

# Transmit Power Allocation for a Downlink Two-User Interference Channel

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**Abstract**—We develop the optimal transmit power allocation scheme that maximizes the total throughput for a downlink two-user interference channel. The derived optimal scheme allocates the total power to one user in better channel state, as in the greedy scheme, when the degree of interference between users exceeds a certain threshold. When it is less than the threshold, on the contrary, the transmit power is divided into two users, as in the water-filling scheme. Numerical results are presented to verify the optimality of the derived scheme and to show throughput gains over the greedy and water-filling schemes.

**Index Terms**—Greedy power allocation, interference channel, transmit power allocation, water-filling.

## I. INTRODUCTION

TRANSMIT power allocation combined with rate adaptation is an effective means for increasing the throughput of wireless communication systems [1], [2]. The water-filling power allocation scheme has been found to be optimal in the sense of maximizing the throughput, when different data signals are transmitted through orthogonal channels [1]. On the other hand, when different data signals can fully interfere with one another, it has been found to be optimal to assign the total power to only one data signal associated with the best channel condition, which is called greedy power allocation [2]. In the case of the fully interfering channel, transmit power allocation problem has also been addressed for an uncoordinated system, where each user has an individual power constraint [3]. However, optimal power allocation problem has not been addressed for the case where multiple data signals can cause partial interference to one another as in code-division multiple-access (CDMA) systems [4]. Such a general case can be modeled as a form of interference channel [5].

In this letter, we develop the optimal transmit power allocation scheme that maximizes the total throughput for a two-user interference channel. We introduce a parameter called *the portion of interference* to stand for the interference channel. The derived optimal scheme is shown to change according to the portion of interference. Whenever the portion

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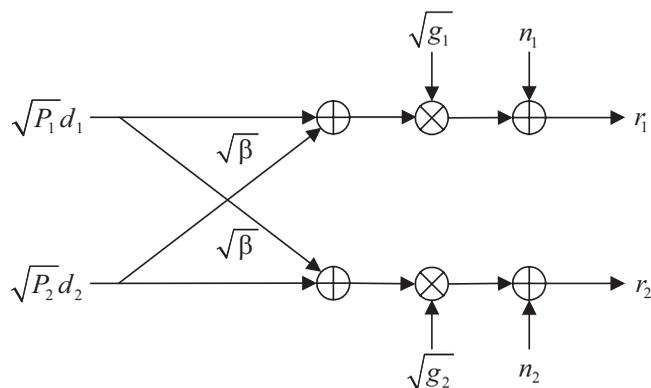


Fig. 1. Downlink two-user interference channel model.

of interference exceeds a certain threshold, the greedy scheme is found to be optimal. When it is less than the threshold, on the contrary, the transmit power is generally divided into two users, as in the water-filling scheme.

## II. INTERFERENCE CHANNEL MODEL AND ACHIEVABLE THROUGHPUT

The interference channel considered in this paper is constructed from a downlink transmission. The base station transmits different data to two different receiving users at the same time. The data  $d_k$  destined for the user  $k$  is transmitted with power  $P_k$ ,  $k = 1, 2$ . Assuming that the two data signals partially interfere with each other in a symmetric form, as illustrated in Fig. 1, we can express the received signal at each user as

$$r_1 = \sqrt{g_1 P_1} d_1 + \sqrt{g_1 P_2 \beta} d_2 + n_1 \quad (1)$$

$$r_2 = \sqrt{g_2 P_2} d_2 + \sqrt{g_2 P_1 \beta} d_1 + n_2 \quad (2)$$

where  $g_k$  denotes the channel gain between the base station and the user  $k$ , and  $n_1$  and  $n_2$  represent independent zero-mean additive white Gaussian noise with variance of  $\sigma^2$ . In Fig. 1,  $\beta$  ( $0 \leq \beta \leq 1$ ) represents *the portion of interference*, which quantifies the degree of interference that each data signal causes to the unintended user. The transmit powers assigned to the two users are assumed to be constrained as

$$P_1 + P_2 = P. \quad (3)$$

Assuming that quadrature amplitude modulation (QAM) is employed, the achievable throughput of the user  $k$  normalized by the transmission bandwidth can be expressed as [2]

$$R_k = \log_2(1 + \gamma_k / \Gamma) \quad (4)$$

where  $\Gamma \equiv -\ln(5P_E)/1.5$  is the signal-to-interference-plus-noise ratio (SINR) gap, and  $P_E$  denotes the target bit error rate (BER) required at each user. In (4), the received SINR  $\gamma_k$  at each user can be obtained from (1) and (2) as

$$\gamma_1 = \frac{g_1 P_1}{g_1 P_2 \beta + \sigma^2}, \quad \gamma_2 = \frac{g_2 P_2}{g_2 P_1 \beta + \sigma^2}. \quad (5)$$

From (4) and (5), the total achievable throughput normalized by the bandwidth is expressed as

$$\begin{aligned} R &= R_1 + R_2 \\ &= \log_2 \left( \left( 1 + \frac{g_1 P_1}{(g_1 P_2 \beta + \sigma^2) \Gamma} \right) \left( 1 + \frac{g_2 P_2}{(g_2 P_1 \beta + \sigma^2) \Gamma} \right) \right). \end{aligned} \quad (6)$$

### III. OPTIMAL TRANSMIT POWER ALLOCATION

The optimal transmit power allocation  $(P_1^*, P_2^*)$  should maximize  $R$  in (6) under the constraint (3). Since the logarithm is monotonically increasing, the objective function that incorporates the power constraint can be defined as

$$\begin{aligned} J(P_1) &= \left( 1 + \frac{g_1 P_1}{(g_1 (P - P_1) \beta + \sigma^2) \Gamma} \right) \left( 1 + \frac{g_2 (P - P_1)}{(g_2 P_1 \beta + \sigma^2) \Gamma} \right), \\ &0 \leq P_1 \leq P. \end{aligned} \quad (7)$$

Unfortunately,  $J(P_1)$  is not guaranteed to be concave with respect to  $P_1$ . Thus, in order to find the optimal power  $(P_1^*, P_2^* = P - P_1^*)$ , we should search over all possible boundary points ( $P_1 = 0$  and  $P_1 = P$ ) and extreme points ( $P_1$ 's corresponding to  $\partial J / \partial P_1 = 0$ ). First,  $J(P_1)$  at the boundary points are computed as

$$J(P_1 = 0) = 1 + \frac{g_2 P}{\sigma^2 \Gamma}, \quad J(P_1 = P) = 1 + \frac{g_1 P}{\sigma^2 \Gamma}. \quad (8)$$

Next, it can easily be shown that

$$\frac{\partial J}{\partial P_1} = \frac{AP_1^2 + 2BP_1 + C}{D} \quad (9)$$

where

$$A = g_1 g_2 (g_1 - g_2) \beta \sigma^2 (1 - \beta \Gamma), \quad (10a)$$

$$B = g_1 g_2 (g_1 P \beta + \sigma^2) (g_2 P \beta^2 \Gamma + (2\beta \Gamma - 1) \sigma^2), \quad (10b)$$

$$\begin{aligned} C &= -(g_1 P \beta + \sigma^2) (g_2 \Gamma \sigma^2 (g_2 P \beta + \sigma^2) \\ &+ g_1 (g_2^2 P^2 \beta^2 \Gamma + g_2 P (\beta \Gamma - 1) \sigma^2 - \Gamma \sigma^2)), \end{aligned} \quad (10c)$$

$$D = ((g_1 (P - P_1) \beta + \sigma^2) (g_2 P_1 \beta + \sigma^2) \Gamma)^2. \quad (10d)$$

Since  $D$  is always positive,  $\partial J / \partial P_1 = 0$  is equivalent to  $AP_1^2 + 2BP_1 + C = 0$ , which leads to

$$P_1 = \begin{cases} (-B \pm \sqrt{B^2 - AC}) / A, & A \neq 0, \\ -C / 2B, & A = 0. \end{cases} \quad (11)$$

From (11), real-valued extreme points  $P_1$  only within  $[0, P]$  can be identified, and  $J(P_1)$  at those points can be calculated using (7). Then, the optimal transmit power  $P_1^*$  corresponds to the point where  $J(P_1)$  is the largest among the two boundary points and those extreme points. From (3),  $P_2^*$  is given as  $P_2^* = P - P_1^*$ . It can be shown that the optimal power allocation becomes identical to the water-filling scheme when  $\beta = 0$ , and to the greedy scheme when  $\beta = 1$ , respectively, which are given as

$$P_1^* = \max \left( 0, \frac{P}{2} + \frac{(g_1 - g_2) \sigma^2 \Gamma}{2g_1 g_2} \right), \quad \text{if } \beta = 0, \quad (12)$$

$$P_1^* = P \quad (\text{if } g_1 \geq g_2) \text{ or } 0 \quad (\text{if } g_1 < g_2), \quad \text{if } \beta = 1. \quad (13)$$

To look into the behavior of the optimal power allocation for different  $\beta$ , we plot how the throughput of the optimal power allocation varies with  $\beta$  in Fig. 2, where the channel gains are assumed to be fixed to  $g_1 = 2.5$  and  $g_2 = 1.5$ . In Fig. 2, the signal-to-noise ratio (SNR), defined to be  $P/\sigma^2$ , is assumed to be 10 dB, and the target BER  $P_E$  is set to  $10^{-3}$ . It is interesting to observe that the throughput does not change with  $\beta$  if  $\beta \geq \beta_{\text{TH}}$ , where the threshold  $\beta_{\text{TH}} = 0.0611$  in Fig. 2. This implies that the greedy scheme is optimal whenever  $\beta \geq \beta_{\text{TH}}$ , since it is optimal at  $\beta = 1$ . The threshold  $\beta_{\text{TH}}$  can be derived by noting that the greedy scheme provides larger throughput than any other power allocation when  $\beta \geq \beta_{\text{TH}}$ . Without loss of generality, we assume  $g_1 \geq g_2$ , and then from (6) and (13), the normalized total throughput  $R_{\text{GR}}$  for the greedy scheme is given as

$$R_{\text{GR}} = \log_2 \left( 1 + \frac{g_1 P}{\sigma^2 \Gamma} \right). \quad (14)$$

From (3), (6), and (14), the difference  $\Delta(P_1) \equiv R_{\text{GR}} - R$  can be expressed as a function of  $P_1$  as

$$\begin{aligned} \Delta(P_1) &= \\ &\log_2 \left( \frac{(\Gamma + g_1 P / \sigma^2) ((P - P_1) \beta + \sigma^2 / g_1) (P_1 \beta + \sigma^2 / g_2) \Gamma}{(P \beta \Gamma - P_1 (\beta \Gamma - 1) + \Gamma \sigma^2 / g_1) (P + P_1 (\beta \Gamma - 1) + \Gamma \sigma^2 / g_2)} \right). \end{aligned} \quad (15)$$

The range of  $\beta$  that yields  $\Delta(P_1) \geq 0$  for any value of  $P_1 \in [0, P]$  can be calculated from (15), and the minimum  $\beta$  in the range corresponds to  $\beta_{\text{TH}}$ . It is derived in Appendix I as

$$\beta_{\text{TH}} = \max(0, \min(\beta_1, \beta_2)) \quad (16)$$

where  $\beta_1$  and  $\beta_2$  are defined in (20) and (21), respectively. It can be verified that the  $\beta_{\text{TH}}$  calculated from (16) under the conditions of Fig. 2 is the same as the one observed in Fig. 2. In the case of  $g_1 < g_2$ ,  $\beta_{\text{TH}}$  is also given as (16) with the exchange of  $g_1$  and  $g_2$ . In summary, the optimal transmit power allocation scheme can be written as

$$\begin{aligned} P_1^* &= \begin{cases} \operatorname{argmax}_{p=0, P, (-B \pm \sqrt{B^2 - AC})/A} J(p), & \beta < \beta_{\text{TH}}, A \neq 0 \\ \operatorname{argmax}_{p=0, P, -C/2B} J(p), & \beta < \beta_{\text{TH}}, A = 0 \\ P, & \beta \geq \beta_{\text{TH}}, g_1 \geq g_2 \\ 0, & \beta \geq \beta_{\text{TH}}, g_1 < g_2 \end{cases} \\ P_2^* &= P - P_1^*. \end{aligned} \quad (17)$$

It should be noted that the transmit power allocation in (17) maximizes the throughput sum of the two users, not considering the individual throughput of each user. If certain fairness between the two users is desired, an adjustable weighting factor  $w$  can be introduced in (6) as in [3], so that  $R_1 + wR_2$  rather than  $R_1 + R_2$  is maximized.

### IV. NUMERICAL RESULTS

Fig. 3 compares the throughput of the optimal power allocation scheme in (17) with those of the water-filling and greedy schemes in (12) and (13) in a Rayleigh fading channel, when SNR = 10 dB and 20 dB. The throughput of each scheme is averaged over 10,000 independent realizations of  $g_1$  and  $g_2$ . As expected, the optimal power allocation scheme is shown

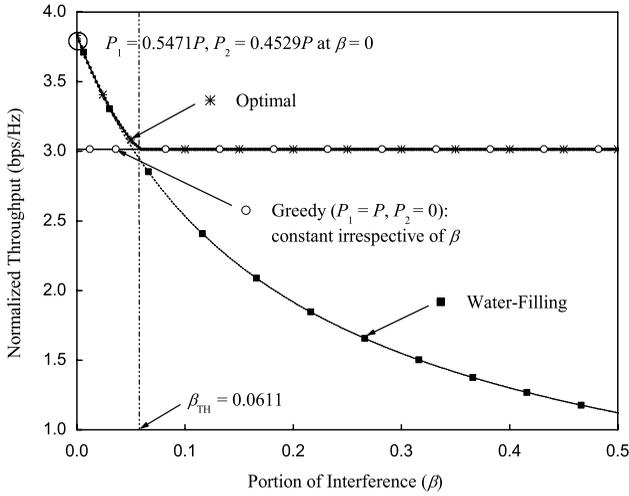


Fig. 2. Normalized throughput versus  $\beta$  for given channel gains ( $g_1 = 2.5$ ,  $g_2 = 1.5$ ).

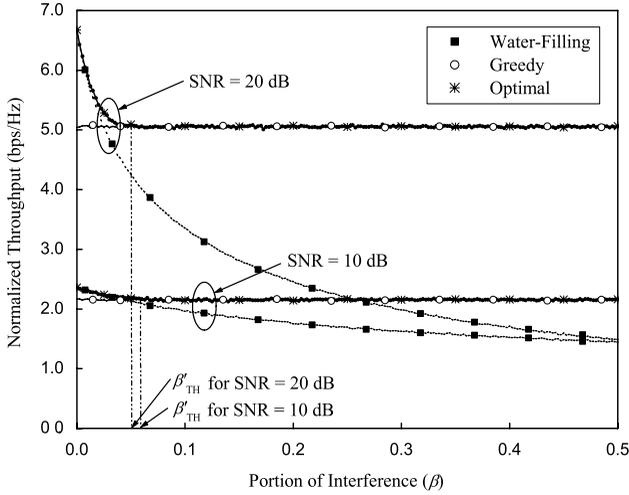


Fig. 3. Average normalized throughput versus  $\beta$  in a Rayleigh fading channel.

to always achieve the maximum throughput. The throughput of the optimal scheme is shown to be indistinguishable as that of the greedy scheme when  $\beta$  is greater than a certain value, denoted as  $\beta'_{TH}$  in Fig. 3. This implies that  $\beta_{TH}$  is less than  $\beta'_{TH}$  in most cases, although  $\beta_{TH}$  is different for different channel realizations. The throughput of the water-filling scheme is almost the same as that of the optimal scheme when  $\beta < \beta'_{TH}$ , but it decreases rapidly as  $\beta$  increases. When  $\beta = 0$  and SNR = 20 dB, the average throughput of the optimal and water-filling schemes is 6.7 bps/Hz, whereas that of the greedy scheme is 5.1 bps/Hz. When  $\beta = 0.5$  and SNR = 20 dB, on the other hand, the optimal and greedy schemes achieve 5.1 bps/Hz, whereas the water-filling scheme achieves only 1.5 bps/Hz.

## V. CONCLUSIONS

We have developed the optimal transmit power allocation scheme that maximizes the total throughput for a downlink two-user interference channel. The greedy scheme is found to be optimal, whenever the portion of interference ( $\beta$ ) is greater than a certain threshold for a given channel condition.

When  $\beta$  is less than the threshold, the optimal scheme is found to behave similarly to the water-filling scheme, which is optimal at  $\beta = 0$ . Numerical results have verified that the optimal scheme can provide significant throughput gain over the greedy scheme for relatively small  $\beta$ , and over the water-filling scheme for large  $\beta$ .

## APPENDIX I

Since the denominator of the argument of the logarithm in the right hand side of (15) is always positive, the condition  $\Delta(P_1) \geq 0$  is equivalent to  $\Omega(P_1) \geq 0$ , where

$$\begin{aligned} \Omega(P_1) \equiv & (\Gamma + g_1 P / \sigma^2) ((P - P_1) \beta + \sigma^2 / g_1) \cdot \\ & (P_1 \beta + \sigma^2 / g_2) \Gamma - (P \beta \Gamma - P_1 (\beta \Gamma - 1) + \Gamma \sigma^2 / g_1) \cdot \\ & (P + P_1 (\beta \Gamma - 1) + \Gamma \sigma^2 / g_2). \end{aligned} \quad (18)$$

Note that  $\Omega(P_1)$  at the boundary points,  $P_1 = 0$  and  $P_1 = P$ , are given as

$$\Omega(0) = (P \beta + \sigma^2 / g_1) (g_1 / g_2 - 1) P \Gamma \geq 0, \quad \Omega(P) = 0. \quad (19)$$

From (19), it can be seen that the range of  $\beta$  that satisfies (18) for any  $P_1 \in [0, P]$  depends on the convexity of  $\Omega(P_1)$ , which can be tested from the sign of the second derivative  $\partial^2 \Omega / \partial P_1^2 = -2(g_1 P \beta^2 \Gamma / \sigma^2 + 2\beta \Gamma - 1)$ .

When  $\partial^2 \Omega / \partial P_1^2 \leq 0$ ,  $\Omega(P_1)$  is concave with respect to  $P_1$ , and thus  $\Omega(P_1) \geq 0$  for any  $P_1$  due to the boundary values in (19). The range of  $\beta$  that satisfies  $\partial^2 \Omega / \partial P_1^2 \leq 0$  is found as

$$\beta \geq \beta_1 \equiv \frac{-1 + \sqrt{1 + g_1 P / \sigma^2 P}}{g_1 P / \sigma^2}. \quad (20)$$

When  $\partial^2 \Omega / \partial P_1^2 \geq 0$ , on the contrary,  $\Omega(P_1)$  is convex with respect to  $P_1$ . For  $\Omega(P_1) \geq 0$  to be satisfied for any  $P_1$ , the minimum point of  $\Omega$  should locate at a  $P_1$  outside  $[0, P]$ . From these conditions, the range of  $\beta$  is found as

$$0 \leq \beta \leq \beta_1,$$

$$\beta \geq \beta_2 \equiv \frac{-(g_1 + g_2) + \sqrt{(g_1 - g_2)^2 + 4g_2^2(1 + g_1 P / \sigma^2 \Gamma)}}{2P g_1 g_2 / \sigma^2}. \quad (21)$$

From (20) and (21), the range of  $\beta$  that satisfies (18) is found as

$$\beta \geq \max(0, \min(\beta_1, \beta_2)) \quad (22)$$

which leads to  $\beta_{TH} = \max(0, \min(\beta_1, \beta_2))$ .

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