel with high speed, then network to be sliced based on user required speed. This method is called network slicing which is the most powerful method used in 5G communication.

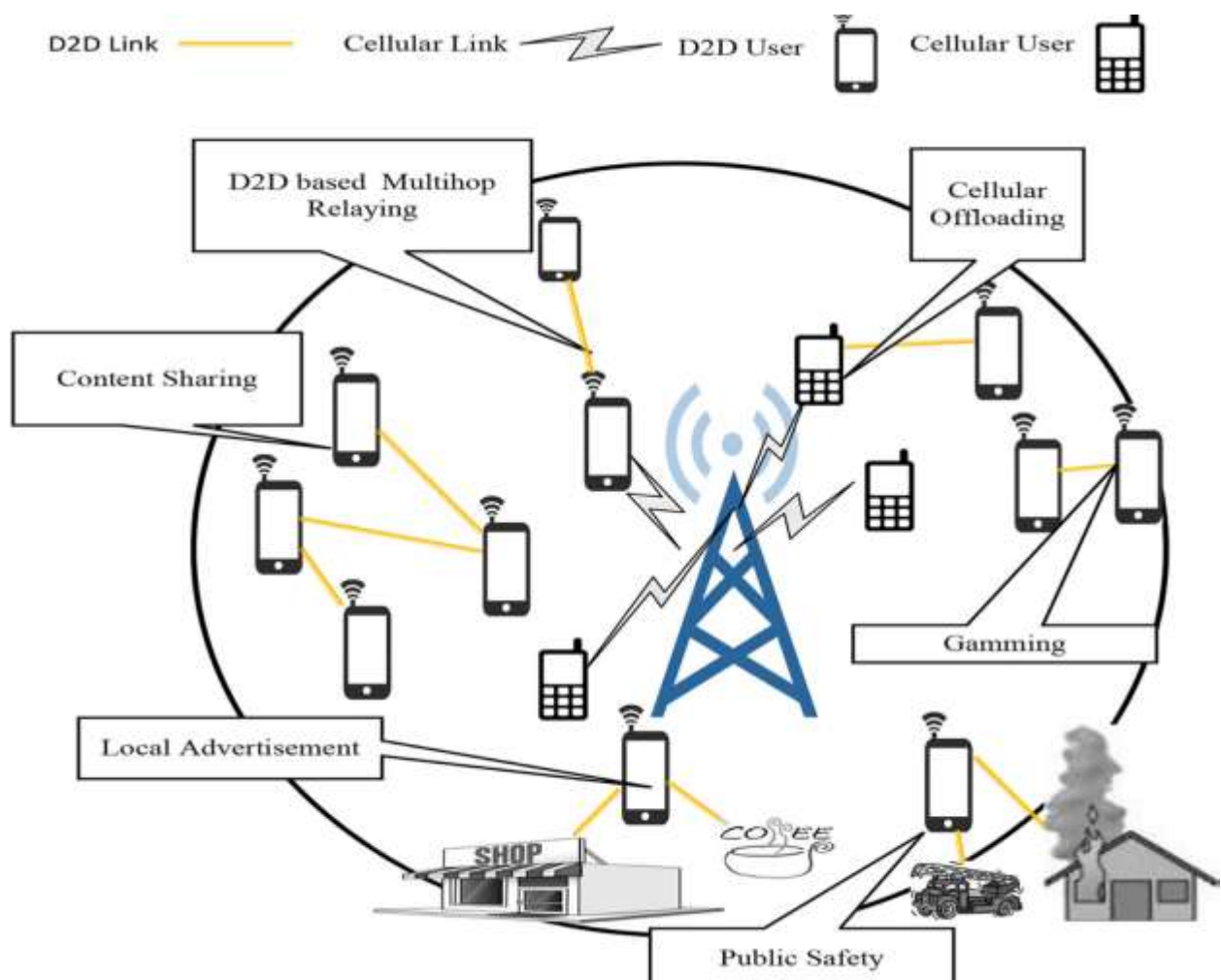


Figure 1: Voice Over Long-Term Evaluation (VOLTE) communication model.

In the realm of 5G networks, effective resource allocation stands as a pivotal consideration, gaining added intricacy and adaptability when intertwined with the concept of network slicing. Network slicing, a fundamental concept, empowers the creation of virtual networks designed for distinct use cases, each endowed with unique sets of resources, configurations, and distinctive characteristics <sup>[4]</sup>. Special features extraction will give need of CNN in resource allocation based on Significance: CNNs demonstrate exceptional proficiency in capturing spatial features within datasets, rendering them particularly apt for scenarios where the spatial distribution of resources and network conditions holds paramount importance <sup>[5-8]</sup>.

Contextual Relevance: In the framework of network slicing, comprehending the spatial arrangement of resources becomes imperative for ensuring effective allocation that aligns with the distinctive requirements of each network slice <sup>[9]</sup>.

### III. Model architecture:

Input Layer:

Define the input layer to align with the preprocessed traffic data structure, accounting for both temporal and spatial dimensions.

Convolutional Layers:

Arrange a series of convolutional layers to efficiently extract spatial features from the traffic data.

Pooling Layers:

Incorporate pooling layers to down sample the data and preserve critical information.

Fully Connected Layers:

Include fully connected layers to capture intricate relationships within the data, fostering comprehensive pattern learning.

Output Layer:

Set up the output layer with a neuron configuration matching the predefined traffic pattern classes, employing a suitable activation function (e.g., SoftMax for classification).

System parameter	Value/range
Macrocell radius	500 m
Macrocell transmission power	46 dBm
Carrier frequency	2 GHz
Bandwidth	20 MHz
White noise power density	- 174 dBm
Number of indoor femtocells	24
Number of outdoor femtocells	6
Number of macrocell active user	90
Number of femtocell active user	90

Table 1: parameters considered for experimental setup.

### III Results and discussion:

By considering the above experimental setup we tested the user equipment processor capacity and memory block usage in different time intervals. The capacity of the system is increased by having the slicing in network for communication based on user required speed, and compared to the D2D model femtocell arrangement was given best performance.

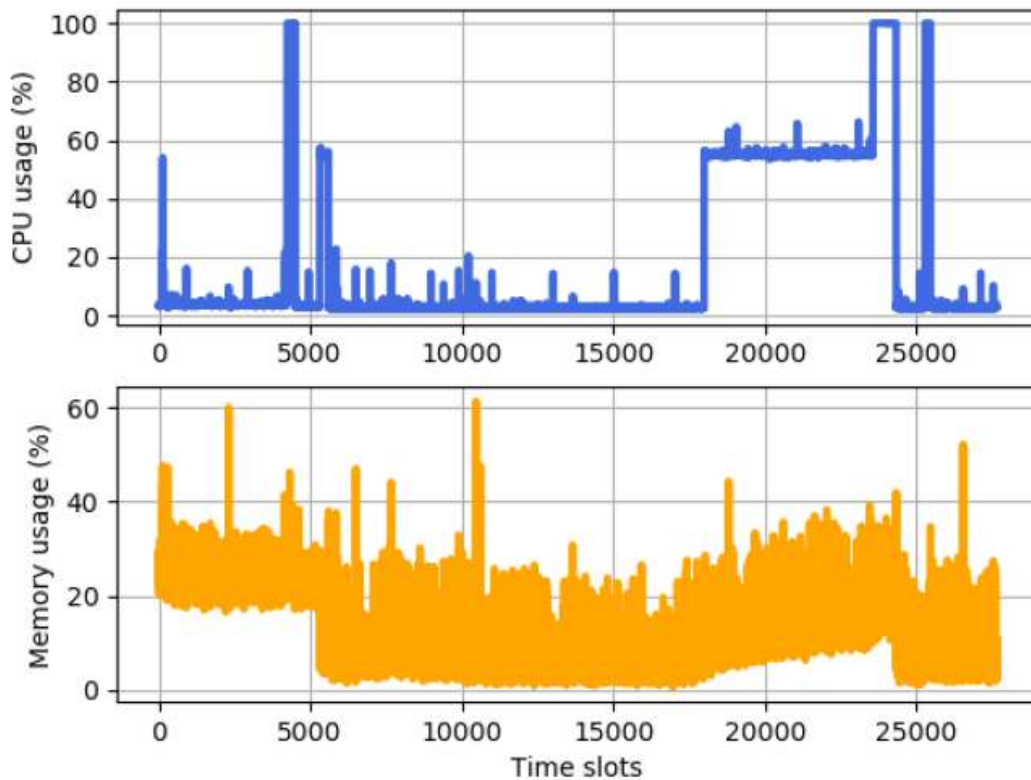


Figure 2: optimizing resource allocation at core level and radio allocation level.

The results are shown in figure 2 are obtained by considering the table 1 system parameters.

#### IV conclusion:

While femtocells prove to be applicable in both indoor and outdoor settings, the utilization of outdoor femtocells is notably lower, constituting less than half of the demand observed for indoor femtocells. Implementing outdoor femtocells with a range exceeding half of the Macro Base Station (MBS) radius yields commendable improvements in spectrum efficiency and signal quality. Leveraging the Channel Quality Indicator (CQI) coefficient enables effective management and release of MBSs, leading to enhanced network capacity. This, in turn, facilitates the provision of higher bandwidth to serve more users or support cellular tracking for individual vehicles within a smart city environment. Network slicing can be changed based on CNN algorithm will give best results by having intelligence in system for capacity analysis of channel.

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