

MMSE Detection and Interference Mitigation in MIMO Antenna Systems in Wireless Communication

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Abstract: The aim of this research is to enhance signal processing methods for large-scale multiple-input, multiple-output (MIMO) antenna systems, targeting quick and low-interference transmission in seventh-generation wireless communication networks. In our modern world, it also encounters challenges like increased computational complexity and inter-user interference. This paper presents innovative approaches such as linear precoding, minimum mean square error (MMSE) detection, and interference alignment to mitigate interference. Additionally, we explore the application of machine learning for adaptive beamforming and channel estimation, especially in fluctuating environments. The effectiveness of these algorithms is evaluated through simulations and theoretical comparisons, focusing on their spectral efficiency, energy consumption, and bit error rate (BER). Results reveal that implementing these advanced signal processing techniques can significantly reduce interference and boost system performance, particularly in highly populated urban settings anticipated in the evolution of 5G and beyond. This research holds promise for developing more efficient massive MIMO systems, paving the way for future networks to satisfy the increasing demand for improved communication capabilities.

Keywords: MIMO antenna Systems, Signal Processing, BER, Bit error rate, Signal systems, Wireless Communication, Interference Mitigation.

I. INTRODUCTION

Mobile networks are evolving rapidly, enabling new forms of communication. 5G beamforming is one of the key technologies driving this transformation. This upper-layer signal processing technology is gaining traction in connectivity between base stations and user devices, bringing efficiency, speed, and capacity to unprecedented levels. By precisely targeting radio power to targeted users, this improvement is expected to optimize call signal quality and minimize interference to other users—a key pillar of the 5G offering.

In the deployment of 5G networks, beamforming plays a key role in addressing the challenges posed by the use of high-frequency millimeter wave (mmWave) bands and large-scale multiple-input (MIMO) systems. Therefore, this paper provides an overview of beamforming algorithms, with a particular focus on the Uniform Linear Array (ULA) and its impact on Channel State Information (CSI) estimation. In addition, techniques for achieving maximum signal-to-noise ratio (SINR) are examined, as well as the key hardware elements to bring these concepts from the design phase into practice, enabling improvements in the quality of wireless communication (Figure 1).[1]. The goal is to enable the connection of significantly more devices, in many cases 10 to 100 times more than previous generations of this

technology. A key element of Massive MIMO is beamforming, which directs the radio signal in specific directions, minimizing interference levels. Massive MIMO systems use CSI to optimize the transmission and reception processes.

5G and Massive MIMO Fundamentals: 5G Requirements and Use Cases

5G technology is the next generation of wireless communication systems and promises high levels of speed and efficiency to satisfy the growing number of users. This new generation of mobile networks is expected to offer extremely high data rates, very low latency, and much greater network capacity. All of these improvements are essential to enable numerous applications across all possible sectors. The first use case for 5G is enhanced mobile broadband, where peak data speeds of 20 Gbps and reliability of at least 100 Mbps are expected. These improvements lead to increased performance to support high-quality video, virtual and augmented reality, and large file downloads. The fifth reason is the vast scale of IoT that 5G can support. The goal is to enable the connection of significantly more devices, in many cases 10 to 100 times more than previous generations of this technology. It offers the advanced capabilities required for smart cities, industrial IoT, and connected vehicles, as well as for use cases in critical

infrastructure and public safety. With its ultra-low latency of 0.1 milliseconds or less and an unprecedented error rate of 10^{-9} , 5G can be used for real-time control and monitoring of critical infrastructure, telesurgery, and autonomous vehicles.[2]

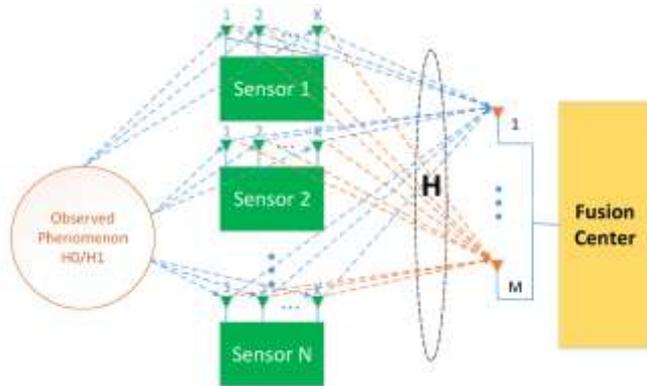


Figure 1: Beamforming in 5G: Optimizing Signal Processing for Massive MIMO

Principles of Massive MIMO

MIMO technologies of 5G networks, and performance targets depend heavily on it. Massive MIMO systems use multiple fundamentals of Massive MIMO are based on three general concepts: spatial diversity, spatial multiplexing, and beamforming.

Spatial diversity addresses fading and also facilitates wireless communication. Spatial multiplexing allows multiple data streams to be sent simultaneously, improving overall data throughput. A key element of Massive MIMO is beamforming, which directs the radio signal in specific directions, minimizing interference levels. Massive MIMO systems use CSI to optimize the transmission and reception processes. The CSI provides information about how a signal travels, as well as additional information such as dispersion, fading, or power loss. This knowledge enables the base station to adapt transmission strategies accordingly, thus improving system performance.[3]

Basics of Beamforming

It is worth explaining the concept of beamforming, one of the most important technologies in 5G networks, which is particularly applicable in combination with MMIMO. This is achieved by directing a wireless signal in a specific direction without sending it in random directions. This significantly improves signal quality, coverage of a specific area, and the overall performance of the network compared to "frequency

combining." In 5G systems, beamforming plays a crucial role. In these high-frequency bands, a wide range of spectrum is still available to innovators, especially for wireless communications. However, as the center frequency increases, the dominant path loss increases, making signals less sensitive to obstacles. Beamforming avoids these limitations by directing the signal power to the intended receiver.

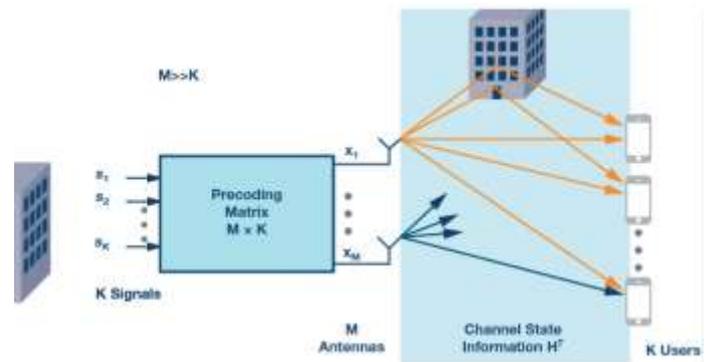


Figure 2: The Signal Processing Behind the 5 G Buzzwords

In 5G technology, more sophisticated beamforming techniques, such as 3D beamforming, not only control the horizontal and vertical direction of the beam. This feature is particularly useful in urban environments, improving coverage and capacity for users working in multi-story buildings. The use of massive MIMO in conjunction with beamforming in 5G networks also improves SINR and thus coverage and data rates, while reducing power consumption. Together, these technologies create a more reliable and resilient wireless network for any 5G application.[4]

II. ANTENNA ARRAY TECHNOLOGIES

The use of antenna array technologies poses significant signal processing challenges in massive MIMO systems of fifth-generation networks. These superior configurations enable beamforming strategies that significantly improve the performance and efficiency of wireless links.[5-6]

Uniform Linear Arrays (ULAs)

ULAs are a central antenna array in 5G systems and are called Uniform Linear Arrays (ULAs). They consist of multiple antenna elements and have a linear geometry with equal spacing between them. This arrangement provides an excellent method for beam steering and focusing signal transmission and reception. They are particularly used in beamforming, as they allow the beam to be directed in specific directions to increase its

amplitude and eliminate interference. In a 5G network, ULAs with a relatively high number of antenna elements are used to improve the MMI structure. For example, 5G NR (New Radio) networks were described in the 3GPP Release 15 specifications with 32 antennas; more are planned for version 16 and later, up to 64 or more. This growth in network size has led to the term Massive MIMO, which uses spatial diversity, spatial multiplexing, and beamforming to improve network capacity. As with standard beamforming, the beamforming performance of ULAs depends on the number of elements, the spacing between elements, and the phase and amplitude control of each element. By modifying these parameters, base stations can form highly directional beams that improve SINR and energy efficiency.

Uniform Planar Networks

Uniform Planar Arrays (UPAs) are the two-dimensional extension of ULAs, in which the antenna elements are arranged in a planar fashion. This configuration offers several advantages over linear networks, particularly with regard to 5G mmWave communications. Using UPAs, the system supports 3D beamforming, in which both horizontal and vertical beam directions can be controlled horizontally. This feature is particularly useful in urban infrastructure, especially in high-rise buildings, as it improves overall coverage and user performance across multiple floors. Another advantage of UPAs is that their planar structures can be easily integrated into compact devices, a limitation primarily encountered in 5G devices. They are used in base stations and user equipment for practical implementations of massive MIMO handoff and beamforming in 5G communication systems. UPAs can control both the phase and amplitude of each array element, thus shaping the radiation pattern to achieve maximum signal gain with minimal interference. This is also important for maintaining excellent link quality in the propagation environment characterized by millimeter-wave frequencies.

Distributed Antenna Systems

Distributed Antenna Systems (DAS) offer a unique solution for improving link density and coverage capacity in 5G networks, especially indoors or with complex geometries. Unlike centralized antenna systems, which are typically installed with multiple nodes within or across a facility or site, DAS nodes are installed in different parts of a building or area and connected to a control unit. Due to the low penetration capacity of high-frequency millimeter-wave signals, the application of DAS is crucial in 5G. This can be achieved through the correct

implementation of antenna nodes, which will be discussed in more detail later. DAS is used to maintain constant signal strength in areas where signals from external base stations cannot penetrate (e.g., very deep inside buildings or below the ground).

Deployments for 5G involve the use of small cells

In many cases, DAS deployments for 5G involve the use of small cells, or "gNodeBs," as signal sources. These compact units act as microcellular stations within the operator's network, connecting to its core network via the internet or fiber optics. This distributed architecture also aims to increase network capacity and signal quality by handling the high data rates of 5G services. Furthermore, DAS is easily scalable to support multiple operators and technologies simultaneously. This multi-operator capability is even more useful in mega-facilities such as stadiums or airports, where a single service provider among a large number of operators is critical. With the gradual evolution of 5G technology, DAS designs are evolving to incorporate new frequency bands such as C-band and additional enhancements such as 2x2 or 4x4 MIMO. Therefore, its continuous development further reinforces the importance of 5G for the provision of reliable and widespread 5G signals, especially indoors.[7] Channel Estimation in Massive MIMO

This paper clearly demonstrates the importance of channel estimation for optimizing signal processing in massive MIMO systems of the 5G network. High reliability of channel state information (CSI) is required for the use of methods such as beamforming, which can significantly improve the efficiency of 5G communication systems. Channel estimation in massive MIMO presents several challenges due to the large number of antennas used and the need for resource optimization.

Uplink Pilot Transmission

In the massive MIMO scenario, the uplink pilot transmission is most commonly used for channel estimation. This approach exploits the inverse symmetry of the airborne transmission channels between the antenna and the user terminal. Using uplink pilots, the base station can capture the effective properties of spatial channels, thus facilitating the estimation process.

The use of uplink drivers offers several advantages:

1. Reduced complexity: The difficulty of channel estimation is actually due to the number of user terminals rather than the number of antennas in the network. This facilitates the

scalability of massive MIMO systems.

2. Energy efficiency: Inaccurate and intensive channel estimation and signal processing are implemented and performed at the base station, while the use of user resources on the portable device is minimized, thereby improving the energy efficiency of the portable user terminal.
3. Minimal pilot overhead: Only one pilot signal needs to be transmitted from each user terminal, while all antenna elements in the base station can receive it. This results in significantly lower pilot overhead than if the estimations were performed via the downlink.

Downlink Channel Estimation

Although an uplink-based channel estimation scheme is feasible for important, especially for FDD systems where the uplink and downlink channels are not reciprocal. A key feature of that use different frequency bands. This creates the need for channel-independent estimation. To address the challenges of downstream channel estimation in massive MIMO, several advanced techniques have been proposed:

1. Compressive Sensing (CS): CS-based methods have taken on the task of reducing driver overhead due to the sparse nature of the channel. These techniques enable accurate CSI estimation with relatively short pilots, making them suitable for massive MIMO structures.
2. Block Iterative Support Detection (ISD): This algorithm leverages the block-wise nature of MIMO channel impulse responses (CIRs) to improve estimation and reduce pilot costs.
3. Adaptive algorithms: Techniques such as the Sparsity Adaptive Matching Pursuit (SAMP) algorithm do not require a priori information about channel dispersion levels and can therefore be more effective in practical applications.

III. BEAMFORMING ALGORITHMS

Beamforming algorithms are important for determining signal processing associated with future 5G networks with massive MIMO systems. These algorithms are used to improve signal quality, minimize interference, and increase overall system performance. To date, the three most commonly used beamforming techniques have been Maximum Ratio Transmission (MRT), Zero-Forced (ZF) beamforming, and Minimum Mean Square Error (MMSE) beamforming.

Maximum Ratio Transmission

Maximum Ratio Transmission (MRT) is a method that uses location information at the transmitter to optimize the signal-to-noise ratio (SNR) at the receiver. With MRT, the base station adjusts the transmitted signal based on channel gains to deliver more energy to the target user. This occurs especially when the interference level is low. MRT works by using channel gain conjugates as factors in the modulation symbols that accompany the signal during transmission. It ensures synchronization of the signals of all antennas at the receiving node, thus enabling constructive interference of electromagnetic energy. The result is reception with a stronger signal-to-power ratio, which, while very advantageous at most frequencies, is particularly suitable for addressing propagation problems associated with very high frequencies such as millimeter wave frequencies. Contrary to what was discussed above, while the MRT algorithm is capable of achieving a high signal-to-noise ratio (SNR), it does not provide a solution to multi-user interference when deployed in heavily congested networks (Table 2).

Zero-Force Beamforming

Another technique, Zero-Force Beamforming (ZF), aims to completely eliminate interference from multiple users. Unlike MRT, which aims for maximum signal strength, ZF takes other users of the system into account. This makes it particularly useful in situations where such interference is a major problem. The ZF algorithm works by generating orthogonal beams to different users, so that interfering elements in the channel matrix space have zero values. This approach allows each user to receive only the desired signal, without interference from other users' signals. However, this interference cancellation presents an additional problem, especially on low-recovery channels where the noise level increases. Since ZF is very effective in systems with a large number of antennas at the base station in massive MIMO systems, training more antennas allows the algorithm to generate more precise beams, which suppresses interference and thus improves system performance. Minimum mean square error (MSE) beamforming is an advanced technique designed to avoid interference between multiple users. Unlike MSE, which is solely concerned with maximizing signal strength, IF takes into account the impact of other users on the system. This makes it particularly suitable for scenarios where interference is a major concern. The IF algorithm works by generating orthogonal beams for different users, effectively forcing zeros into the interference terms of the channel matrix. This approach ensures that each user receives only the desired signal and that

interference from other users' transmissions is minimal.

However, this interference cancellation leads to a potential increase in noise, especially on low-gain channels. IF is particularly effective the algorithm can generate more precise beams, resulting in better interference cancellation and improved overall system performance.[10] Minimum Mean Square Error Beamforming

This method is a compromise between the MRT and IF approaches and is called minimum mean square error (MMSE) beamforming. Its goal is to reduce the mean square error between the transmitted signal and the estimated received signal with respect to noise and interference. Previous studies have shown that MMSE beamforming achieves higher reliability compared to other algorithms such as IF Hybrid, Kalman, Fully Digital MSE, and purely analog spectral efficiency-based precoding schemes. This makes it a good candidate for the 5G network, as the key factor in developing a 5G network is achieving maximum spectral efficiency to meet the increasing demand for data rates.

Another advantage of MMSE beamforming is its multipath configuration. By estimating signal and noise power, MMSE should perform well under all conditions, from noise-dominated to interference-dominated. Therefore, analog and digital beamforming can be combined in the practical implementation of 5G systems. This approach is most effective for millimeter-wave communication systems. This hybrid beamforming enables coarse beamforming at the analog level followed by finer digital beamforming within the analog beam.

Chen et al. suggested that this approach can fully utilize the advantages of various beamforming algorithms to achieve high performance. With the progressive deployment of the 5G network, efforts are focused on developing these beamforming algorithms to improve SINR, minimize power consumption, and expand system capacity. The selection of the beamforming algorithm is based on the deployment scenario, user traffic intensity, and the physical characteristics of base stations and user terminals.[11] Multi-User MIMO Considerations

One of the key challenges of 5G networks is the MU-MIMO solution, which reroutes the transmission and reception of multiple inputs and outputs simultaneously, thus significantly improving spectral efficiency. However, the introduction of MU-MIMO in 5G systems brings its own challenges, which must be discussed in more detail below.

User Grouping Strategies

This is because proper user grouping plays a crucial role in the overall performance of MU-MIMO in 5G design. The goal is to group users with good channel conditions to optimize system capacity for fairness. One solution is the OUG (Overlapped User Grouping) technique, which allows users with good channel conditions to belong to multiple beamforming groups. This method takes advantage of the intensive use of propagation in massive MIMO structures of the 5G network.

The second method for grouping users is to use the K-means clustering method to categorize users. However, in the case of MU-MIMO, there are limitations because the cluster size cannot be controlled, an important factor when the number of users exceeds the number of base station antenna elements.

In response to this challenge, clustering algorithms have been proposed in the literature that address not only user selection but also ordering to improve system capacity.² Silent Interference Management (MU-MIMO) technology is critical in 5G networks because it enables simultaneous transmission to multiple users and significantly improves spectral efficiency. However, implementing MU-MIMO in 5G systems presents several challenges that require careful consideration.

Managing Interference Between Users

Managing interference between users is a significant challenge in MU-MIMO systems, especially in 5G networks, where high data rates and network capacity are critical. Using beamforming algorithms significantly reduces this interference. A two-stage precoding scheme, originally standardized by 3GPP for 5G New Radio (NR), uses linear and nonlinear precoding to optimize excess capacity and reduce system complexity. One of the most effective solutions is the zero-forcing (ZF) precoding technique, which aims to minimize interference between users. Combined with Tomlinson-Harashima (THP) precoding, this scheme achieves a good balance between increased capacity and required computational load. Numerical results also showed that this combined online and offline learning strategy is particularly useful in 5G systems, where a base station must handle many users who may experience different channel conditions.

Fairness in Resource Allocation

Because the MU-MIMO system is designed to seamlessly support various applications in 5G networks, it is important to

develop a method that enables fair signal scheduling. The challenge is to maximize the system's effective throughput while maintaining reasonable fairness among users with the worst channel quality. This is especially important for stationary IoT devices with low power requirements, as they may be continually blocked by other users with better channel conditions.

IV. HARDWARE IMPLEMENTATION ASPECTS

Beamforming in the context of new 5G networks raises important questions for hardware designers and system architects. The industry's transition to higher 5G frequency bands, especially mmWave, requires more sophisticated and unconventional hardware solutions, including RF chain architecture, phase shifters, and power amplifiers.

The RF chain architecture in the context of 5G beamforming systems must be optimized for performance, economic impact, and power consumption. In multi-tier MIMO configurations, the number of RF chains directly impacts the overall complexity and power requirements of the system. In response to these issues, new approaches such as hybrid beamforming structures have been developed for implementation in both analog and digital architectures.

Hybrid beamforming divides the number of required RF chains and uses both analog phase shifters and digital processing. This approach is more cost-effective to implement while still allowing for simultaneous service to multiple users. The available bandwidth required by 5G systems must also be considered when designing RF chains. Therefore, high-speed data converters and low-noise amplifiers capable of operating at millimeter waves must be integrated.

Phase-shifting technologies

Phase shifters are one of the most important components of beamforming systems because they control the phase of the transmitted or received signal at a specific antenna element. In 5G applications, phase shifters must operate at high frequencies with minimal insertion loss and high linearity. To meet these requirements, various technologies have been investigated, including the use of microelectromechanical systems (MEMS)-based phase shifters, ferroelectric materials, and CMOS manufacturing technology.

MEMS-based phase shifters offer excellent results in terms of insertion loss and operating linearity, but their reliability and conditioning are limited as the frequency increases into the millimeter-wave range. Pulsed ferroelectric phase shifters offer good price-performance but can be nonlinear at high power levels. The attached M-Series CMOS phase shifters have gained widespread acceptance due to their compatibility with manufacturing features and their suitability for low-cost, high-volume applications.

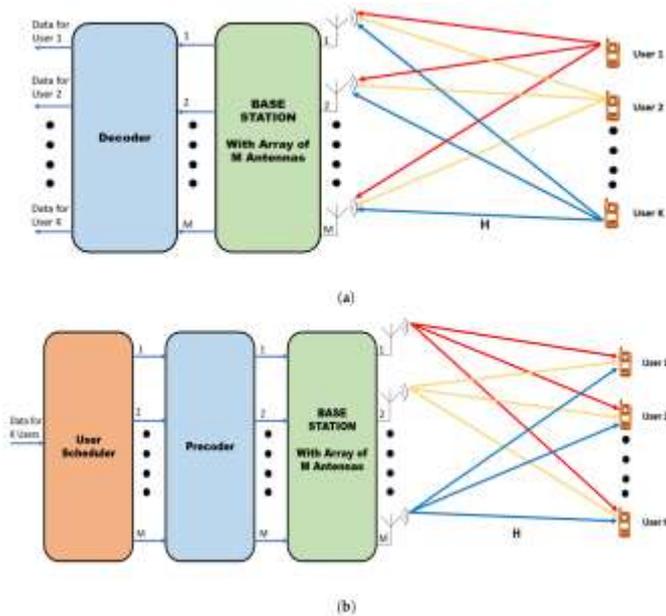


Figure 3: Hardware Implementation Aspects

Another method for measuring fairness is the Jain Fairness Index (JFI), which incorporates the aggregate rate of users from previous scheduling decisions into the calculation of preferences for future scheduling periods. This ensures that all users, especially those who will face challenging channel conditions in the future, receive satisfactory service over time (Figure 2).

In 5G networks, scheduling decisions must be made within a 1 ms time interval. Machine learning-based user programming approaches are considered. These algorithms are designed to optimize network load when fair and work with nodes that have only partial or outdated knowledge of the channel. MUMIMO techniques are expected to continue to improve with the evolution of 5G network environments. Other research areas include user clustering, user interference, and fair resource allocation. These advances will be critical to achieving the expected benefits of 5G technology: the provision of high-capacity, low-latency networks for a wide range of applications.

Power Amplifier Efficiency

The efficiency of the power amplifier (PA) plays a crucial role in the energy efficiency of 5G base stations. For general MIMO transmission, the relative hardware complexity of APs is not significant, but in widespread application scenarios of massive MIMO base stations, the power consumption of APs is a significant drawback. Conventional PA designs cannot maintain high efficiency for PAPR signals operating in 5G networks most of the time. To address this challenge, monolithic or multi-stage PA structures such as Doherty amplifiers and envelope tracking techniques are available for millimeter-wave frequencies. These approaches aim to improve efficiency over a wide range of output power levels, which is essential for working with fluctuating 5G signals. Furthermore, reports on the implementation of GaN-based PAs have demonstrated that efficiency and power density can be achieved at millimeter-wave frequencies. Beamforming is part of the 5G system and must be implemented in hardware, meaning the components are interdependent systems.

When proposing new solutions that can function in next-generation wireless networks, designers must define which types of features are associated with high performance and value, and which energy costs are acceptable. As the 5G communications industry continues to push the boundaries of mmWave communications, advances in RF chain design, phase shifting, and improved power amplifier efficiency are critical to fully exploiting the potential of beamforming across the network.

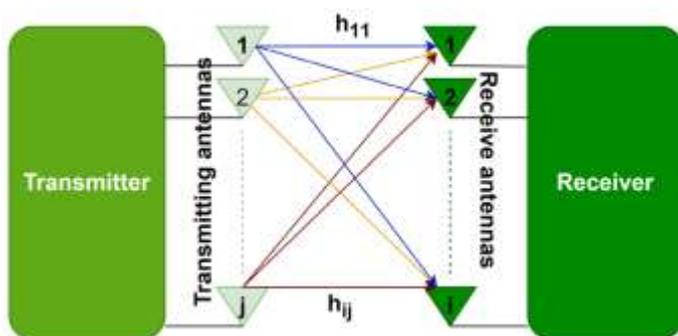


Figure 4: System-Level Optimization

System-Level Optimization

To fully realize the benefits of beamforming technology in 5G networks, system-level optimization is essential. Optimizing these systems can be achieved in different ways, and different strategies can be implemented to increase the effectiveness and

efficiency of the massive MIMO systems that support them.

Cell-free Massive MIMO

Cell-free Massive MIMO (CF-mMIMO) is one of the most promising technologies for 5G wireless networks and beyond. This approach involves distributing a large number of access points (APs) across a large coverage area, allowing each user equipment (UE) to connect to multiple APs simultaneously. Unlike other macrodiversity techniques, the CF-mMIMO architecture offers small cell-like features such as a massive MIMO network with zero inter-cell interference, as well as user-based JT-CoMP, which provides a strong signal with path loss protection. The CFmMIMO system is very useful for achieving a nearly equal quality of service (QoS) for all connected UEs in a given geographic area. This is achieved through various power control methods that isolate driver contaminants and enhance the power of the desired received signal. Furthermore, CF-mMIMO systems can significantly improve energy efficiency (EE) in terms of operating ratio and the overall power consumed by the network. MIMO Network

Network MIMO, also known as Multi-Cell MIMO, generalizes the principle of Multi-User MIMO by applying it to the entire network and allowing multiple base stations to cooperate in both transmit and receive. This approach optimizes the use of available network resources and improves SE, as multiple base stations can be modeled as a distributed antenna. In the case of 5G beamforming and MIMO networks, it is useful to explore ways to minimize intercell interference and increase system capacity. When the beamforming strategies of multiple base stations are synchronized, Network MIMO can provide narrower coverage areas and thus higher SINR values for UEs at the edge of networks.

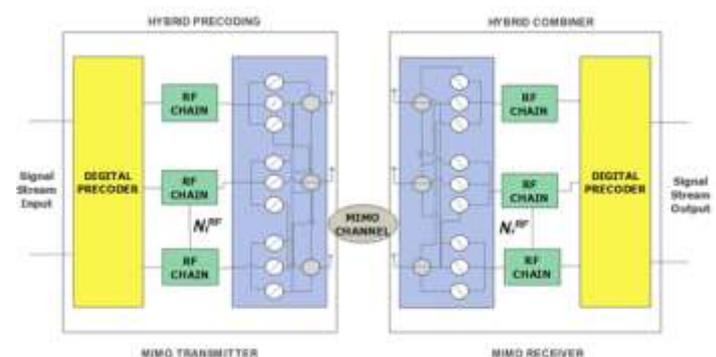


Figure 5: System model for massive MIMO system with hybrid beamforming

Inter-Layer Design

Inter-layer design is critical for optimizing system beamforming in the 5G network. This approach aims to improve multiple layers of the network stack simultaneously. For millimeter wave and beamforming, a multilayer design can further improve some aspects of the system, such as CSI updating, resource allocation, and especially user scheduling. Another area of multilayer design with great potential for impact is the integration of beamforming with other non-orthogonal multiple access techniques. Once integrated, these two technologies improve spectral efficiency and enable the support of more users on the same time-frequency resource.

Power consumption is a major concern in multilayer design approaches for wireless networks. This is achieved in a way that provides an optimal balance between spectral efficiency, power consumption, and the complexity of the implemented hardware. For example, systems that use both analog and digital beamforming are preferable because they strike a balance between performance and power consumption, especially in the millimeter wave system.

In this article, we demonstrate how optimizing the network at all layers can fully exploit the benefits of beamforming in 5G systems. This requires the implementation of next-generation technologies such as CF-mMIMO, Network MIMO, cross-layer design, etc. to increase coverage, capacity, and energy efficiency in a variety of use cases and scenarios.

V. CONCLUSION

5G beamforming is revolutionizing wireless communications in profound ways. Along with the widespread adoption of massive MIMO systems, this revolutionary technology represents a fundamental shift in signal processing, enabling higher speeds, greater coverage, and better spectrum utilization. These recent advances in antenna arrays, channel estimation, and beamforming algorithms are indicators of an increasingly integrated and interactive wireless environment. In conclusion, it is important to emphasize that further scientific research on the topics covered in this article is needed to overcome the difficulties associated with implementing such systems.

While the obvious steps are on the hardware side, there is more to it than just changes at this level. With the development of future 5G networks, improvements in beamforming technology will be instrumental in building a new generation of

wireless environments and create countless opportunities for the development of new technologies across all sectors.

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