On Multi-Cell Uplink-Downlink Duality with Treating Inter-Cell Interference as Noise

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Abstract—We consider the information-theoretic optimality of treating inter-cell interference as noise in downlink cellular networks modeled as Gaussian interfering broadcast channels. Establishing a new uplink-downlink duality, we cast the problem in Gaussian interfering broadcast channels to that in Gaussian interfering multiple access channels, and characterize an achievable GDoF region under power control and treating inter-cell interference as (Gaussian) noise. We then identify conditions under which this achievable GDoF region is optimal.

I. INTRODUCTION

The age-old robust interference management strategy of power control and treating interference as noise (TIN) in wireless networks has recently been given renewed vitality, attracting increasing attention of late [1]–[9]. This revived interest is largely due to the findings of Geng et al. [1], who showed that power control and TIN are sufficient to achieve the entire generalized degrees-of-freedom (GDoF) region, and capacity region to within a constant gap, of the $K$-user Gaussian interference channel (IC) in a broad regime of parameters, described in terms of channel strength levels. The approach and results of Geng et al. were generalized and extended in a number of directions reported in [2]–[9].

The TIN framework of [1] was recently extended to multi-cell networks in uplink scenarios [9], modeled as the Gaussian interfering multiple access channel (IMAC). TIN is defined in [9] for such setting as the employment of “a MAC-type, capacity-achieving strategy, with Gaussian codebooks and successive decoding...in each cell while treating all inter-cell interference as noise...complemented with power control to manage inter-cell interference”. Under this TIN scheme, the achievable GDoF region (with no time-sharing) was explicitly characterized as a finite union of polyhedra, and broad GDoF-friendly model (see [1]), i.e. each cell employs a power-controlled, degraded BC-type, capacity-achieving strategy, with superposition coding and successive decoding, while treating inter-cell interference as (Gaussian) noise. We establish a new uplink-downlink duality between IBCs and IMACs in the sense that the corresponding TIN-achievable GDoF regions are unchanged if the roles of transmitters and receivers are switched. This duality, in conjunction with the results in [9], leads to an explicit characterization of the TIN-achievable GDoF region for the IBC. We further identify conditions under which this achievable GDoF region is also optimal.

Notation: For positive integers $z_1$ and $z_2$ where $z_1 \leq z_2$, the sets $\{1, 2, \ldots, z_1\}$ and $\{z_1, z_1+1, \ldots, z_2\}$ are denoted by $\langle z_1 \rangle$ and $\langle z_1 : z_2 \rangle$, respectively. For any $a \in \mathbb{R}$, $(a)^+ = \max\{0, a\}$. Bold lowercase symbols denote tuples, e.g. $a = (a_1, \ldots, a_z)$. For $\mathcal{A} = \{a_1, \ldots, a_K\}$, $\Sigma(\mathcal{A})$ is the set of all cyclically ordered sequences of all subsets of $\mathcal{A}$ (see [9] Sec. 1.3]).

II. SYSTEM MODEL

Consider a $K$-cell cellular network in which each cell $k$, where $k \in (K)$, comprises a base station denoted by BS-$k$ and two user equipments, each denoted by UE-$l_k$, $k \in (2)$. The set of tuples corresponding to all UEs in the networks is given by $\mathcal{K} = \{l_k, k \in (2), k \in (K)\}$. For ease of exposition, we limit our attention to the 2-user-per-cell case. Nevertheless, the results can be generally extended to scenarios with an arbitrary number of users in each cell.

A. Interfering Broadcast Channel

When operating in the downlink mode, the above network is modeled by a Gaussian IBC, e.g. Fig. 1 (left). Adopting a GDoF-friendly model (see [1]), the input-output relationship at the $t$-th use of the channel, where $t \in \mathbb{N}$, is described as

$$Y_k^{[t]}(t) = \sum_{i=1}^{K} h_{ki}^{[t]} X_i(t) + Z_k^{[t]}(t)$$

$$= \sum_{i=1}^{K} \sqrt{P_{\mathrm{tx}}^{[t]} e^{j\phi_{ki}^{[t]}}} X_i(t) + Z_k^{[t]}(t),$$

where $Y_k^{[t]}(t)$ is the signal received by UE-$l_k$, $h_{ki}^{[t]}$ is the channel coefficient from BS-$i$ to UE-$l_k$, $X_i(t)$ is the transmitted symbol of BS-$i$ and $Z_k^{[t]}(t) \sim \mathcal{N}(0, 1)$ is

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the additive white Gaussian noise (AWGN) at UE-\(\{k_1, k_2\}\). All symbols are complex and each BS-\(i\) is subject to the average power constraint \(\frac{1}{n} \sum_{t=1}^{n} \mathbb{E}[|X_i(t)|^2] \leq 1\), where \(n\) is the duration of the communication. \(\sqrt{P_{\text{BS}}[k]}\) and \(\alpha_{k}[l]\) are the magnitude and phase of the channel coefficient \(h_{k}[l]\), where \(P > 0\) is a nominal power value and \(\alpha_{k}[l] \geq 0\) is the corresponding channel strength level\(^1\) Without loss of generality, we assume that the following order holds
\[
\alpha_{k}[1] \leq \alpha_{k}[2], \forall k \in \{K\}. \tag{2}
\]

Each transmitter in the IBC, e.g., BS-\(k\) for some \(k \in \{K\}\), has the independent messages \(W_{1}[k]\) and \(W_{2}[k]\) intended to UE-\((1, k)\) and UE-\((2, k)\), respectively. Codes, error probabilities, achievable rates \(R = (R_{1}[1], R_{1}[2], \ldots, R_{K}[1], R_{K}[2])\), and the capacity region \(C_{\text{IBC}}\) are all defined in the standard Shannon theoretic sense. A GDoF tuple is denoted by \(d = (d_{1}[1], d_{1}[2], \ldots, d_{K}[1], d_{K}[2])\) and the GDoF region is denoted by \(D_{\text{IBC}}\), where both are defined in the standard fashion.

### B. Dual Interfering Multiple Access Channel

The dual IMAC is obtained by reversing the roles of the transmitters and receivers in the IBC, e.g. Fig. 1 (right). Building upon the GDoF-friendly model in 1, the input-output relationship for the IMAC is given by
\[
Y_i(t) = \sum_{k=1}^{K} \left[ h_{k}[1]X_{k}[1](t) + h_{k}[2]X_{k}[2](t) \right] + Z_i(t), \tag{3}
\]
where \(Y_i(t)\) and \(Z_i(t)\) \(\sim \mathcal{N}(0, 1)\) are the received signal and the AWGN at BS-\(i\) respectively, and \(X_{k}[l](t)\) is the transmitted symbol of UE-\((k, l)\). Each UE-\((k, l)\) is subject to the average power constraint \(\frac{1}{n} \sum_{t=1}^{n} \mathbb{E}[|X_{k}[l](t)|^2] \leq 1\). For cell \(k\), \(k \in \{K\}\), UE-\((1, k)\) and UE-\((2, k)\) have the independent messages \(W_{1}[k]\) and \(W_{2}[k]\), respectively, intended to BS-\(k\). We denote the capacity region and the GDoF region of the above IMAC by \(C_{\text{IMAC}}\) and \(D_{\text{IMAC}}\), respectively.

**Remark 1.** When defining dual uplink channels, it is common to impose a sum transmit power constraint on the UEs so that the total transmit power does not exceed the transmit power of the BS in the downlink channel. We do not impose this on the dual IMAC here, which exhibits a transmit power gain of \(K\). This gain is inconsequential for the GDoF results.

### III. Treating Inter-cell Interference as Noise and Uplink-Downlink Duality

We consider TIN in the cellular sense [9], where each cell employs an adequately modified single-cell capacity-achieving strategy while treating all inter-cell interference as noise.

#### A. TIN in the IBC

Each cell \(k\) of the IBC employs superposition coding and successive decoding according to an order \(\pi_{k}\), which is a permutation on the set \(\{1, 2\}\). In particular, the transmitted signal of BS-\(k\) is composed as
\[
X_k(t) = X_k^{[1]}(t) + X_k^{[2]}(t), \tag{4}
\]
where each message \(W_{k}[l]\), for \(l \in \{2\}\), is independently encoded into the codeword \(X_k^{[l]}(n) \triangleq X_k^{[l]}(1, \ldots, X_k^{[l]}(n))\), drawn from a Gaussian codebook with average power \(1 \sum_{t=1}^{n} \mathbb{E}[|X_k^{[l]}(t)|^2] = q_k\). Note that the powers \(q_{k}\) and \(q_{k}^{[l]}\) satisfy \(q_{k} + q_{k}^{[l]} \leq 1\). On the other end, UE-\((\pi_{k}(1))\) decodes its own signal \(X_{k}^{[\pi_{k}(1)]}\) while treating all other signals (i.e. both intra-cell and inter-cell interference) as noise. UE-\((\pi_{k}(2))\), however, starts by decoding and cancelling \(X_{k}^{[\pi_{k}(1)]}\) before decoding its own signal \(X_{k}^{[\pi_{k}(2)]}\), while treating inter-cell interference as noise.

Using the above scheme, the effective signal-to-interference-plus-noise ratio (SINR) for decoding the signal \(X_{k}^{[\pi_{k}(1)]}\) is denoted by \(\text{SINR}_{k}^{[\pi_{k}(1)]}\) (given by a minimum of two SINRs), while the SINR for \(X_{k}^{[\pi_{k}(2)]}\) is denoted by \(\text{SINR}_{k}^{[\pi_{k}(2)]}\) (SINR expressions are omitted for brevity). For fixed decoding order and power allocation, the message \(W_{k}[\pi_{k}(l)]\) is hence reliably communicated to UE-\((\pi_{k}(k), l)\) at any rate satisfying
\[
0 \leq R_{k}^{[\pi_{k}(l)]} \leq \log \left(1 + \text{SINR}_{k}^{[\pi_{k}(l)]}\right). \tag{5}
\]

For GDoF purposes, we may assume that \(q_{k}^{[l]}\) scales with \(P_{k}\) as \(O\left(P_{k}^{\gamma}\right)\), where \(\gamma \leq 0\) is the corresponding transmit power exponent. It follows that the achievable GDoF tuple in cell \(k\) is given by all \(d_{k}^{[\pi_{k}(1)]}\) and \(d_{k}^{[\pi_{k}(2)]}\) that satisfy (6) (top of next page). A power allocation tuple is given by \(\pi = (\pi_{1}^{[1]}, \pi_{1}^{[2]}, \ldots, \pi_{K}^{[1]}, \pi_{K}^{[2]}) \leq \mathbf{0}\) and a network decoding order tuple is given by \(\pi \in \{(\pi_{1}^{[1]}, \pi_{1}^{[2]}, \ldots, \pi_{K}^{[1]}, \pi_{K}^{[2]})\} \leq \mathbf{0}\) where \(\Pi\) is the set comprising all possible network decoding orders. For fixed \(\pi\), we use \(p^{\text{IBC}}(r)\) to denote the set of all GDoF tuples \(d\) with components satisfying (6). The **TIN-achievable GDoF region** with fixed \(\pi \in \Pi\) is obtained by taking the union over all feasible power allocations, i.e. \(p^{\text{IBC}}(\pi) = \bigcup_{\pi} p^{\text{IBC}}(r)\). The general TIN-achievable GDoF region for the IBC, denoted by \(p^{\text{IBC}}\), is obtained by further considering all possible decoding orders in \(\Pi\) and is defined as
\[
p^{\text{IBC}}(\pi) \triangleq \bigcup_{\pi \in \Pi} p^{\text{IBC}}(r). \tag{7}
\]
\[0 \leq d_k^{(1)} \leq \max \left\{ 0, \min_{m_k \in \{1,2\}} \left\{ \alpha_k^{(m_k)} + r_k^{(1)} \right\} - \left( \max \left\{ \alpha_k^{(m_k)} + r_k^{(2)} \right\}, \max_{(l,j), j \neq k} \left\{ \alpha_k^{(m_k)} + r_j^{(l)} \right\} \right) \right\} \]

\[0 \leq d_k^{(2)} \leq \max \left\{ 0, \alpha_k^{(2)} + r_k^{(2)} - \left( \max_{(l,j), j \neq k} \left\{ \alpha_k^{(2)} + r_j^{(l)} \right\} \right) \right\}. \]

As time-sharing is not allowed, each GDoF tuple \( d \in \mathcal{P}^{IBC*} \) is achieved through a strategy identified by a decoding order and a power allocation tuple, i.e. \((\pi, r)\).

### B. TIN in the Dual IMAC

For the dual IMAC, we adopt the TIN scheme given in [9]. We highlight the aspects most relevant to this paper. Readers are referred to [9] for a more detailed exposition.

Each UE-\((l_k,k)\) uses an independent Gaussian codebook with power \( p_k^{l_k} \), where the transmit power exponent \( j_{l_k}^{l_k} \) and the power allocation tuple \( r_k \leq 0 \) are defined similarly to their counterparts in the IBC. Each BS-\(k\) successively decodes its in-cell signals according to the order \( \pi_k \), such that \( X_k^{\pi_k(2)} \) is decoded and cancelled before decoding \( X_k^{\pi_k(1)} \). As for the IMAC, the network decoding order tuple is given by \( \pi \in \Pi \). For a decoding order \( \pi \) and a power allocation \( r \), UE-\((\pi_k(k), k)\) achieves any GDoF \( d_k^{\pi_k(l_k)\pi} \) satisfying (8) (top of this page).

In the IBC, while fixing \((\pi, r)\), we achieve \( \mathcal{P}^{IMAC*}(r) \) given by all all GDoF tuples \( d \) with components satisfying (8). The TIN-achievable GDoF region, for fixed \( \pi \), is given by \( \bigcup_{r \leq 0} \mathcal{P}^{IMAC*}(r) \), while the general TIN-achievable GDoF region for the dual IMAC is given by

\[ \mathcal{P}^{IMAC*} \triangleq \bigcup_{\pi \in \Pi} \mathcal{P}^{IMAC*}. \]

**Remark 2.** For any given cell \( k \) and permutation \( \pi_k \), the uplink decoding order is the reverse of the counterpart downlink decoding order. This reverse relationship is commonly exhibited in uplink-downlink dualities (see [13], Ch. 10.3.4).

### C. Uplink-Downlink Duality under TIN

Here we present the first result of this paper.

**Theorem 1.** The IBC and IMAC general TIN-achievable GDoF regions \( \mathcal{P}^{IBC*} \) and \( \mathcal{P}^{IMAC*} \) are identical.

To prove Theorem 1, we first consider an arbitrary IBC GDoF tuple \( d \in \mathcal{P}^{IBC*} \). From the earlier parts of this section, we know that there must exist a decoding order \( \pi \in \Pi \) and a feasible power allocation tuple \( r \leq 0 \) such that \( d \in \mathcal{P}^{IMAC*}(r) \).

We show that for the same \( \pi \), there exists \( r \leq 0 \) such that \( d \in \mathcal{P}^{IMAC*}(r) \) also holds. This proves that \( \mathcal{P}^{IBC*} \subseteq \mathcal{P}^{IMAC*} \) in general, as the selected GDoF tuple \( d \) is arbitrary. We then apply a similar argument in the opposite direction and prove that \( \mathcal{P}^{IMAC*} \subseteq \mathcal{P}^{IBC*} \). Details of this proof are presented in Appendix A. It is worthwhile noting that a similar uplink-downlink duality of achievable GDoF tuples and power allocations was shown for the regular \( K\)-user IC in [12].

From the duality in Theorem 1 and the characterization of \( \mathcal{P}^{IMAC*} \) in [9], we obtain a characterization of \( \mathcal{P}^{IBC*} \) as a finite union of polyhedra. Moreover, from [9, Th. 3], we obtain conditions under which \( \mathcal{P}^{IBC*} \) is a polyhedron, i.e. one of the polyhedra in the union includes all others.

### IV. On TIN-Optimality for the IBC

In the following theorem, we obtain TIN-optimality conditions under which the TIN scheme described in Section II-A achieves the entire GDoF region of the IBC.

**Theorem 2.** For the IBC described in Section II-A if the following conditions are satisfied

\[ \alpha_k^{l_i} \geq \max_{j \neq i} \left\{ \alpha_k^{l_j} \right\}, \forall i \in \{K\} \]

\[ \alpha_k^{l_i} \geq \max_{j \neq i} \left\{ \alpha_k^{l_j} \right\} + \max_{(l,k), k \neq i} \left\{ \alpha_k^{l_k} \right\}, \forall (l,i) \in \mathcal{K}, \]

then the optimal GDoF region is given by \( \mathcal{D}^{IBC} = \mathcal{P}^{IBC*} \), which is characterized by all GDoF tuples that satisfy

\[ d_k^{l_i} \geq 0, \forall (l,i) \in \mathcal{K} \]

\[ \sum_{s \in \{l_i\}} d_k^{s_l} \leq \alpha_k^{l_i}, \forall l_i \in \{1,2\}, i \in \{K\} \]

\[ \sum_{j \in \{m\}} \sum_{s \in \{l_i\}} d_k^{s_l} \leq \alpha_k^{l_i} - \alpha_k^{l_i-1}, \forall l_i \in \{1,2\}, \]

\[ (i_1, \ldots, i_m) \in \Sigma(\mathcal{K}), m \in \{2 : K\}. \]

It is readily seen that the IBC TIN-optimality conditions identified in Theorem 2 imply (i.e. stricter than) the counterpart IMAC TIN-optimality conditions identified in [9, Th. 4]. From this observation, it follows that \( \mathcal{P}^{IBC*} \) is characterized by (12)-(14) under such conditions. The rest of the section is hence dedicated to proving the converse part of Theorem 2.

We start with an auxiliary result that plays a key role in the proof. This result essentially shows that under condition (10), UE-\((k,1)\) is more capable than UE-\((1,k)\) in a GDoF sense.

### A. Auxiliary Lemma

Consider the independent input sequences \( X_1^n \) and \( X_2^n \) with average power constraints \( \frac{1}{n} \sum_{i=1}^n \mathbb{E}[|X_i(t)|^2] \leq 1, i \in \{1,2\} \). Let \( Y_0^n \) and \( Y_2^n \) be noisy outputs given by

\[ Y_0(t) = a_1 X_1(t) + a_2 X_2(t) + Z_0(t) \]

\[ Y_2(t) = b_1 X_1(t) + b_2 X_2(t) + Z_2(t) \]

where \( a_1, a_2, b_1, b_2 \in \mathbb{C} \) are constants and \( Z_0(t), Z_2(t) \sim \mathcal{N}_C(0,1) \) are AWGN terms. Finally, let \( W \) be a discrete random variable such that \( W \rightarrow X_1^n \rightarrow Y_0^n \) and \( W \rightarrow X_2^n \rightarrow Y_2^n \) are Markov chains. The following result holds.
Lemma 1. Given that the two conditions \( \frac{|b_1|^2}{p_1} \geq |a_1|^2 \) and \(|b_2|^2 \geq |a_2|^2 \geq 1 \) hold, then

\[
I(X_1^n; Y_0^n | W) \leq I(X_1^n; Y_0^n | W) + n.
\]

The proof of the above result is relegated to Appendix B.

B. Proof of Theorem 2

For each cell \( i \), the inequalities in (13) follow from the capacity region of the degraded Gaussian BC [13]. Therefore, we focus on the cyclic bounds in (14). Cells and users participating in a given cyclic bound are identified by the sequences \( (t_1, \ldots, t_m) \in \Sigma(K) \) and \( (l_1, \ldots, l_m) \in \{1, 2\}^m \). Next, we go through the two following steps:

- Eliminate all non-participating UEs, all non-participating BSs and the corresponding messages.
- Eliminate all interfering links except for links from BS-\( j \) to participating UEs in cell \( i_{j+1} \), for all \( j \in \langle m \rangle \).

We obtain the cyclic (w.r.t cells) IBC modeled as

\[
Y_{ij}(s_{ij}) = h_{ij}^{[s_{ij}]} X_{ij}(t) + h_{ij}^{[s_{ij}]} X_{ij-1}(t) + Z_{ij}(s_{ij}),
\]

for all \( s_{ij} \in \langle l_{ij} \rangle \) and \( j \in \langle m \rangle \). As the above steps do not decrease the rates of all remaining messages, we restrict our attention to the channel in (18) henceforth. We further define the following side information signal

\[
S_{ij}(t) = h_{ij}^{[l_{ij}+1]} X_{ij}(t) + Z_{ij}^{[l_{ij}+1]} (t), j \in \langle m \rangle.
\]

which is eventually provided to the stronger UE in cell \( i_j \).

For cells \( i_j \) with \( l_{i_j} = 1 \), Fano’s inequality yields

\[
n \left( R_{ij}^{[l_{ij}]} - \epsilon \right) \leq I(X_{ij}^n; Y_{ij}^{[l_{ij}]} | S_{ij}).
\]

On the other hand, for cells \( i_j \) with \( l_{i_j} = 2 \), we obtain

\[
n \left( R_{ij}^{[l_{ij}]} + R_{ij}^{[l_{ij}]} - 2 \epsilon \right) \leq I(W_{ij}^n; X_{ij}^{[l_{ij}]}|W_{ij}^n) + I(W_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n)\]
\[
= I(X_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n) + I(W_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n)\]
\[
= I(X_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n) + I(W_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n)\]
\[
\leq I(X_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n) + n + I(W_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n)\]
\[
= I(X_{ij}^n; Y_{ij}^{[l_{ij}]}|W_{ij}^n) + n\]
\[
\leq I(X_{ij}^n; Y_{ij}^{[l_{ij}]}|S_{ij}).
\]

In the above, (21) is obtained through a direct application of Lemma 1 while taking into consideration the TIN-optimality condition in (10). By adding the bounds in (20) and (22) for all \( i_j \), \( j \in \langle m \rangle \), we bound the sum-rate for this cycle as

\[
n \sum_{j \in \langle m \rangle} \sum_{i_j \in \langle l_{ij} \rangle} \left( R_{ij}^{[l_{ij}]} - \epsilon \right) \leq mn + \sum_{j \in \langle m \rangle} I(X_{ij}^n; Y_{ij}^{[l_{ij}]}|S_{ij}).
\]

The bound in (23) is obtained by first noting the setting has essentially reduced to a regular \( m \)-user IC with receivers given by UE-\( i_{j} \), \( j \in \langle m \rangle \), and side information signals as defined in (19), and then applying the steps in [1] Appendix C. Combining (23) with the condition in (11), the corresponding GDOF inequality in (14) is obtained (after rearranging indices).

Remark 3. The capacity outer bound in (23) leads to a constant-gap characterization of the capacity region \( C_{IBC} \) when the TIN-optimality conditions in Theorem 2 hold.

V. CONCLUSION

In this paper, we established an equivalence between the achievable GDoF regions for the IBC and IMAC under power controlled single-cell transmissions, while treating inter-cell interference as noise. This uplink-downlink duality, on the achievability side, leads to a characterization of the IBC TIN-achievable GDoF region. We also identified a regime in which the IBC TIN-achievable GDoF region is optimal. This IBC TIN-optimal regime is included in its counterpart IMAC TIN-optimal regime in [9]. It is of interest to investigate whether the IBC TIN-optimal regime in Theorem 2 can be enlarged to coincide with the IMAC TIN-optimal regime in [9].

APPENDIX A
PROOF OF THEOREM 1

A. \( \mathcal{P}_{\text{IBC}} \subseteq \mathcal{P}_{\text{IMAC}} \)

Consider \( \delta \in \mathcal{P}_{\text{IBC}}(\pi, \alpha) \) with arbitrary \( \pi, \alpha \). We observe that (6) is equivalently expressed as

\[
0 \leq d_k^{[\pi_k(l_k)]} \leq \max \left\{ 0, \epsilon^{[\pi_k(l_k)]} + \beta^{[\pi_k(l_k)]} \right\}, l_k \in \{1, 2\}
\]

where \( \beta^{[\pi_k(l_k)]} \), \( l_k \in \{1, 2\} \), is given by

\[
\beta_k^{[\pi_k(1)]} = \min_{m_k \in \{1, 2\}} \min_{l_j \neq k} \left\{ \alpha_{k,m_k}^{[\pi_k(m_k)]} - \delta_k^{[\pi_k(2)]} \right\},
\]

\[
\beta_k^{[\pi_k(2)]} = \min_{l_j \neq k} \left\{ \epsilon_k^{[\pi_k(2)]} \right\}.
\]

Now consider the following power allocation for the IMAC

\[
\delta_k^{[\pi_k(l_k)]} = -\alpha_k^{[\pi_k(l_k)]} + \beta_k^{[\pi_k(l_k)]}.
\]
As $\beta_k^{[\pi_1]}(1) \leq \min \{ \alpha_{kk}^{[\pi_2]}(1), \alpha_{kk}^{[\pi_3]}(2) \}$ and $\beta_k^{[\pi_2]}(2) \leq \alpha_{kk}^{[\pi_2]}(2)$, the power allocation in (26) is feasible. Using (26), we achieve the set of IMAC GDoF tuples $P_\pi^{IMAC^*} (\bar{r})$ that satisfy

$$d_k^{[\pi_k]}(l_k) \leq \max \{0, \beta_k^{[\pi_k]}(l_k) - \left( \max_{(i,j):j \neq k} \{ \alpha_{ij} \} + \bar{r}_j \right) \}.$$  (27)

From (24) and (27), to show that $d$ is also in $P_\pi^{IMAC^*} (\bar{r})$, it is sufficient to show that $r_k^{[\pi_k]}(l_k) \leq \min \{0, \beta_k^{[\pi_k]}(1) \} \cdot \max_{(i,j):j \neq k} \{ \alpha_{ij} \} + r_j$.

Since $r_k^{[\pi_k]}(l_k) \leq 0$, we only need to show that the inequality in (28) holds for the two other terms in the min{0, ·}. We start with $-\beta_k^{[\pi_k]}(1)$, which only has an influence when $l_k = 2$, as follows:

$$-\beta_k^{[\pi_k]}(1) = \max_{m \in M_{\pi_1}} \max \left\{ -\alpha_{kk}^{[\pi_k]}(m), r_k^{[\pi_k]}(2) \right\} - \alpha_{kk}^{[\pi_k]}(m) + \max_{(i,j):j \neq k} \{ \alpha_{ij}^{[\pi_k]}(m) + r_j \}.$$

Now, next, we observe that for any $(l_j, j)$ with $j \neq k$, we have

$$-\alpha_{ij}^{[\pi_k]}(m) + \max_{(i,j):j \neq k} \{ \alpha_{ij}^{[\pi_k]}(m) + r_j \} \quad \geq r_k^{[\pi_k]}(2).$$

From (29) and (30), we conclude that the inequality in (28) holds. This in turn implies that $P_\pi^{IMAC^*} (\bar{r}) \subseteq P_\pi^{IMAC^*} (\bar{r})$, and hence completes this part of the proof.

B. $P_\pi^{IMAC^*} \subseteq P_\pi^{IBC^*}$

Consider $d \in P_\pi^{IMAC^*} (\bar{r})$ with arbitrary $(\pi, \bar{r})$. We start by observing that, without loss of generality, we may assume $r_k^{[\pi_k]}(2) + \alpha_{kk}^{[\pi_k]}(2) \geq r_k^{[\pi_k]}(1) + \alpha_{kk}^{[\pi_k]}(1)$, $k \in (K)$, as otherwise, there exists another satisfactory power allocation $\bar{r}$ with an achievable GDoF tuple such that $d \geq \bar{d}$ (see end of this section). Next, we observe that (31) can be expressed by

$$0 \leq d_k^{[\pi_k]}(l_k) \leq \max \left\{ 0, \alpha_{kk}^{[\pi_k]}(l_k) + \bar{r}_l^{[\pi_k]}(l_k) - r_k^{[\pi_k]}(l_k) \right\}.$$  (32)

where we define $\gamma_k^{[\pi_k]}(l_k)$, $l_k \in \{1, 2\}$, as follows:

$$\gamma_k^{[\pi_k]}(l_k) = \max \left\{ 0, (\alpha_{kk}^{[\pi_k]}(l_k) + \bar{r}_l^{[\pi_k]}(l_k)) \cdot \max_{(i,j):j \neq k} \{ \alpha_{ij} \} + \bar{r}_j \right\}.$$  (33)

Now consider the IBC power allocation given by

$$r_k^{[\pi_k]}(l_k) = -\gamma_k^{[\pi_k]}(l_k).$$  (34)

which is clearly feasible. Using (34), we achieve the set of IBC GDoF tuples $P_\pi^{IMAC^*} (\bar{r})$ that satisfy

$$0 \leq d_k^{[\pi_k]}(1) \leq \max \left\{ 0, \min_{m \in M_{\pi_1}} \left\{ \max \left\{ \alpha_{kk}^{[\pi_k]}(m) - r_k^{[\pi_k]}(1) \right\} + \left( \max \left\{ \alpha_{kk}^{[\pi_k]}(m) - r_k^{[\pi_k]}(1) \right\} + \bar{r}_j \right) \right\} \right\}.$$  (35a)

$$0 \leq d_k^{[\pi_k]}(2) \leq \max \left\{ 0, \alpha_{kk}^{[\pi_k]}(2) - r_k^{[\pi_k]}(2) \right\} + \left( \max_{(i,j):j \neq k} \{ \alpha_{ij} \} + \bar{r}_j \right) \right\}.$$  (35b)

By examining (32) and (33), $P_\pi^{IMAC^*} (\bar{r}) \subseteq P_\pi^{IMAC^*} (\bar{r})$ is shown by proving that the following inequalities hold

$$\alpha_{kk}^{[\pi_k]}(2) \leq \min_{m \in M_{\pi_1}} \left\{ \max \left\{ \alpha_{kk}^{[\pi_k]}(m) \right\} + \bar{r}_j \right\}.$$  (36a)

$$\alpha_{kk}^{[\pi_k]}(2) \leq \min_{m \in M_{\pi_1}} \left\{ \max \left\{ \alpha_{kk}^{[\pi_k]}(m) \right\} + \bar{r}_j \right\}.$$  (36b)

We start by showing that (36a) holds. For the first of the three terms inside the min{·, ·} in (36a), we note that $\alpha_{kk}^{[\pi_k]}(1) + \bar{r}_j(1) \leq \alpha_{kk}^{[\pi_k]}(m)$, $m \in \{1, 2\}$, holds for (37).

For the second term, we observe that $\alpha_{kk}^{[\pi_k]}(1) + \bar{r}_j(1) \leq \gamma_k^{[\pi_k]}(2)$ follows directly from (33). For the final term, we observe that for any $(l_j, j) \in K$, such that $j \neq k$, we have

$$\alpha_{kk}^{[\pi_k]}(m) + \left( \gamma_k^{[\pi_k]}(1) - r_k^{[\pi_k]}(1) \right) \geq \alpha_{kk}^{[\pi_k]}(m) + \left( r_k^{[\pi_k]}(1) - \gamma_k^{[\pi_k]}(1) \right) \geq r_k^{[\pi_k]}(1).$$

where (38) is obtained by setting $(l_i, i) = (\pi_k(m), k)$ in (37), and the inequality in (39) holds due to (31). Next, we consider (36b). It is clear that the inequality holds for the first of the two terms in the min{·, ·}. The inequality also holds for the second term in the min{·, ·} due to (38). As (36) holds, we have $P_\pi^{IMAC^*} (\bar{r}) \subseteq P_\pi^{IMAC^*} (\bar{r})$, which concludes the proof.

C. Justification for (31)

To show that the assumption in (31) is justified, suppose that the contrary, i.e., $\pi_k^{[\pi]}(2) + \alpha_{kk}^{[\pi]}(2) < \pi_k^{[\pi]}(1) + \alpha_{kk}^{[\pi]}(1)$, holds for some $k \in (K)$. From (31), it follows that in this case we would necessarily have $d_k^{[\pi_k]}(1) = 0$. Now consider an alternative scheme $(\tilde{\pi}, \tilde{r})$, which is a modification of $(\pi, \bar{r})$ such that: $\tilde{\pi}_k(1) = \pi_k(2)$ and $\tilde{\pi}_k(2) = \pi_k(1)$ (i.e. swapping the order in cell $k$), and $\tilde{\pi}_k^{[\pi_k]}(1) = -\infty$, while maintaining the order and power allocation for all remaining cells. For cell $k$, this modified scheme achieves any GDoF pair satisfying $\tilde{d}_k^{[\pi_k]}(1) = 0$ and $0 \leq \tilde{d}_k^{[\pi_k]}(2) \leq \max \left\{ 0, \alpha_{kk}^{[\pi_k]}(2) + \bar{r}_j(2) \right\} - \left( \max_{(i,j):j \neq k} \{ \alpha_{ij} \} + \bar{r}_j \right) \right\}$. Recalling that the decoding
order is swapping in cell $k$, it follows that all GDoF pairs of cell $k$ achieved using $(\pi, \tau)$ are also achievable using $(\tilde{\pi}, \tilde{\tau})$. Moreover, all GDoF pairs of all remaining cells achieved using $(\pi, \tau)$ are also achievable using $(\tilde{\pi}, \tilde{\tau})$. This last statement holds as the power allocation for cells $i \in (K) \setminus \{k\}$ is unaltered, while the transmit power of cell $k$ is reduced in $(\tilde{\pi}, \tilde{\tau})$. Therefore, we have $\mathcal{P}^{\text{IMAC}}(\tilde{\tau}) \subseteq \mathcal{P}^{\text{IMAC}}(\tilde{\pi})$. We apply the argument to all cells $k$ that violate (31).

**Appendix B**

**Proof of Lemma 1**

First, we observe that $I(X_1^n; Y_a^n)$ is bounded above as

$$I(X_1^n; Y_a^n) \leq I(X_1^n; a_1X_1^n + Z_b^n)$$

(40)

which follows from the independence of $X_1^n$ and $X_2^n$. Next, we bound $I(X_1^n; Y_b^n)$ below as

$$I(X_1^n; Y_b^n) = I\left(X_1^n; \frac{b_1}{b_2}X_1^n + X_2^n + \frac{1}{b_2}Z_b^n\right) \geq I\left(X_1^n; \frac{b_1}{b_2}X_1^n + X_2^n + Z_b^n\right)$$

(41)

$$\geq I\left(X_1^n; \frac{b_1}{b_2}X_1^n + X_2^n + Z_b^n\right) - I\left(X_2^n; X_1^n + Z_b^n\right)$$

(42)

$$\geq I\left(X_1^n; a_1X_1^n + Z_b^n\right) - I\left(X_2^n; X_1^n + Z_b^n\right).$$

(43)

Inequality (42) holds due to $|b_2|^2 \geq 1$ and $Z_b(t) \sim \mathcal{C}(0, 1)$, while (43) is obtained from the chain rule as follows

$$I\left(X_1^n; \frac{b_1}{b_2}X_1^n + X_2^n + Z_b^n\right) = I\left(X_1^n; \frac{b_1}{b_2}X_1^n + X_2^n + Z_b^n\right) + I\left(X_2^n; X_1^n + Z_b^n\right)$$

(44)

$$= I\left(X_1^n; \frac{b_1}{b_2}X_1^n + X_2^n + Z_b^n\right) + I\left(X_2^n; \frac{b_1}{b_2}X_1^n + X_2^n + Z_b^n\right)$$

(45)

$$\geq I\left(X_1^n; \frac{b_1}{b_2}X_1^n + Z_b^n\right).$$

(46)

On the other hand, inequality (44) holds since $|b_1|^2/|b_2|^2 \geq |a_1|^2$.

From (40) and (44), we obtain

$$I(X_1^n; Y_a^n) - I(X_1^n; Y_b^n) \leq I(X_2^n; X_1^n + Z_b^n) \leq n$$

(47)

where (a) in (48) is obtained by taking $X_2^n \sim \mathcal{C}(0, I_n)$. Equipped with (48), we obtain (17) as follows

$$I(X_1^n; Y_a^n|W) - I(X_1^n; Y_b^n|W) = \sum_{w} p(w) \left[ I(X_1^n; Y_a^n|W = w) - I(X_1^n; Y_b^n|W = w) \right]$$

(49)

$$= \sum_{w} p(w) \left[ I(X_1^n; Y_a^n) - I(X_1^n; Y_b^n) \right] \leq n$$

(50)

where $X_2^n \sim X_2^n|\{W = w\} \sim F(x_2^n|w)$. The transition from (49) to (50) is due to the Markov chains $W \to X_1^n \to Y_a^n$ and $W \to X_1^n \to Y_b^n$ (see [13] Ch. 5.6.1). On the other hand, (b) in (50) follows from (48), which completes the proof.

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