

Optimal Power Allocation for OFDMA Systems Under I/Q Imbalance

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Abstract—A direct-conversion architecture can offer highly integrated low-cost hardware solutions to communication transceivers. However, it has been demonstrated that radio frequency impairments, such as amplifier nonlinearities, phase noise, and in-phase/quadrature-phase imbalances (IQI), can lead to a severe degradation in the performance and fairness. To this end, we study the power allocation (PA) problem in an orthogonal frequency-division multiple access system, when the served user equipment (UE) suffers from different levels of IQI. Additionally, we present a novel low-complexity solution with directly calculated PA policies, given the Lagrange multiplier, which mitigates the impact of IQI and achieves fairness in terms of capacity for the served UE, by maximizing the minimum achievable capacity of the UE. The effectiveness of the offered solution is validated through simulation results, which reveal that it can drastically increase the minimum achievable UE's capacity.

Index Terms—Fairness, hardware-constrained communications, in-phase/quadrature (I/Q) imbalance, multicarrier communications, orthogonal frequency-division multiple access (OFDMA), power allocation.

I. INTRODUCTION

THE ever-increasing demand for high-data-rate applications and multimedia services has led to the development of flexible and software-configurable transceivers that are capable of supporting the desired quality of service requirements. In this context, a direct-conversion architecture provides an attractive front-end solution, as it requires neither external intermediate frequency filters nor image rejection filters. Instead, the essential image rejection is achieved through signal processing methods. The direct-conversion architecture is low cost and can be easily integrated on chip, which render it an excellent candidate for modern wireless technologies [1], [2]. However, it is typically sensitive to front-end radio frequency (RF) impairments, which are often inevitable due to component mismatch and manufacturing defects [3]–[7]. An indicative example is the in-phase (I) and quadrature (Q) imbalance (IQI), which corresponds to the amplitude and phase mismatch between the I and Q branches of a transceiver and ultimately leads to imperfect image rejection that incurs considerable performance degradation [8], [9]. Furthermore, in multicarrier systems, IQI creates an

additional image signal from the mirror subcarrier, which leads to a throughput ceil [10].

Various approaches have been proposed so far to eliminate, compensate, and mitigate the effects of IQI using baseband signal processing techniques at the receiver (RX) (see [11]–[15] and references therein). For example, in [11], the authors presented an IQI mitigation method for orthogonal frequency-division multiple access (OFDMA) systems, in which each subcarrier is processed jointly with its counterpart at the image subcarrier. All previously mentioned works deal with IQI at the RX by employing digital signal processing. However, in wireless systems, where low cost, energy efficiency, low complexity, and compactness of the RXs are key design requirements, the extra processes in the RX may be prohibitive. Motivated by this, in the present work, we investigate the power allocation (PA) problem for OFDMA wireless systems, when the served user equipment (UE) suffer from different levels of IQI. To take into consideration the impact of IQI, we propose a novel low-complexity solution with directly calculated PA policies, given the Lagrange multiplier (LM), that maximizes the minimum UE's achievable capacity, with respect to the base station (BS) transmitted power. The proposed PA solution outperforms the conventional one, which does not take into consideration the IQI levels of the served UE, while, at the same time, fairness in terms of capacity of the served UE is achieved. The effectiveness of the offered solution is validated through simulations, which reveal that it can significantly increase the minimum achievable UE's capacity.

Notations: Unless otherwise stated, $(\cdot)^*$ denotes conjugation, whereas $\Re\{x\}$ and $\Im\{x\}$ represent the real and imaginary part of x , respectively. Furthermore, the $\mathbb{E}[\cdot]$ and $|\cdot|$ operators denote statistical expectation and absolute value operations, respectively.

II. SYSTEM AND SIGNAL MODEL

In this section, we revisit the ideal signal model, as well as the realistic IQI signal models in a multicarrier direct-conversion RX scenario in an OFDMA system.

A. Ideal RF Front End

We assume OFDMA transmission, where a transmitted signal at subcarrier k for the UE i , $s_i(k)$, conveyed over a wireless channel $h_i(k)$ with an additive white Gaussian noise (AWGN) $n_i(k)$. The received RF signal is passed through various processing stages, also known as the RF front end of the RX. These stages include filtering, amplification, analog I/Q

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demodulation, downconversion to baseband, and sampling. To this end, the corresponding baseband equivalent received signal can be expressed as $r_{\text{id},i}(k) = h_i(k)s_i(k) + n_i(k)$. Note that $h_i(k)$ is given by $h_i(k) = \frac{g_i(k)}{D_i^n}$, where $g_i(k)$ is a complex Gaussian random variable, n represents the path loss exponent, and $D_i = \frac{d_i}{d_0}$, with d_i and d_0 being the distance between the BS and the i th UE and the reference distance, respectively. Based on this, the instantaneous signal-to-noise ratio (SNR) per symbol at the RX input of the i th UE can be given by $\gamma_{\text{id},i}(k) = \frac{P_s(k)}{N_0} |h_i(k)|^2$, where $P_s(k)$ denotes the power per transmitted symbol at subcarrier k and N_0 is the single-sided AWGN power spectral density.

B. IQ Imbalance Model

The time-domain baseband representation of the IQI impaired signal at the i th UE is given by [1] $g_i^{\text{no}} = K_{1,i}g_i + K_{2,i}g_i^*$, where g_i denotes the baseband IQI-free signal at the i th UE and g_i^* raised due to the involved IQI effects. Furthermore, the IQI coefficients $K_{1,i}$ and $K_{2,i}$ are expressed as $K_{1,i} = \frac{1}{2}(1 + \epsilon_i e^{-j\theta_i})$ and $K_{2,i} = \frac{1}{2}(1 - \epsilon_i e^{j\theta_i})$, where ϵ_i and θ_i account for the RX amplitude and phase mismatch of the i th UE, respectively. It is also noted that the IQI parameters are algebraically linked to each other as $K_{2,i} = 1 - (K_{1,i})^*$. The $K_{1,i}$ and $K_{2,i}$ coefficients are associated with the corresponding image rejection ratio (IRR), which determines the amount of attenuation of the image frequency band and is expressed as $\text{IRR}_i = \frac{|K_{1,i}|^2}{|K_{2,i}|^2}$.

It is recalled here that for practical analog RF front-end electronics, the value of IRR is typically in the range of 20–40 dB [5], [16]–[18]. Furthermore, the second term $K_{2,i}g_i^*$ is due to the associated imbalances, and in multicarrier transmission, it denotes the image aliasing effect, which results into crosstalk between the mirror frequencies in the downconverted signal. This is because, in general, complex conjugate in time domain corresponds to complex conjugate and mirroring in the frequency domain. Therefore, the spectrum of the imbalance signal at the k th subcarrier becomes

$$G_{\text{IQI},i}(k) = K_{1,i}G_i(k) + K_{2,i}G_i^*(-k) \quad (1)$$

where $G_i(k)$ and $G_i(-k)$ denote the spectrum of the IQI free signal at the k and $-k$ subcarriers, respectively. Note that in this letter, we assume frequency-independent IQI; however, the generalization to the frequency-dependent case is straightforward using the methodology in [19].

C. OFDMA Systems Impaired by IQI

In the case of multiuser transmission, we assume that multiple RF subcarriers are downconverted to the baseband by means of wideband direct conversion, where the RF spectrum is translated to the baseband in a single downconversion. Note that the wideband conversion is the most general scenario in multicarrier wireless systems [6]. For notational convenience, we denote the set of subcarriers/UE as $\mathcal{K} = \{-K, \dots, -1, 1, \dots, K\}$. Without loss of generality, it is assumed that the signal carried by the k th subcarrier is intended for the k th UE; a signal carried by the mirror subcarrier, $-k$, is intended for UE $-k$. Moreover, since

the BS is usually a high-complexity device, the RF front end of the TX is considered ideal, while the RX experiences IQI. Hence, by using (1), the baseband equivalent received signal in the k th subcarrier of the i th UE can be represented as

$$r_i(k) = K_{1,i}h_i(k)s_i(k) + K_{2,i}h_i^*(-k)s_{-i}^*(-k) + K_{1,i}n_k(k) + K_{2,i}n_k^*(-k) \quad (2)$$

while the baseband equivalent received signal in the $-k$ th subcarrier at the $-i$ th UE can be expressed as

$$r_{-i}(-k) = K_{1,-i}h_{-i}(-k)s_{-i}(-k) + K_{2,-i}h_{-i}^*(k)s_i^*(k) + K_{1,-i}n_{-i}(-k) + K_{2,-i}n_{-i}^*(k). \quad (3)$$

With the aid of (3), it is shown that IQI is the reason that the received baseband equivalent signal intended for the i th UE, $s_i(k)$, (with $i = k \in \mathcal{K}$) is interfered by the image signal intended for UE $-i$, $s_{-i}^*(-k)$. The instantaneous signal-to-interference plus noise ratio (SINR) per symbol at the input of the RX of the i th UE at subcarrier k can be expressed as

$$\gamma_i(k) = \frac{|h_i(k)|^2 P_s(k)}{\frac{|h_i(-k)|^2}{\text{IRR}_k} P_s(-k) + \left(1 + \frac{1}{\text{IRR}_i}\right) N_0}. \quad (4)$$

Similarly, the instantaneous SINR per symbol at the input of the RX of the $-i$ th UE can be obtained by interchanging i with $-i$ and k with $-k$ and vice versa in (4). Consequently, the achievable rates at UE i , with $i = k \in \mathcal{K}$, can be obtained as

$$R_i(k) = \log_2(1 + \gamma_i(k)). \quad (5)$$

III. PROBLEM FORMULATION AND PROPOSED PA SCHEME

In this section, we first define the PA optimization problem, and then, we present a novel solution. We consider that the optimization is performed by the BS, which has full channel state information¹ as well as the served UE IRR values of the served devices, which are reported to the BS by the UE through a feedback channel, when they enter the wireless system. As we are interested in increasing the achievable capacity of each UE, we aim to maximize the minimum capacity with respect to the transmitted power. The corresponding optimization problem can be written as

$$\begin{aligned} \max_{\mathbf{P}} \quad & \min_{k \in \mathcal{K}} R_k \\ \text{s.t.} \quad & \text{C1} : \sum_{k \in \mathcal{K}} P_s(k) \leq P_{\max} \end{aligned} \quad (6)$$

where $\mathbf{P} = [P_s(-K), \dots, P_s(-1), P_s(1), \dots, P_s(K)]$ and P_{\max} stands for the maximum allowable transmitted power. The optimization problem in (6) is identical to the problem of minimum SINR maximization, and thus, it can be rewritten as

$$\begin{aligned} \max_{\mathbf{P}} \quad & \min_{k \in \mathcal{K}} \gamma_k \\ \text{s.t.} \quad & \text{C1} : \sum_{k \in \mathcal{K}} P_s(k) \leq P_{\max}. \end{aligned} \quad (7)$$

¹Note that, in practice, this can be achieved by estimating the channels at the BS, if the channel reciprocity property is valid, i.e., uplink and downlink occurs within a coherence block [20].

The objective function in (7) is not a purely analytical expression. However, by using the epigraph representation of the optimization problem in (6), it can be equivalently expressed as

$$\begin{aligned} & \mathbf{max}_{\mathbf{P}} \quad \mathcal{R} \\ & \mathbf{s.t.} \quad \mathbf{C}_1 : \sum_{k \in \mathcal{K}} P_s(k) \leq P_{\max} \\ & \quad \quad \mathbf{C}_2 : \gamma_k \geq \mathcal{R} \quad \forall k \in \mathcal{K}. \end{aligned} \quad (8)$$

In the above, \mathbf{C}_2 represents the hypograph of the original optimization problem in (6), with \mathcal{R} being an extra auxiliary variables.

Notice that the optimization problem in (8) is nonconvex. However, it can be easily transformed into a convex one by replacing $P_s(k)$ with $\exp(x(k))$ and \mathcal{R} with $\exp(y)$ and by following similar steps as in [21]. After some mathematical manipulations, this problem can be finally expressed as

$$\begin{aligned} & \mathbf{max}_{\mathbf{P}} \quad y \\ & \mathbf{s.t.} \quad \mathbf{C}_1 : \sum_{k \in \mathcal{K}} \exp(x(k)) \leq P_{\max} \\ & \quad \quad \mathbf{C}_2 : \ln \left(\frac{|h_k(-k)|^2}{\text{IRR}_k} \exp(x(-k) - x(k)) \right. \\ & \quad \quad \left. + \left(1 + \frac{1}{\text{IRR}_k} \right) N_0 \exp(-x(k)) \right) \\ & \quad \quad + y - \ln(|h_k(k)|^2) \leq 0 \quad \forall k \in \mathcal{K}. \end{aligned} \quad (9)$$

Apparently, the constraint \mathbf{C}_1 is convex as a summation of convex functions, while \mathbf{C}_2 is also convex, because its Hessian matrix has nonnegative eigenvalues, given in (10):

$$\begin{aligned} \phi_1 &= \frac{|h_k(-k)|^2 (1 + \text{IRR}_k) N_0 \exp(x(-k))}{\left((1 + \text{IRR}_k) N_0 + |h_k(-k)|^2 \exp(x(-k)) \right)^2} \geq 0, \\ \phi_2 &= 0, \quad \phi_3 = 0. \end{aligned} \quad (10)$$

Consequently, the problem in (9) can be solved by using convex optimization techniques. Next, we solve it by using the dual decomposition method [22]. For this reason, the Lagrangian of (9) is needed, which can be obtained as [23]

$$\begin{aligned} L &= y - \lambda \left(\sum_{k \in \mathcal{K}} \exp(x(k)) - P_{\max} \right) \\ &\quad - \sum_{k \in \mathcal{K}} \mu_k \left(\ln \left(\frac{|h_k(-k)|^2}{\text{IRR}_k} \exp(x(-k) - x(k)) \right. \right. \\ &\quad \left. \left. + \left(1 + \frac{1}{\text{IRR}_k} \right) N_0 \exp(-x(k)) \right) \right. \\ &\quad \left. + y - \ln(|h_k(k)|^2) \right) \end{aligned} \quad (11)$$

where $\lambda \geq 0$ and $\mu_k \geq 0$ are the LMs.

For fixed LMs, the optimal solutions for $x(k)$ and $x(-k)$, or equivalently for $P_s(k)$ and $P_s(-k)$, are, respectively, obtained

by using the Karush–Kuhn–Tucker conditions as

$$\begin{aligned} \tilde{P}_s(k) &= -\frac{\xi_k}{2} + \frac{\mu_{-k} - \mu_k}{2\lambda} \\ &\quad + \sqrt{\frac{\xi_k^2}{4} + \frac{(\mu_{-k} - \mu_k)^2}{4\lambda^2} + \frac{\xi_k}{2\lambda} (\mu_{-k} + \mu_k)} \end{aligned} \quad (12)$$

where the coefficients $\xi_k = (\text{IRR}_k + 1) \frac{N_0}{|h_{-k}(k)|^2}$.

For given LMs, (12) is low-complexity directly calculated PA optimization solution, given the LM, that can be calculated in parallel for each UE. Interestingly, it takes into consideration the RX's nonideal characteristics and guarantees fairness in terms of UE's achievable capacity. Additionally, we point out that, according to (12), the power allocated to the UE k is dependent from the RF characteristics and the channels of all the $2K$ UE that are served via the LMs, λ and μ_k . Note that for $K_{1,k} = 1$ and $K_{2,k} = 0$ ($k \in \{-K, \dots, -1, 1, \dots, K\}$), the proposed optimization solution is simplified to the PA for an ideal RF front-end scheme. This PA is used by the BS that is unaware of the UE's RF imperfections, and to what follows, we refer to as "classical PA" [22]. The constants λ and μ_k can be easily estimated, in polynomial time, by an iterative algorithm such as subgradient method, which is out of the scope of the current letter. Interested readers are referred to [23] and [24] for further information.

Proposition 1: The minimum achievable capacity is maximized when the inequality constraint in (6) is satisfied with equality.

Proof: This can be straightforwardly proven by following the same steps as [25, Corollary 1]. ■

Proposition 2: At the optimal, the k and $-k$ UE will have the same achievable capacity.

Proof: This can be straightforwardly proven by following the same steps as [25, Corollary 2]. ■

IV. NUMERICAL RESULTS AND DISCUSSION

In this section, we demonstrate the efficiency of the proposed PA scheme by presenting simulation results. In particular, we consider that a BS serves $2K$ UE. Each UE suffers from different levels of IQI. Furthermore, it is important to note that, unless otherwise is stated, in the following figures, we consider that $n = 3$, $\theta_k = 3^\circ$, $\epsilon_k < 1$ and $D_k = 1$ for $k \in \mathcal{K}$. Finally, without loss of generality, we assume that $P_{\max} = 1$.

Fig. 1 illustrates the detrimental effects of IQI on the achievable capacity of each UE and the efficiency of the proposed PA scheme. We observe that for low $\frac{P_{\max}}{N_0}$ values, IQI does not affect the UE's capacity performance. However, as $\frac{P_{\max}}{N_0}$ increases, the impact of IQI has adverse effects on its UE's achievable capacity. Furthermore, it is evident that the proposed PA scheme can mitigate the performance degradation, due to IQI and positively contribute to the increase of the UE's achievable capacity. For example, for $\frac{P_{\max}}{N_0} = 35$ dB, $\text{IRR}_1 = \text{IRR}_{-1} = 20$ dB, the use of the proposed PA scheme increases the average achievable rate about 22.3%. This indicates the importance of taking into consideration the effects of the UE's IQI, when designing a PA scheme.

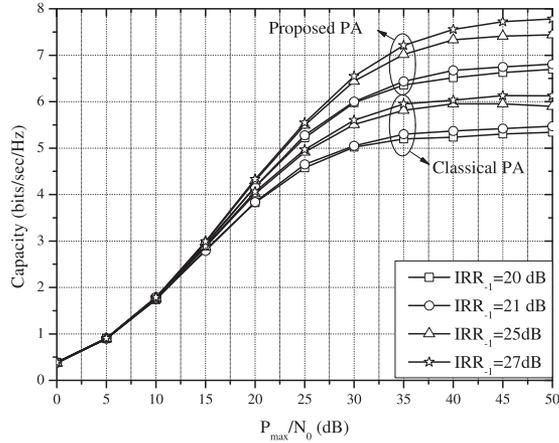


Fig. 1. Capacity as a function of $\frac{P_{\max}}{N_0}$ for different levels of IRR_{-1} , when $IRR_1 = 20$ dB and $K = 1$.

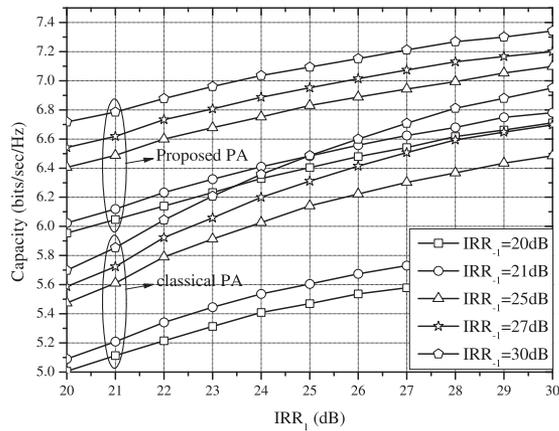


Fig. 2. Capacity as a function of IRR_1 for different levels of IRR_{-1} , $\frac{P_{\max}}{N_0} = 30$ dB and $K = 1$.

In Fig. 2, the UE's capacity as a function of the IRR_1 , for different values of IRR_2 and $\frac{P_{\max}}{N_0} = 30$ dB for both classical and the proposed PA schemes, is plotted. From this figure, it is evident that the proposed PA scheme outperforms the classical one for all the IRR values. Moreover, it is observed that, for a given IRR_{-1} , as IRR_1 increases, the signal leakage of the mirror subcarrier decreases; hence, the performance of the proposed PA scheme tends to these of the conventional scheme. For example, for $IRR_{-1} = 30$ dB, and $IRR_1 = 20$ dB, the use of the proposed PA scheme increases the average achievable rate about 18%, whereas for $IRR_1 = 30$ dB and the same IRR_1 , the increase of the average achievable rate is about 5.6%.

In Fig. 3, the average achievable capacity of each UE of the proposed PA scheme as a function of $\frac{P_{\max}}{N_0}$ for different values of K is depicted, when $IRR_k = 20$ dB, with $k \in \mathcal{K}$. Again, it is observed that the proposed PA scheme outperforms the classical PA scheme for any value of K and in all the $\frac{P_{\max}}{N_0}$ regime. Furthermore, from this figure, it is evident that as K increases, the effects of IQI become more detrimental. For instance, for $K = 2$, in the high $\frac{P_{\max}}{N_0}$ regime, each UE capacity is limited to 4.21 bits/s/Hz, while for $K = 3$, it is constrained

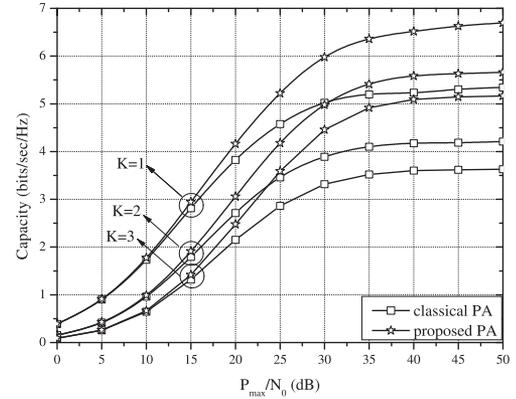


Fig. 3. Capacity as a function of $\frac{P_{\max}}{N_0}$ for $IRR_1 = IRR_{-1} = 20$ dB and different values of K .

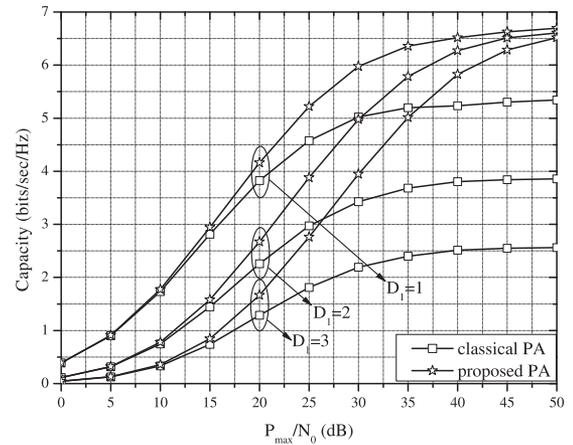


Fig. 4. Capacity as a function of $\frac{P_{\max}}{N_0}$ for $IRR_1 = IRR_{-1} = 20$ dB, $K = 1$, $D_{-1} = 1$, and different values of D_1 .

to 3.63 bits/s/Hz. Additionally, we observe that as K increases, the effectiveness of the proposed PA scheme increases. For example, for $\frac{P_{\max}}{N_0} = 40$ dB, the use of the proposed PA scheme results to 24.47%, 33.79%, and 41.36% increase of the average UE capacity for $K = 1$, $K = 2$, and $K = 3$, respectively.

In Fig. 4, the average achievable capacity of each UE as a function of $\frac{P_{\max}}{N_0}$ for different values of D_1 , when $K = 1$, $D_{-1} = 1$, and $IRR_1 = IRR_{-1} = 20$ dB, is plotted. From this figure, it is evident that the proposed PA scheme outperforms the classical PA scheme for any value of D_1 and in all the transmitted SNR region. Also, for fixed $\frac{P_{\max}}{N_0}$, as D_1 increases, the impact of IQI in the average achievable capacity becomes more severe, when the classical PA is employed. For example, for $\frac{P_{\max}}{N_0} = 40$ dB and $D_1 = 1$, each UE's capacity equals 6.5 bits/s/Hz, whereas, for the same $\frac{P_{\max}}{N_0}$ value and $D_1 = 3$, each UE's capacity is 2.5 bits/s/Hz. Moreover, we observe that as D_1 increases, the effectiveness of the proposed PA scheme increases. For instance, for $\frac{P_{\max}}{N_0} = 40$ dB, the use of the proposed PA scheme results to 24.5%, 64.9%, and 131.7% increase of the average UE's capacity for $D_1 = 1$, $D_1 = 2$, and $D_1 = 3$, respectively. This reveals that the proposed PA scheme provides even larger gain as compared to classic PA, when the UE has different channel qualities.

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