



American International University-Bangladesh (AIUB)

# **A Study on Coverage and Capacity Optimization Techniques for 5G Networks**

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## **Abstract**

The fifth generation of mobile networks promises to meet the demand for high-speed data connectivity and seamless network experience. But it can only achieve that feat if the networks are optimized for coverage and capacity. In this paper we explore various techniques as well as research papers that tries to optimize the coverage and capacity of 5g network. We looked at techniques that can be useful in a self-organizing network. According to our study, various techniques and schemes can be used to improve the capacity and coverage, if the techniques or scheme are used in the right scenarios. This study suggests that integration of various techniques in SON is the way of achieving optimum performance.

## Declaration by author

This thesis is composed of our original work, and contains no material previously published or written by another person except where due reference has been made in the text. We have clearly stated the contribution of others to our thesis as a whole, including statistical assistance, survey design, data analysis, significant technical procedures, professional editorial advice, financial support and any other original research work used or reported in our thesis. The content of our thesis is the result of work we have carried out since the commencement of Thesis / Software project.

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## Approval

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## **Publications included in this thesis**

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## **Submitted manuscripts included in this thesis**

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## **Other publications during candidature**

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## **Research involving human or animal subjects**

No animal or human subjects were involved in this research

## Contributions by authors to the thesis

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	<i>16-31037-1</i>	<i>17-34322-1</i>	<i>17-34325-1</i>	<i>19-40293-1</i>	
Conceptualisation	25%	25%	25%	25%	100(%)
Data curation	25%	25%	25%	25%	100(%)
Formal analysis	25%	25%	25%	25%	100(%)
Investigation	25%	25%	25%	25%	100(%)
Methodology	25%	25%	25%	25%	100(%)
Validation	25%	25%	25%	25%	100(%)
Theoretical derivations	25%	25%	25%	25%	100(%)
Preparation of figures	25%	25%	25%	25%	100(%)
Writing – original draft	25%	25%	25%	25%	100(%)
Writing – review & editing	25%	25%	25%	25%	100(%)

If your task breakdown requires further clarification, do so here. Do not exceed a single page.

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## **Keywords**

5g, self-organizing network(SON), capacity and coverage optimization(CCO), heterogeneous network (HetNet)



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# Contents

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Abstract . . . . .	ii
<b>Contents</b>	<b>ix</b>
<b>List of Figures</b>	<b>x</b>
<b>List of Tables</b>	<b>xi</b>
<b>List of Abbreviations and Symbols</b>	<b>xiii</b>
<b>1 Introduction</b>	<b>1</b>
1.1 Background . . . . .	1
1.2 Problem Statement . . . . .	2
1.3 Research Objective . . . . .	3
1.4 Significance of Study . . . . .	3
<b>2 Literature review</b>	<b>5</b>
<b>3 CCO Techniques</b>	<b>7</b>
3.1 Inclusion of ML techniques . . . . .	7
3.2 MIMO for capacity optimization . . . . .	8
3.3 Network Slicing . . . . .	9
3.4 Hybrid Beamforming . . . . .	9
3.5 Edge Computing . . . . .	10
3.6 Cooperative Communications . . . . .	10
3.7 HetNet Optimization . . . . .	11
3.8 Carrier Aggregation . . . . .	11
3.9 Optimization Algorithms . . . . .	11
<b>4 Discussion</b>	<b>13</b>
<b>5 Conclusion</b>	<b>17</b>
<b>Bibliography</b>	<b>19</b>

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# List of Figures

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1.1 5G Requirements . . . . . 2

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# List of Tables

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# List of Abbreviations and Symbols

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Abbreviations	
CS	Computer Science
SON	Self-Organizing Network
SLPN	Smart Low Power Node
CCO	Coverage and Capacity Optimization
HetNet	Heterogeneous Network
<i>etc.</i>	<i>etc.</i>



# Chapter 1

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## Introduction

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In today's world, technology has become an essential part of our everyday lives. We use technology for various purposes, from staying connected with our loved ones to conducting business transactions. With the increasing use of technology, the demand for faster and more reliable internet connectivity has become stronger than ever before. That is where 5G comes in – the fifth generation of wireless technology that promises to revolutionize how we connect to the internet. The emergence of fifth-generation (5G) wireless networks have ushered in a new era of connectivity, promising faster data speeds, lower latency, and enhanced capacity [Ge et al., 2016]. Fig. 1.1 shows the requirements of 5G technologies. With 5G, users can enjoy download and upload speeds of up to 20 Gbps and 10 Gbps, respectively, which is over 20 times faster than 4G. The lower latency in 5G networks also means that there will be less delay in data transmission, resulting in better performance for real-time applications like online gaming and video conferencing. In addition, 5G networks are expected to support a larger number of devices per network area, making it possible for more devices to connect to the internet simultaneously [Kamel et al., 2016]. These advancements in 5G technology have the potential to revolutionize the way we use mobile devices and unlock new possibilities for industries like healthcare, transportation, and entertainment.

The 5G network promises to revolutionize the way we communicate and access information, enabling a range of new applications such as virtual and augmented reality, autonomous vehicles, and the Internet of Things (IoT). However, as the demand for wireless data continues to grow, it is crucial to ensure that the network is both reliable and efficient. This requires careful optimization of both coverage and capacity. Hence, it is critical to ensure that 5G networks are optimized for both coverage and capacity.

### 1.1 Background

The implementation of 5G technology has become a major focus in the telecommunications industry due to the potential benefits it offers such as higher data transfer speeds, reduced latency, and increased capacity for more connected devices. However, to achieve the full potential of 5G technology, it is

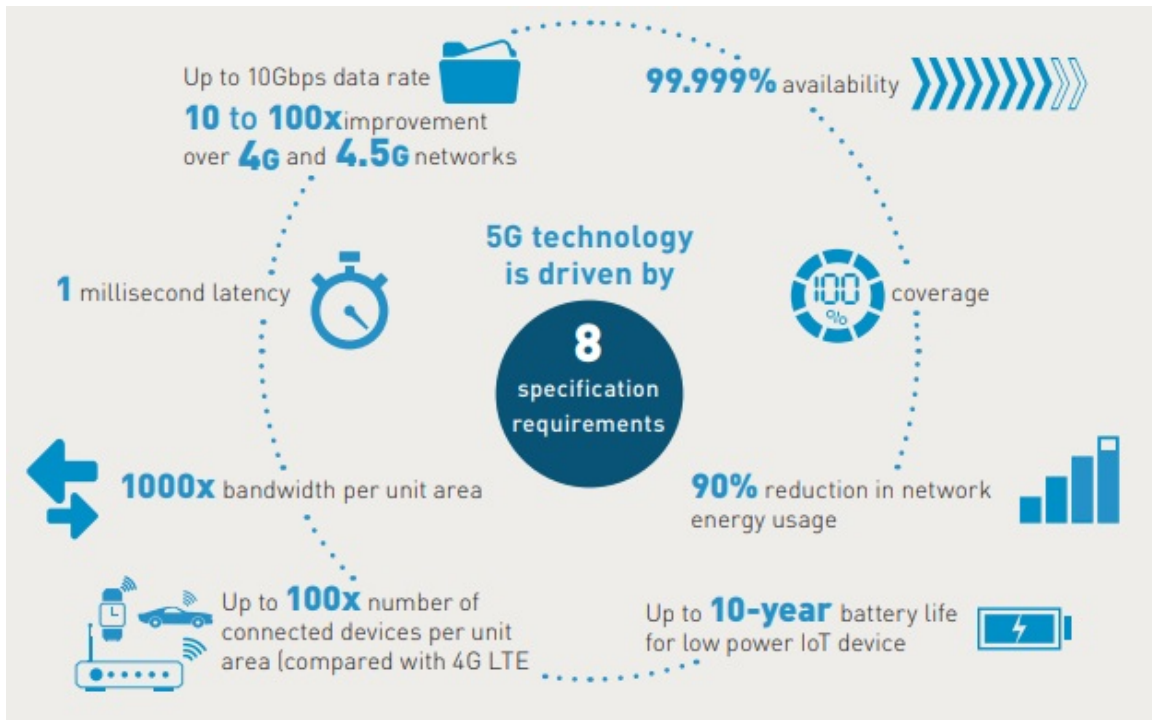


Figure 1.1: 5G Requirements

necessary to optimize the coverage and capacity of the network. This requires the development of efficient and effective 5G coverage and capacity optimization (CCO) systems. Research studies have been conducted to investigate the challenges associated with the implementation of 5G technology, and to propose solutions to address them. However, these studies demonstrate the importance of having different 5G CCO techniques to optimize the coverage and capacity of 5G networks. By addressing the challenges associated with 5G technology, systems can be developed that can ensure that users have access to fast and reliable connectivity, regardless of their location or network demands.

## 1.2 Problem Statement

Despite the promising benefits of 5G technology, the implementation of this technology presents significant challenges in terms of coverage and capacity optimization. Traditional cellular networks often struggle to provide reliable and efficient coverage in highly populated urban areas, and the implementation of 5G networks will exacerbate this challenge due to the high frequency and limited range of 5G signals. Additionally, the increasing demand for high-speed connectivity and the proliferation of connected devices present challenges for network capacity management. To address these challenges, there is a need for effective 5G coverage and capacity optimization (CCO) systems that can dynamically manage network resources to provide reliable and efficient coverage and capacity. These systems must be capable of optimizing the placement of small cells in indoor and outdoor environments, managing network resources to support different network slices, and integrating satellite communication and terrestrial networks to provide reliable connectivity in rural areas. Therefore, the problem statement is developing efficient and effective 5G CCO systems that can optimize the



coverage and capacity of 5G networks, in order to ensure reliable and efficient connectivity for all users, regardless of their location or network demands.

## 1.3 Research Objective

The research objective of this study is to develop efficient and effective 5G coverage and capacity optimization (CCO) systems that can optimize the coverage and capacity of 5G networks. Specifically, the objectives are:

1. To identify the key challenges associated with 5G coverage and capacity optimization in indoor and outdoor environments, as well as in rural areas.
2. To compare the performance of the various 5G CCO techniques.
3. To provide insights and recommendations for the practical implementation and deployment of 5G CCO systems in real-world scenarios.

## 1.4 Significance of Study

The deployment of 5G networks has the potential to revolutionize the way we communicate, enabling new applications and services that were previously not possible. However, the success of these networks relies heavily on their ability to provide reliable and efficient coverage and capacity. Therefore, understanding the different techniques used to optimize coverage and capacity is essential to ensure the full potential of 5G networks is realized. This thesis paper is significant because it aims to compare and evaluate the effectiveness of different techniques used to optimize coverage and capacity in 5G networks. By analyzing real-world data and simulations, this study will provide valuable insights into the strengths and weaknesses of different optimization strategies, enabling network planners and operators to make informed decisions about which techniques to employ. The findings of this study will have practical implications for the development of 5G networks. Specifically, it will provide insights into how to design and deploy 5G networks that can achieve optimal coverage and capacity, thus improving the overall user experience. Additionally, this research will contribute to the body of knowledge on 5G networks, providing a foundation for future research in this area. Finally, this thesis paper is significant because it addresses a critical research gap. Despite the growing interest in 5G networks, there is a lack of research comparing the effectiveness of different techniques for optimizing coverage and capacity. Therefore, this study will provide valuable insights into this important area of research, which will be of interest to both academics and industry practitioners.



## Chapter 2

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### Literature review

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5G Coverage and Capacity Optimization (CCO) is an important aspect of 5G network design and deployment. It involves optimizing the coverage and capacity of the network to ensure that it can handle the increasing demand for data traffic and support a wide range of services and applications. The following literature review provides an overview of recent research and development in 5G CCO. The successful deployment of 5G networks requires efficient coverage and capacity optimization techniques. In recent years, researchers have proposed various methods to optimize coverage and capacity in 5G networks. This literature review aims to summarize the recent research on 5G coverage and capacity optimization with relevant citations.

[Soszka et al., 2015] proposed optimization of coverage and capacity in cellular radio networks using advanced antennas. A highly complex optimization problem is faced when using 2-dimensional antenna arrays for concurrent coverage and capacity optimization. In the proposed technique, the complexity of the problem is reduced while achieving adequate performance gains. The complexity reduction is achieved by considering only a limited set of basic beams which are flexibly combined with each other in order to obtain the overall beam pattern of the actual transmission. A Nelder-Mead and a Q-Learning based approach is proposed in order to adjust the remaining parameters, i. e. vertical and horizontal angles of every basic beam. Based on system level simulations it can be concluded that 2-dimensional antenna arrays can significantly outperform linear arrays when carrying out concurrent coverage and capacity optimization.

[Roth-Mandutz et al., 2017] proposed a self-optimized approach on the network management level to minimize the capacity impact for ultra-reliable low latency communication (URLLC) services. Network Management (NM) parameters are discussed regarding their impact on URLLC and present a structure, which allows a fast selection of the appropriate parameter set. Device-to-device (D2D) communication has a high potential to meet both, the reliability and latency requirements, so the focus is on that. A scheme to optimize the reuse of radio resources for a mixed D2I and D2D traffic scheme was presented in this study. An initial study on radio resource reuse indicates that the gain in network capacity is promising, while D2D reuses the same resources as the cellular users in the network.

[Chen et al., 2018] presented capacity and coverage optimization scheme that is self-organizing,

using power adaptation to enhance the capacity and improve the coverage of a heterogeneous two-tier network with hybrid deployed small cells and macrocells. The proposed scheme can effectively improve the capacity and coverage according to the simulation results.

In conclusion, various approaches have been proposed to optimize coverage and capacity in 5G networks, including machine learning-based techniques, beamforming, power allocation, cooperative relaying, spectrum sharing, and transmission schemes. These techniques can be used to improve signal quality, coverage area, and network capacity, enabling the successful deployment of 5G networks.

## Chapter 3

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# CCO Techniques

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The efficient optimization of coverage and capacity in 5G networks is a critical challenge faced by the telecommunications industry. The growing demand for high-capacity data transfer from various devices, including smartphones, IoT devices, wearables, and autonomous vehicles, requires the deployment of infrastructure capable of handling high-capacity traffic. This infrastructure includes base stations and small cells, which need to be installed at higher densities, especially in densely populated areas, due to the high-frequency bands used by 5G networks that offer higher data transfer rates. To optimize coverage and capacity efficiently, operators must employ advanced technologies such as fiber-optic cables to support high-capacity data transmission. They must also optimize the use of available spectrum by implementing radio access technologies such as massive MIMO and beamforming, which improve signal strength and reduce interference (Qualcomm, 2020).

A significant challenge in optimizing coverage and capacity in 5G networks is signal attenuation caused by obstacles such as buildings, trees, and other structures. Due to the high frequency used by 5G networks, signal attenuation occurs as radio waves are absorbed, scattered, and reflected. Operators must overcome this issue by using advanced technologies such as beamforming, which directs narrow beams of radio waves towards the user's device to ensure a stronger signal is received (Qualcomm, 2020). The deployment of additional infrastructure, including base stations and small cells, is another challenge in optimizing coverage and capacity, mainly due to the cost involved. However, the deployment of 5G networks can provide significant benefits to network operators, such as cost savings through the consolidation of equipment and the use of virtualized network functions. Virtualization allows for the virtualization of network functions previously provided by physical equipment, enabling operators to deliver services in a more flexible and scalable way with greater control over their networks (Ericsson, 2020).

### 3.1 Inclusion of ML techniques

ML has the potential to revolutionize coverage and capacity optimization in wireless networks, including 5G. ML-based solutions can help network operators efficiently manage network resources,

optimize coverage, and enhance capacity, ultimately leading to improved network performance and customer experience. One way ML can be used to optimize coverage and capacity in 5G networks is by predicting and preventing network congestion. With the increasing number of devices and data transfers, network congestion can significantly impact network performance. Additionally, ML can be used to optimize the placement of antennas and adjust the beam direction of base stations to strengthen signal strength and reduce interference.

ML can also be used to optimize the allocation of network resources, including spectrum and bandwidth. ML algorithms can analyze network traffic patterns, predict future traffic demand, and allocate resources accordingly, ensuring efficient use of available resources and avoiding network congestion. To implement ML-based coverage and capacity optimization in 5G networks, data is essential. Network operators can use data collected from various sources, including user behavior, network traffic, and environmental conditions, to train ML models. This data can be collected using sensors, network monitoring tools, and user devices. However, data privacy and security must be a top priority when collecting and using data, and network operators must ensure that appropriate measures are in place to protect users' privacy. In [Sousa et al., 2021], The spectrum efficiency and the sum rate of a network are improved, using a simplified path loss model and reinforcement learning (RL) algorithm. Furthermore, in [Dreifuerst et al., 2021] two machine learning algorithms are proposed to optimize the coverage and capacity. And according to their finding data-driven techniques can be used to self-optimize coverage and capacity in cellular networks in an effective manner.

Hence, ML-based solutions offer significant potential to optimize coverage and capacity in 5G networks. By predicting network congestion, optimizing the deployment of base stations and small cells, and allocating network resources efficiently, network operators can improve network performance, enhance customer experience, and reduce costs. However, to implement ML-based solutions, network operators must collect and analyze vast amounts of data, and ensure that appropriate data privacy and security measures are in place.

## 3.2 MIMO for capacity optimization

Multiple-Input Multiple-Output (MIMO) technology is a crucial tool for capacity and coverage optimization in wireless networks, including 5G. MIMO enables wireless systems to transmit multiple data streams simultaneously through multiple antennas at both the transmitter and receiver ends. This technology increases the number of spatial dimensions available for transmission, which can significantly improve the network's capacity and coverage.

One of the key benefits of MIMO is its ability to mitigate signal fading caused by multipath propagation. Multipath propagation is a phenomenon where signals from a single transmitter reach the receiver via multiple paths, resulting in signal attenuation and distortion. MIMO technology can effectively mitigate multipath fading by exploiting the diversity of signals received by multiple antennas. The receiver can then combine these signals to reconstruct the original signal and improve its quality. In addition to mitigating multipath fading, MIMO can also improve network capacity

by increasing the number of data streams transmitted simultaneously. MIMO systems can transmit multiple data streams through different antennas at the same time, which significantly increases the amount of data that can be transmitted in a given time period. This increases the network's capacity, which is essential for meeting the growing demand for high-speed data transfer in 5G networks.

MIMO technology can also enhance coverage in 5G networks by improving signal strength and reducing interference. By utilizing multiple antennas, MIMO systems can focus signals on specific areas of the network, improving the strength of the signal in those areas. This approach can increase the coverage area of the network while maintaining a high level of signal quality. Additionally, MIMO can reduce interference by canceling out signals from unwanted sources, ensuring that only the desired signals reach the receiver. To fully realize the benefits of MIMO for coverage and capacity optimization in 5G networks, network operators must implement appropriate antenna configurations and deployment strategies. The optimal configuration and deployment of antennas depend on various factors, including the network topology, traffic patterns, and environmental conditions. Network operators must also consider factors such as power consumption, cost, and regulatory constraints when designing MIMO-based systems.

Therefore, MIMO technology is a crucial tool for capacity and coverage optimization in 5G networks. By mitigating signal fading, increasing network capacity, and improving signal strength and reducing interference, MIMO technology can significantly enhance network performance and improve customer experience. However, to fully realize the benefits of MIMO, network operators must carefully design and deploy MIMO-based systems, considering various factors such as network topology, traffic patterns, and environmental conditions. [Chataut and Akl, 2020]

### 3.3 Network Slicing

Network slicing is a technique used in telecommunications networks, particularly in the context of 5G networks, to create separate logical networks that can be tailored to specific use cases or services. With network slicing, a single physical network infrastructure can be partitioned into multiple virtual networks, each with its own set of characteristics such as bandwidth, latency, security, and reliability. The concept of network slicing allows for a more flexible and efficient use of network resources, as it enables network operators to allocate network resources dynamically based on the needs of different applications or services. [Zhang, 2019]

### 3.4 Hybrid Beamforming

Hybrid beamforming is a technique used in wireless communication systems, particularly in millimetre-wave (mm Wave) frequencies, to achieve the high data rates required for 5G and beyond. It is a combination of analog and digital beamforming techniques that enable the transmission and reception of high-speed data over long distances. In hybrid beamforming, the signal is first processed digitally to reduce the complexity of the analog components. The signal is then passed through an array of antenna

elements, which are grouped into subarrays. The subarrays are then connected to a network of analog phase shifters that steer the beam in the desired direction. The digital component is responsible for processing the signals that are transmitted or received by the antenna array. This processing includes filtering, amplification, and modulation. The analog component is responsible for adjusting the phase of the signals, which determines the direction of the beam. Hybrid beamforming offers several advantages over traditional beamforming techniques, such as lower power consumption, reduced complexity, and improved efficiency.

### 3.5 Edge Computing

Edge computing is a technique for optimizing 5G CCO by offloading processing and storage to edge devices. This approach can help to improve network capacity and reduce latency by reducing the amount of data that needs to be transmitted to the core network. It refers to a distributed computing architecture that brings computation and data storage closer to the devices and sensors that generate data, rather than relying on centralized data centers or the cloud. This approach enables faster response times, reduces network latency, and can be more cost-effective than traditional cloud computing models. With edge computing, devices and sensors can perform local processing and storage of data, reducing the amount of data that needs to be sent over the network to the cloud or a central data center. This approach can also provide greater privacy and security, as sensitive data can be processed and stored locally rather than being sent over the internet. Edge computing has many applications, including in the fields of healthcare, manufacturing, transportation, and smart cities. For example, in healthcare, edge computing can be used to process data from medical devices, such as wearable sensors or implanted devices, in real-time to monitor patients and deliver personalized treatments. [Hassan et al., 2019]

### 3.6 Cooperative Communications

Cooperative communications is a technique for optimizing 5G CCO by enabling devices to cooperate and share resources. This approach can help to improve network capacity and coverage by enabling devices to work together to transmit and receive data. In cooperative communication, multiple devices form a cooperative network, and each device may act as a relay to help transmit data from a source device to a destination device. This can increase the overall coverage and range of wireless communication, as well as improve signal quality and reduce interference. Cooperative communication can be used in a variety of applications, such as mobile ad-hoc networks, sensor networks, and wireless mesh networks. [Guo et al., 2022]



### 3.7 HetNet Optimization

Het Net optimization is a technique for optimizing 5G CCO by deploying heterogeneous networks (HetNets) that combine different types of access technologies. This approach can help to improve network coverage and capacity by leveraging the strengths of different access technologies. The goal of Het Net optimization is to improve network performance by balancing traffic loads and ensuring that users have a high-quality wireless experience, regardless of their location within the network. This can involve adjusting the transmission power and coverage of different cells, optimizing handover procedures, and managing interference between cells.

### 3.8 Carrier Aggregation

In carrier aggregation, multiple carriers, each with a different frequency band, are aggregated or combined together to create a larger bandwidth. This larger bandwidth allows for higher data transfer rates, which can be used to support high-bandwidth applications such as virtual reality, augmented reality, and 4K/8K video streaming.

One of the advantages of carrier aggregation is that it allows operators to utilize their existing spectrum resources more efficiently. By combining multiple carriers, operators can create a larger bandwidth, which allows them to serve more users simultaneously and support high-bandwidth applications.

Overall, carrier aggregation is an essential technology in 5G networks, which enables higher data rates and better network capacity by aggregating multiple frequency bands. [Nidhi et al., 2020]

### 3.9 Optimization Algorithms

Providing high data rates, low latency along with enhanced coverage and capacity is incredibly challenging for cellular networks. This is where the optimization algorithms come in. They reduce the network complexity which in turn enhance the key parameters of 5G. [Sudhamani et al., 2023]



## Chapter 4

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### Discussion

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[Zhao et al., 2018] proposed a novel CCO scheme to maximize the utility function of the integrated coverage and capacity. Initially, they analyze the throughput proportional fairness (PF) algorithm. After that their novel Coverage and Capacity Proportional Fairness (CCPF) allocation algorithm is proposed along with proof of the algorithms' convergence. Then they simulate the proposed CCO scheme based on telecom operators' real network data from two aspects: 1) regular 6+1 BSs (Base station) sim-scene which only contains one central BS and 6 neighbor BSs. 2) real-world network area, i.e. Beijing e-Town. And then compare it with some typical resource allocation algorithms: round robin, proportional fairness and max C/I. The performance metrics used were: received signal power for coverage and user's throughput and cell's Resource Occupation (RO) ratio for capacity. In the first aspect, the authors find that their proposed CCO scheme increased more than 30%, edge users. In terms of RO comparison, the neighbor BSs have RO that is higher than 80%, which is considered "heavy" and the central source BS has a "light" RO before the CCO. But after the CCO all the BSs have a more reasonable and similar RO, which indicates a good balance between coverage and capacity. However, the CCO process with a traditional resource allocation algorithm causes the edge users to have low scheduling priority. To solve this issue CCPF is adopted after CCO. The edge user's throughput with CCPF reaches 7.2 Mbps which, when compared to the edge user's throughput on Round Robin(RR), and Proportional Fairness(PF), is 2.11 and 2.57 times increment respectively. Max C/I has zero edge user throughput. The mean throughput of CCPF can reach up to 10.1 Mbps which is 1.34 and 1.61 times compared to PF and RR respectively. In peak throughput comparison, max C/I is considerably better than PF, RR, and CCPF. CCPF has the lowest peak throughput performance among the algorithms, 90.3% and 95.2% of performance on RR and PF respectively. In the second aspect, Beijing e-Town is selected for simulation scenarios for investigation. In this scenario, after the CCO process is applied, the low-traffic loaded BSs balance the high-traffic loaded BSs' load by the transmission power accommodation, which causes the RO of the whole area to be more reasonable. In terms of edge users and mean throughput performance, CCPF is better than PF, RR, and Max C/I. Max C/I allows the central users to get high mean throughput but CCPF can be 1.96 and 1.54 times better than the RR and PF respectively in mean throughput performance. CCPF scores 0.81 on Jain's

Fairness Index compared to 0.53, 0.72, and 0.90 scores of Max C/I PF and RR respectively.

[Su et al., 2018] proposed a mathematical combined optimization function to balance the conflicting key performance indicators in a densely deployed small cell network. For generating new antenna transmit power to optimize the performance under this model, they propose the Tabu Search Algorithm. In this paper, a scenario case based on a hexagonal 19-site network deployment was considered. Which had one original site deployed in the middle and six others wrapped around it symmetrically. Other twelve sites were attached to the six ones' sides. In the simulation, the Tabu Search(TS) algorithm is compared to the Simulated Annealing (SA) algorithm. The simulation area considered was a hotspot area with MBSs deployed outdoors and SBSs indoors. The indoor clusters are uniformly random within 2 sectors of the macro geographical area. And the SBSs are uniformly random dropping within the cluster areas. In the simulation, for the same neighbor range of  $r = 4$ , compared to SA the proposed TS algorithm has significantly faster convergence in iteration, all while reaching almost the same near global optimum as the contrast approach with about a 32% gain. As the TS algorithm converges quickly to a constant during iteration, it can gain a 21% capacity improvement compared to the SA algorithms 15%. Even in the Cumulative Distribution Function (CDF) of the UE throughput in the cells of the final solution, the proposed TS algorithm improves both the average throughput and edge throughput.

[Sun et al., 2015] present a novel scheme of adaptive self-organizing network(SON) by combining the spectrum sensing function from Cognitive Radio(CR) and the radio resource layering function from the inter-cell interference coordination(ICIC). The author also develop a Hungarian algorithm based self-organization strategy to improve the Smart Low-Power Node(SLPN) optimization. Initially a 5G SLPN concept is developed. The deployment of SLPN in this paper, have four principles: mitigation of cross-tier interference, maximization of the overall capacity of Heterogeneous Networks(HetNets), flexibility and adaptability, and self-configuration and self-optimization strategies with low computational complexity. The paper has for parts for SLPN for adaptive SON, these are : construction of interference matrix, multilayer frequency reuse and power assignment, multi-sector frequency reuse and SLPN optimization algorithm. The simulation setup consists of nineteen cell sites in a hexagonal grid with three sectors per site. In the simulation proposed adaptive SON is compared with the performance of homogeneous network and HetNet without SON. The simulation shows the HetNet with plain LPNs improves the coverage performance around coverage holes to some extent but it fails to improve the overall network performance in a significant way. On the other hand SLPN achieves a SINR layout that overcomes the challenge of cross-tier and co-tier interference because of the dynamic algorithm proposed. Alleviating the interference in its own cell is not the only thing the proposed algorithm achieves, it can also improves the total network performance by reducing interference to other power nodes.

[Franco et al., 2016] propose a cooperative and cognitive SON framework (CCSON) for 5G mobile radio access networks. Providing RAN elements with cognitive capabilities is the main idea here.

The CCSON framework consists of following design principles : Distributed solutions, Modu-

larity, Self-awareness, Learning capabilities, Cross-layer design and Status and Knowledge sharing capabilities.

The functional architecture of this framework consists of three layers:

1) Perception layer: In this layer functions that support the optimization and learning tasks are included.

2) Decision making and learning layer: Implementation of the self-organizing functions through independent modules called self-organizing modules(SOM) is the purpose of this layer.

3) Operation Layer: This layer adjusts and re-configures the network parameter in RAN elements based on the decision taken by the SOMs at decision making and learning layer. Individual Operation Element(OE) controls this information.

In the simulation it can be observed that the scheme proposed by the authors can achieve the optimum performance achieved by the Adaptive Bias Configuration (ABC) technique but with better results with regards to convergence time.



## Chapter 5

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### Conclusion

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5G is the fifth generation of mobile networks, promising to deliver faster data speeds, lower latency, and more reliable connections compared to its predecessors. With 5G, users can enjoy high-quality video streaming, faster downloads, and improved gaming experiences, among other things. It also has the potential to enable new technologies, such as self-driving cars and smart cities. 5G operates on higher frequencies than previous generations, which allows for more data to be transmitted at once. However, the higher frequency also means that 5G signals have a shorter range and can be more easily blocked by obstacles such as buildings and trees. Despite these challenges, 5G is expected to have a significant impact on the way we use and interact with technology.

5G Self-Organizing Networks (SON) are designed to automate and optimize the deployment, operation, and maintenance of 5G networks. SON is a critical aspect of 5G network architecture that enables network operators to efficiently manage the complexity of the 5G network. With SON, the network can automatically configure, optimize, and heal itself in response to changing network conditions and traffic demands. This eliminates the need for manual configuration and troubleshooting, reduces network downtime, and improves network performance. SON enables network operators to quickly adapt to new services and applications, and efficiently manage the growing number of connected devices and users. The use of AI and machine learning technologies in SON also improves the network's ability to predict and prevent issues before they occur, further enhancing the network's reliability and performance.

Coverage and capacity optimization are two critical aspects of self-organizing networks (SON) that enable efficient management of 5G networks. Coverage optimization ensures that the network provides a consistent and reliable signal to all users, regardless of their location. SON achieves this by automatically adjusting the transmission power, antenna orientation, and other network parameters in response to changing network conditions. Capacity optimization, on the other hand, ensures that the network can handle the increasing volume of data traffic generated by 5G devices. SON addresses this challenge by optimizing the allocation of network resources, such as radio spectrum and bandwidth, to maximize network capacity. By automating these optimization processes, SON reduces the need for manual intervention, eliminates human error, and improves the network's overall efficiency and

performance. With coverage and capacity optimization, SON enables 5G networks to provide the high-speed, reliable, and seamless connectivity that users demand.

Optimizing coverage and capacity in 5G networks is crucial to meet the increasing demand for high-capacity data transfer. To overcome the challenges associated with 5G technology, operators must invest in advanced technologies such as fiber-optic cables, massive MIMO, and beamforming to ensure efficient use of available spectrum and strengthen signal strength. Additionally, operators must prioritize the deployment of additional infrastructure, such as base stations and small cells, to support the increased demand for data transfer. Finally, the security of 5G networks must be a top priority for operators, who must implement security measures to prevent cyber-attacks that could compromise the privacy and security of users.



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