



Interference Alignment Technique to Improve Energy Efficiency in MIMO-IFBC

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ABSTRACT: Characterising the energy efficiency (EE) and spectral efficiency (SE) performance of multiple-input–multiple-output interfering broadcast channels (MIMO-IFBC) for the design of green wireless system. In this work, develop a new network architecture based on EE and SE maximization for Multi-Cell MIMO-IFBC within the context of interference alignment (IA). This work develops two methods to optimize energy efficiency for different signal-to-noise ratio regions. For high SNR operating regions, develop a grouping-based interference alignment scheme to jointly cancel intra- and inter-cell interferences and thus transform the MIMO-IFBC to a single-cell MIMO broadcast channel. A gradient-based power adaptation scheme is developed based on singular value decomposition and water filling power adaptation to maximize energy efficiency for each cell. For moderate SNR cases, employ an approach using dirty paper coding (DPC) based on the principle of multiple access channel and broadcast channel duality to perform interference alignment while maximizing EE in each cell. The algorithm in its dual form is solved by using a subgradient method and a bisection searching scheme. However, the extent to which interference can be aligned over a finite number of signalling dimensions remains unknown. The important concern for interference alignment scheme is the requirement of global channel knowledge. So this work can achieve interference alignment with only local channel knowledge at each node. In the proposed work determining the sum rate based on channel coefficients and can calculate spectral efficiency at high signal to noise region. By using maximum SINR algorithm, analyse sum rate Vs signal to noise ratio to get the maximum achievable spectral efficiency. Simulation results give the performance of the proposed schemes over several existing approaches.

KEYWORDS: Energy efficiency, Spectral efficiency, interference alignment, Water filling, dirty paper coding.

I. INTRODUCTION

Dense deployment of base stations (BSs) with multiple antennas is considered as a solution for supporting the projected massive data traffic growth. This trend causes ever-rising network power consumption which has severe implications in terms of both operational and environmental costs. Mobile operators are consequently encompassed with the difficult but fascinating challenge of improving spectral efficiency (SE) and energy efficiency (EE) of network infrastructure at the same time. Traditionally, SE has been the sole performance indicator for the design of wireless communication networks [1]. It is an important measure for quantifying the effectiveness of cellular systems and has been extensively studied for various technologies and scenarios. A prominent transmission technology for the next generation of cellular networks such as long-term-evolution advanced (LTE-A) is multiuser (MU) multiple-input multiple-output (MIMO) which facilitates multiplexing data streams across multiple users. The fundamental challenge for multi-cell scenario is the mitigation of inter- and intra-cell interferences. Random beamforming is a practically good transmission scheme for multiuser multi-antenna downlink systems and a closed-form expression of the achievable average sum-rate in a MISO system. Considering multiple antennas at the mobile stations, a capacity-maximizing zero-forcing (ZF) scheme for two mutually interfering broadcast channels (IFBC) is proposed. Furthermore, the authors proposed a grouping based interference alignment (IA) technique for a network with multi-user MIMO under a Gaussian IFBC scenario with multiple BSs [11]. Compared to the zero forcing scheme in multi-cell scenario, the grouping based interference alignment scheme requires fewer transmit antennas at the BS. Hence if the same number of antennas is used, the grouping-based approach will have extra spatial dimension than the zero forcing scheme, which will result in diversity gain. An energy efficient transmission strategy for multi-cell MU-MISO downlink system by jointly optimizing the

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transmit powers and beamforming vectors was studied. However there is a general lack of research study on the EE aspect of multi-cell MU-MIMO system, which is a very practical scenario. There is only which extended the energy efficient optimization problem to a partial-cooperative multi-cell MU-MIMO system by employing the interference zero-forcing (I-ZF) technique

II.SYSTEM ANALYSIS

In this section, introduce the system model of MIMOIFBC and formulate the EE optimization problem.

Consider a downlink cellular system with multiple cells (L) and each cell serves K multiple users with N_k transmit and N_r receive antennas, corresponds to a multiple input multiple output interfering broadcast channels scenario. The information transmitted from each BS to any of its respective users consists of d_s data streams where $d_s < \min(N_k, N_r)$. In the l th cell, the intended signal from the serving BS to the user is expressed as

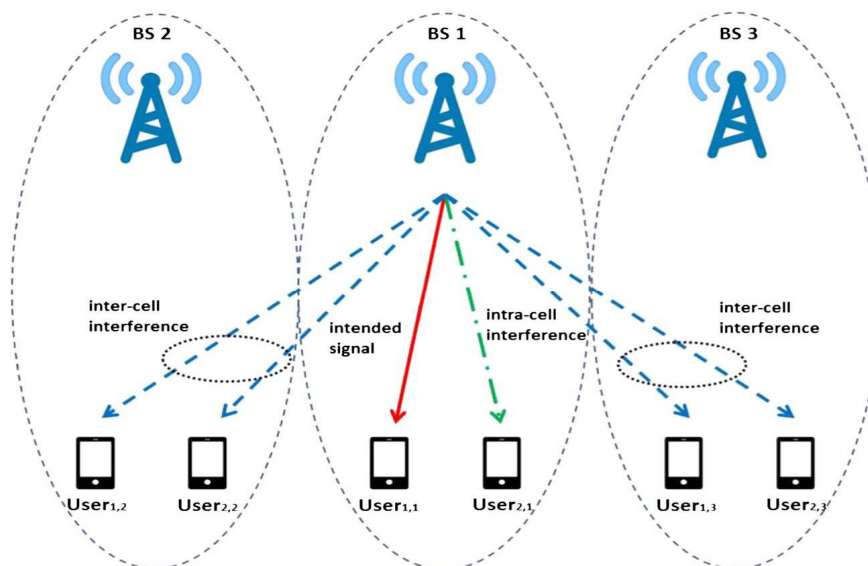


Fig 1: System model of MIMO-IFBC with three cells

$$x_{[k,l]} = V_{[k,l]} s_{[k,l]} \quad (1)$$

The received signal in k th user at l th cell is defined as

$$y_{[k,l]} = \sum_{i=1}^L H_{[k,l]}^i \sum_{j=1}^k x_{[j,i]} + n_{[k,l]} \quad (2)$$

The received signal for each user is multiplied by a receiver beamforming matrix to obtain the desired signal from its corresponding BS. It is well-known that BSs are the primary source of energy consumption in cellular networks. Due to the recent advances in circuit technology, it has been made possible for wireless transceivers to consume different power levels in different operational modes. These include BS sleep, idle, transmit, and receive modes which can be accordingly adjusted based on the daily fluctuations in network load for the purpose of saving energy.

Conventional energy efficiency for downlink transmission is defined as the total number of delivered bits per energy, where energy consumption includes transmission energy consumption and circuit energy consumption in active mode. Hence, we define energy efficiency of the l th cell in multi-Cell multiple input multiple output interfering broadcast channels as



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$$\lambda_{EE}^{[l]} \cong \frac{C_{BC}^{[l]}}{P} = \frac{\sum_{k=1}^K C_{[k,l]}}{\zeta P_T^{[l]} + P_c} \quad (3)$$

where $C_{[k,l]}$ is the capacity achieved by the k th user in the l th cell and P_T^l is the total transmission power for BS l . The objective of this paper is to maximize the energy efficiency of each cell in multi-Cell multiple input multiple output interfering broadcast channels whilst achieving a desirable throughput. It is therefore reasonable to maximize energy efficiency subject to satisfying a minimum throughput requirement. The optimization problem for cell l can be formulated as

$$\max \quad \lambda_{EE} \quad (4)$$

$$\text{S.t} \quad \sum_{k=1}^K P_{[k,l]} \leq P_{\max}^{[l]} \quad (5)$$

$$\sum_{k=1}^K C_{[k,l]} \geq \delta_{\min}^{[l]} \quad (6)$$

where $P_{\max}^{[l]}$ and $\delta_{\min}^{[l]}$ are the maximum total transmit power constraint at BS l and minimum throughput constraint for cell. Due to the existence of inter-channel interference and intra-cell interference in multi-Cell multiple input multiple output interfering broadcast channels, the solution of the above problem is nontrivial and cannot be solved directly. Therefore, in the following sections, we develop resource allocation schemes for interference alignment based systems to solve the above optimization problem.

III. ENERGY EFFICIENCY IN MULTI-CELL MIMO-IFBC WITH INTERFERENCE ALIGNMENT

In this section, design interference mitigation techniques in order to jointly remove the intra- and inter-cell interferences. Therefore, to remove both intra- and inter-cell interferences, grouping-based interference alignment solution is capable of removing both the intra- and intercell interferences but with less transmit antennas at each BS. As a result, we employ the grouping-based interference alignment solution from in this work. The interference alignment scheme under consideration group users to a subspace and then designs the receiver beamforming matrices. Once the effective inter- and intra-cell interfering channels are identified, we design the transmitter beamforming matrices. The interference alignment scheme using the grouping method can ensure zero intra- and inter-cell interference at the receiver of each user. Considering the k th user in the l th BS, the effective channel after applying interference alignment is

$$\bar{H}^{-1} = U_{[k,l]}^H H_{[k,l]}^l V_{[k,l]} \quad (7)$$

Since there does not exist any intra- and inter-cell interference, the multiple-cells with multiple MIMO users scenario has been transformed to a single-cell single user MIMO case. SVD precoding is known to achieve the MIMO channel capacity since the transmitter emits multiple streams in the eigen-directions of the channel covariance matrix. The SVD of the effective channel is given by

$$\bar{H}^{-1}_{[k,l]} = U_{[k,l]}^s \sum_{[k,l]}^l V_{[k,l]}^{sH} \quad (8)$$

After applying IA and SVD, the rate achieved by this user is

$$r_{[k,l]} = \sum_{i=1}^{d_c} B \log_2 (1 + P_{[k,i]}^l g_{[k,i]}^l) \quad (9)$$

where B represents the transmission bandwidth, $p_{[k,i]}^l$ denotes the power allocated on the i th data stream of the k th user, $g_{[k,i]}^l$ denotes the effective channel-gain-to-noise ratio of user $[k,l]$



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The maximum achievable energy efficiency at a certain total transmit power, P_T , is achieved with transmit power P_T , is achieved with transmit power that satisfy the constraints namely,

$$\lambda_{EE}(P_T) = \max_{p[k,i]>0} \frac{\sum_{k=1}^K r_k}{\zeta \sum_{k=1}^K \sum_{i=1}^{d_s} p[k,i] + p_c} \quad (10)$$

The next step is to find out the optimal power allocation scheme to maximize sum-rate. the optimal power can be calculated using the following water-filling scheme

$$\bar{p}_{[k,i]} = \left(\mu_s - \frac{1}{g_{[k,i]}} \right) \quad (11)$$

$$\sum_{k=K} \sum_{i \in S_i} W \log_2(\mu_s g_{[k,i]}) = \delta_{\min} \quad (12)$$

$$\tilde{p}_{[k,i]} = \bar{p}_{[k,i]} + \left(\mu - \frac{1}{g_{[k,i]}} \right) \quad (13)$$

$$\sum_{k \in K} \sum_{i \in S_i} \left(\mu - \frac{1}{g_{[k,i]}} - \bar{p}_{[k,i]} \right) = P_T - \sum_{k \in K} \sum_{i \in S_i} \bar{p}_{[k,i]} \quad (14)$$

Gradient based optimal power adaptation algorithm

1. Do single user water filling to get $\bar{p}_{[k,n]}$ and μ_s . Calculate the power consumption P_o .
2. If $P_o > P_{\max}$
3. infeasible
4. ELSE
5. Initial power $P_T(0) \in [P_o, P_{\max}]$
6. REPEAT
7. For the remaining power, do water filling
8. Update transmission power using gradient of EE
9. STOP when $|p_T(n) - p_T(n-1)| \leq \epsilon$
10. STOP

The idea of the water-filling process includes two steps. We first allocate power to satisfy the minimum rate requirement. The total power used in the first step is P_o . It must be noted that if the power used to satisfy the minimum rate requirement is larger than the power budget, i.e., $P_o > P_{\max}$, the energy efficiency optimization problem is infeasible. We then allocate the remaining power ($P_T - P_o$) using to further maximize the sum rate. Consequently, we can combine a derivative-assisted gradient scheme and water-filling approach to obtain the optimal solution. We call this algorithm as gradient based power adaptation approach.

V. EFFICIENCY IN MULTICELL MIMO-IFBC WITH HYBRID INTERFERENCE ALIGNMENT

The interference alignment scheme presented fully utilizes the available degrees of freedom for transmission. The grouping based interference alignment method is therefore best suited, and in fact capacity optimal, in high SNR operating regimes. On the other hand, pure interference alignment may lead to lower network capacity in low to intermediate SNR region due to the lack of coherent array gain. Hence, we propose a hybrid approach by applying interference alignment only for the inter-cell users and tackle the intra-cell interference among users by using dirty paper coding, which is a capacity-achieving scheme for MIMO-BC, to maximize energy efficiency of each cell.



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Authors proved that the capacity region of the MIMO MAC with a total power constraint P for all K transmitters is equal to the dirty paper region of the dual MIMO broadcast channel with power constraint P. Also any rate vector that is in the dirty paper region of the BC is also in the dual MIMO MAC region with the same total power constraint. Hence, the dirty paper region of a MIMO BC with power constraint P is equal to the capacity region of the dual MIMO medium access channel with total power constraint P. We develop a more efficient two-layer scheme to obtain the maximum energy efficiency in a MIMO-BC scenario. Therefore, the problem can be decomposed into two layers and solved iteratively through the following processes.

- Inner-layer: For a given transmission power, P_T , finds the maximum energy efficiency
- Outer layer: Finds the optimal energy efficiency, η , via a gradient based algorithm.

Defining $f(Q_1^m, \dots, Q_K^m) = \log |I_{N_r \times N_r} + 1/\sigma^2 \sum_{k=1}^K H_k^H Q_k^m H_k|$ we rewrite the optimization problem as

$$\max_{Q_k^m} f(Q_1^m, \dots, Q_K^m) \quad \text{s.t.} \quad \sum_{k=1}^K \text{Tr}(Q_k^m) = P_T \quad (15)$$

Since the positive semi-definiteness of Q_m^k is equivalent to the non-negativeness of the eigen-values of Q_m^k . In this work, we use an iterative method to obtain the optimal Q_m^k for the dual MAC problem. Q_m^k is updated using the

gradient based on Q_m^k as follows $\nabla Q_k^m L = I_{N_r \times N_r} - \eta df \frac{[Q_1^m(n), \dots, Q_{k-1}^m(n), Q_k^m(n-1), \dots, Q_K^m(n-1)]}{\partial Q_k^m(n-1)}$

(16)

The gradient can be promptly determined as

$$\frac{\partial f(Q_1^m, \dots, Q_K^m)}{\partial Q_k^m} = H_k (I_{N_r \times N_r} + 1/\sigma^2 \sum_{k=1}^K H_k^H Q_k^m H_k)^{-1} H_k^H \quad (17)$$

Bisection based resource allocation algorithm

1. Initialise η_{\min} and η_{\max} ;
2. REPEAT
3. $\eta = (\eta_{\min} + \eta_{\max})$
4. REPEAT. Initialise $Q_1^m(0), \dots, Q_K^m(0), n = 1$;
5. FOR $k=1, \dots, K$
6. $Q_k^m(n) = |Q_k^m(n-1) + t \nabla Q_k^m L|^+$
7. END FOR;
8. $n = n + 1$;
9. UNTIL Q_k^m for $k=1, \dots, K$ converge i.e.;
10. $\|\nabla Q_k^m\|^2 \leq \epsilon$ for a small present ϵ
10. $\sum_{k=1}^K \text{Tr}(Q_k^m) > P_T, \eta_{\min} = \eta$
- Else if $\sum_{k=1}^K \text{Tr}(Q_k^m) < P_T, \eta_{\max} = \eta$



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11. UNTIL $|\eta_{\min} - \eta_{\max}| \leq \epsilon$

After the dual MAC covariance matrices Q_k^m are determined for all users, we need to obtain the optimal η . Due to the convexity property of the Lagrangian function $g(\eta)$, the optimal η can be determined through a one-dimensional search process. However, since $g(\eta)$ is not necessarily differentiable, the gradient algorithm is not suitable in this case. Alternatively, we can apply the subgradient method to find the optimal solution. In each iterative step η is updated according to the subgradient. Upon convergence of the transmit covariance matrix Q_k^m we compare the current transmission power in dual MAC with P_T and decrease otherwise. This process will continue until converges. The complexity of dirty paper coding are based on QR decomposition in while the complexity of grouping interference alignment is based on . We calculate the computational complexity based on the number of floating points . As can be seen from the I-ZF scheme proposed in has a higher computational complexity compared to the two proposed schemes

VI. SPECTRAL EFFICIENCY USING MAXIMUM SINR ALGORITHM

In the proposed work results establish the optimality of interference alignment to approach the Shannon capacity of interference networks at high SNR. However, the extent to which interference can be aligned over a finite number of signalling dimensions remains unknown. Another important concern for interference alignment scheme is the requirement of global channel knowledge. So one can achieve interference alignment with only local channel knowledge at each node. The recent emergence of the idea of interference alignment for wireless networks has shown that the capacity of wireless networks can be much higher than previously believed. For the Gaussian interference channel with K interfering transmitters-receiver pairs with each transmitting and receiving node having M antennas each, and with random, time varying channel coefficients drawn from a continuous distribution, reference characterizes the network sum capacity. Here SNR is defined as the total transmit power of all the transmitters in the network when the local noise power at each receiving node is normalized to unity. Here this work determining the sum rate based on channel coefficients and we can calculate spectral efficiency at high signal to noise region. By using maximum SINR algorithm one can analyse sum rate Vs signal to noise ratio to get the maximum achievable spectral efficiency

VII. SIMULATION RESULTS

In this section, this work present simulation results to verify the theoretical findings and analyze the effectiveness of the proposed algorithms in terms of energy efficiency. We refer to the scheme in SVD-IA and the approach in DPC-IA. All the cells are ordered in an warped around linear array and each BS is surrounded by uniformly-distributed users. The drain efficiency of the power amplifier is set to 38 percentage in our simulation. Signal to noise ratio is defined as P_{\max}/σ^2 and is set to 10 dB. Unless stated otherwise, the power budget for each BS is set to 46 dBm while the minimum cell throughput requirement is set to 100 Mbps. We assume there are three cells with three users in each cell, and BS transmit $d_s=2$ data streams for each user . We evaluate the energy efficiency to transmission power relationship for the case with different total circuit power ($P_c = 5, 10, 20$ W) for both SVD-IA and DPC-IA.

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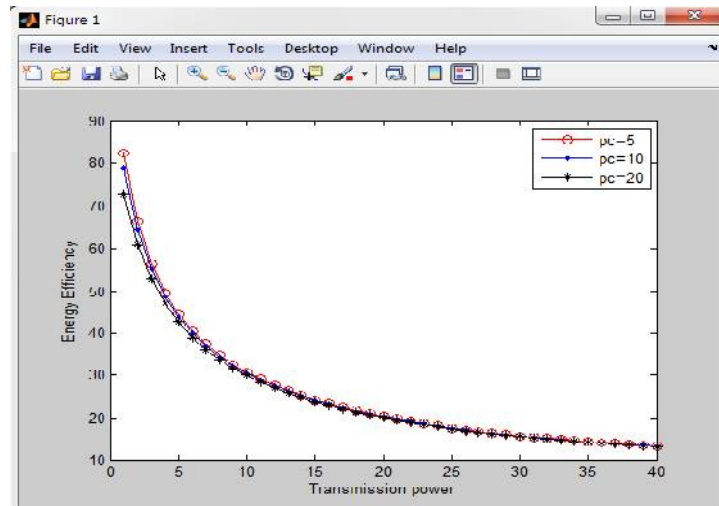


Fig 2:Energy efficiency Vs transmit power using water filling and dirty paper coding scheme

It can be seen from Fig. 2 that the energy efficiency transmission power relationship has an abell shape curve and is quasiconcave. This quasiconcavity is the foundation of the proposed methodology and infers that the proposed water-filling based resource allocation algorithm for SVD-IA always leads to the maximum energy efficiency performance. It also implies that the proposed bisection based resource allocation algorithm for DPC-IA can serve as an optimal inner layer step for energy efficiency maximization. From there, as expected, λ_{EE}^{opt} decreases with increased circuit power due to the higher power consumption. On the other hand, we can observe that the respective P_T^{opt} increases.

In the proposed work, determining the sum rate based on channel coefficients and we can calculate spectral efficiency at high signal to noise region. By using maximum SINR algorithm we can analyse sum rate Vs signal to noise ratio to get the maximum achievable spectral efficiency. It can be seen from Fig. 3 that the sum rate and SINR has a linear shape curve and is quasiconcave. This quasiconcavity is the foundation of this proposed methodology and infers that the proposed maximum SINR algorithm leads to the maximum spectral efficiency performance.

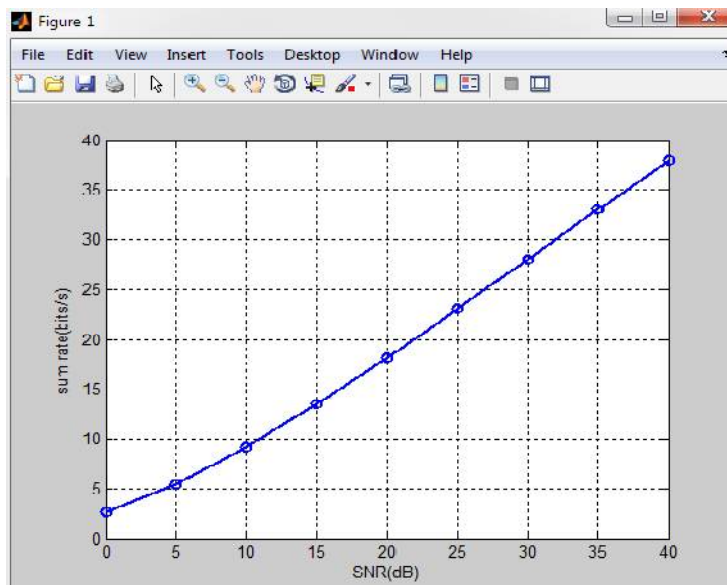


Fig 4:Sum rate Vs SINR using maximum SINR algorithm



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VIII. CONCLUSION

In this work, addressed the energy efficiency optimization problem for multi-cell MIMO-IFBC with interference alignment. This work developed two schemes to optimize energy efficiency for different SNR regions. For high SNR region, employ a grouping-based interference alignment scheme to cancel inter- and intra-cell interference and transform the MIMO-IFBC to a single-cell single user MIMO scenario. A gradient-based power adaptation scheme has been proposed based on the water filling approach and SVD to maximize energy efficiency for each cell.

Here developed a new network architecture proposition based on energy efficiency and spectral efficiency maximization for Multi-Cell MIMO-IFBC within the context of interference alignment (IA). In the proposed work the extent to which reference can be aligned over a finite number of signalling dimensions remains unknown. Another important concern for interference alignment scheme is the requirement of global channel knowledge. So one can achieve interference alignment with only local channel knowledge at each node and determined the sum rate based on channel coefficients and calculate spectral efficiency at high signal to noise region. By using maximum SINR algorithm, analysed sum rate Vs signal to noise ratio to get the maximum achievable spectral efficiency. Simulation results demonstrate that the proposed schemes outperform several existing approaches in all SNR range

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BIOGRAPHY

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