Effective Power Allocation Using for Transmit Source In MIMO Relay Communication

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Abstract — The Multiple Input Multiple Output (MIMO) systems using relays are of interest for high-speed radio communication systems. Currently, most of the articles are interested in the model of three nodes with purposes such as increasing the channel capacities (mutual information) or reducing the minimum mean square of error. This paper extends to more than one relay and is concerned with the maximum channel capacity. It is assumed that the channel matrices between source and relay as well as relay and receiver are random matrices; the relay precoders are also assumed to be random and known at the receiver. The article proposes that the Lagrange multiplier finding algorithm using the Newton – Raphson optimization method is more straightforward than the traditional finding algorithm using the first and second derivatives but still gives a higher channel capacity.

Index Terms — Channel capacity, Lagrangian operator, MIMO, relay.

I. INTRODUCTION

Relay-based transmission leads to the advantages of high system reliability, large throughput, and wide coverage in radio communications [1]. In a two-way amplify-and-forward (AF) relay, both terminals know their transmitting signal, and they can ignore self-noise from the received signal before detecting the requested signal if channel state information (CSI) available [1]. When AF is used, the relay amplifies the received signals to overcome the power loss and subsequently forwards them to the destination. In general, AF needs less signal processing at the relay and, thus, is less complex [2], [3].

Combining MIMO with relay-based transmission can better improve system quality due to the advantages of spatial diversity and diversity gain. The MIMO transmission technology is well known to improve spectral efficiency by transmitting multiple data streams across multiple antennas. Multiple antennas could be installed at the source (S), the relay (R), and the destination (D) to form a MIMO relay network [2], [4]. Amongst the various existing relay-based communication strategies (e.g., amplify-and-forward (AF), decode-and-forward, compress-and-forward), AF remains one of the most popular strategies given its simplicity and practicality for enabling multi-input multi-output (MIMO) cooperative communication [4]-[5].

Articles are focused on designing precoders at both source and relay [3], [6], [7]. They also only give a simple model of 3 nodes [4], [5], [7], adding a direct link and no direct link [3].

They are also interested in the simultaneous design of the precoder at the transmitter and the combined relay. In addition, figuring out how to allocate and update power is introduced through [4] using the Lagrange method, which follows the water-filling principle. [6] also added a complex algorithm that guarantees to find the multipliers of the Lagrange method. Here, finding this parameters value is sometimes difficult when the algorithm goes through many loops to find. The article [2] presents a precoder design model for relay only, considering channel state information.

Among the above works, there are few works related to the use of the Lagrange algorithm, such as [2], [4]-[7]. [2] is considered to find the optimal power distribution for the relay. [4] is used to find the optimal power distribution for both transmitter and relay to convert the relay channel MIMO into a multi-channel relay single input single output. However, in [3], [4], to ensure the optimal function property with convexity, it is necessary to keep the power distribution coefficient of the relay or generator constant. The article [7] is more general when giving different optimal cases: only relay precoder or find source precoder when relay precoder is known or vice versa. [6], then study the following case of [7] when finding the optimal source precoder first, then finding the optimal relay precoder.

To summarize, some articles talk about how if joint source/relay precoders make the optimized function convex impossible, it makes it challenging to find the optimization for global [4], [6], [7]. The solution should be separated into the optimizations for the source and relay separately, then the optimized functions for each component are convex [4], [6], [7]. Some articles mentioned Lagrange or KKT conditions for optimal source or relay [4]-[7]. However, with the water-filling algorithm, as shown in [6], it is not very easy to give the optimal eigenvectors and the optimal water-filling algorithm because many conditions need to be satisfied. One method that does not require too many constraints is the Newton - Raphson method, which is described to find the results of the optimal eigenvectors quickly.

II. CHANNEL MODEL

Suppose we have a channel model with a transmitter and a receiver, plus relays in between them. Source, each of the relays, and receiver are all equipped with multiple antennas to transmit and receive data.
Signals received from direct path:

\[ y_{d,1} = \sqrt{\rho_d} H_{d} F_d d + n_{d} \]  
\[ y_{d,2} = \sqrt{\rho_d} H_{d} F_d y_r + n_{d} \]  

(1)

Here \( \rho_d \) is the power distribution factor for the direct link, \( d \) is the data vector to be transmitted, \( F_d \) is the source precoder, \( H_d \) is the channel matrix between the source and the receiver, and \( n_d \) is the noise acting on the transmission from the source to the receiver.

The signal is received through the relay \( k = 1: K \):

\[ y_{r,1} = \sqrt{\rho_d} H_{r} F_r y_r = \sqrt{\rho_d} H_{r} F_r d + n_{r} \]  
\[ y_{r,2} = \sqrt{\rho_d} H_{r} F_r y_r + n_{r} \]  

(2)

Here \( F_r \) is the receive vector matrix of \( K \) relays, denoted by the \( n_{r} \).

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\[ y_r = \left[ y_{r,1}^T, \ldots, y_{r,K}^T \right]^T \] 
\[ y_r = \sqrt{\rho_d} H_{r} F_r y_r \]  
\[ n_{r} = \left[ n_{r,1}, \ldots, n_{r,K} \right]^T \]  

(3)

Similarly, \( H_{r} \) is the matrix of the channel matrices between the receiver and the receiver: \( H_{r,k} = 1: K \):

\[ H_{r} = \left[ H_{r,1}, \ldots, H_{r,K} \right] \]  

(4)

The total signal at the receiver:

\[ y = \left[ y_{d,1}, y_{d,2} \right]^T = H_{\text{eff}} y_r + n_{\text{eff}} \]  
\[ H_{\text{eff}} = \left[ \sqrt{\rho_d} H_{d} F_d, \sqrt{\rho_d \rho_r} H_{r} F_r \right] \]  
\[ n_{\text{eff}} = \left[ n_{d}, \sqrt{\rho_r} n_{r} + n_{d} \right] \]  
\[ H_{r} = \left[ H_{r,1}', \ldots, H_{r,K}' \right]^T \]  
\[ n_{r} = \left[ n_{r,1}', \ldots, n_{r,K}' \right]^T \]  

From there, [6] estimates the channel capacity \( L \) as follows:

\[ \text{P1: max } L = \log \det \left( I_{N_r} + F_r^H M_F F_r \right) \]  
\[ s.t. tr \left( F_r^H F_r \right) = 1, tr \left( F_r^H BF_r \right) = 1 \]  

where \( M_F \) is calculated as follows:

\[ M_F = G_d + G_r - \rho_r \left[ H_{n} \Theta_F, H_{n} + \frac{tr \left( \Theta_F \Sigma_r^H \right)}{1 + K_{rs}} \right] \]  
\[ G_d = \rho_d E \left[ H_{d}^H H_{d} \right] \]  
\[ \Theta_F = \left( I_{N_r} + F_r^H G_d F_r \right)^{-1} \]  

and

\[ B = \frac{\rho_r}{1 - tr \left( F_r^H F_r \right)} \left[ H_{n} \Sigma_r F_r F_r^H H_{n} + \frac{tr \left( F_r^H \Sigma_r F_r \right)}{1 + K_{rs}} \right] \]  

III. CURRENT ALGORITHM FOR THE SOURCE

With [6], we can find the optimum eigenvectors and eigenvalues for the source. Optimum source beam eigenvectors are the transpose conjugate of the eigenvectors of \( M_F = U_a M_a U_a^H \).

The eigenvalues of the corresponding eigenvectors of the relay are \( \Lambda_M = \text{diag} \left\{ \sigma_{M,1}, \ldots, \sigma_{M,N_r} \right\} \) so that \( \left\{ \sigma_{M,1} \right\} \) sorted in descending order.

\[ \text{P2: min } J = -\sum_{i=1}^{N_r} \log \left( 1 + \sigma_{M,i} \chi_i \right) \]  
\[ s.t. \sum_{i=1}^{N_r} x(i) = 1, \sum_{i=1}^{N_r} x(i) \chi_i = 1, \chi_i \geq 0, i = 1, \ldots, N_r \]  

where \( x = \left[ x_1, \ldots, x_{N_r} \right] \) is the vector of the square of the power control factor of \( A_c \) and \( b_i \) is the i th element of \( U_a^H B U_a \).

The problem is divided into the following cases based on the Lagrange operator for (7):
\[ L(v_i) = \sum_{n=0}^{a_i} \log \left( \frac{a_i \sigma_{M,j}}{(a_i - a_j)} b_i \right) \]  
\[ L(v_i) = \sum_{n=0}^{a_i} (a_i - a_j) b_i \]  
\[ L(v_i) = -\sum_{n=0}^{a_i} \left( \frac{a_i - a_j b_i}{(a_i - a_j) v_i + [I_n]b_i} \right)^2 \]  
with \[ |I_0| = N_s \]

\[ a_1 = \sum_{n=0}^{a_i} \sigma_{M,j} > 0, a_2 = \sum_{n=0}^{a_i} b_i \sigma_{M,i} > 0 \]

1. \( v_b < v_{ab} \) then:
   - when \( L(v_{ab}) \leq 0 \), we have \( V_{opt}^v = v_b \). If \( V_{opt}^v = v_b = v_b \)
     then \( I_0 = I_b / j \). Continue to calculate for new \( I_0 \).
   - when \( L(v_{ab}) \geq 0 \), we have \( V_{opt}^v = v_{ab} \). If \( V_{opt}^v = v_{ab} = v_{ab} \)
     then \( I_0 = I_b / j \). Continue to calculate for new \( I_0 \).
   - when \( L(v_{ab}) > 0 \) and \( L(v_{ab}) < 0 \), then \( V_{opt}^v \) will make
     \( L(V_{opt}^v) = 0 \).

2. \( v_b > v_{ab} \), then \( I_0 \) is modified.

3. \( v_b = v_{ab} \), then \( V_{opt}^v = v_b = v_{ab} \).

4. Find \( x_i = a_2 \left[ (a_2 - a_1 b_j) v_{opt}^v + [I_n]b_j \right]^{-1} - \sigma_{M,i}^{-1} \)

Note:
\[
V_{opt}^v = \begin{cases} 
(\frac{I_0}{b_j}) & \text{for } a_2 - a_1 b_j > 0 \\
(\frac{a_2 \sigma_{M,j} - I_0 b_j}{a_2 - a_1 b_j}) & \text{for } a_2 - a_1 b_j < 0 \\
\end{cases}
\]
\[
V_{opt}^{s} = \begin{cases} 
(\frac{a_2 \sigma_{M,j} - I_0 b_j}{a_2 - a_1 b_j}) & \text{for } a_2 - a_1 b_j > 0 \\
(\frac{a_2 \sigma_{M,j} - I_0 b_j}{a_2 - a_1 b_j}) & \text{for } a_2 - a_1 b_j < 0 \\
\end{cases}
\]
\[
v_{opt} = \max_{n=0} v_{opt}^{v} \quad \text{and} \quad v_{opt} = \min_{n=0} v_{opt}^{s}
\]

IV. PROPOSED METHOD

According to [8], page 66 mentions that the Newton-Raphson method is considered a Newton quasi method, has the advantage of fast convergence, and is suitable for fast fading environments.

The steps for Newton-Raphson are applied as below:

1. The initial decision value \( v_1(1) = 0 \)
\[
L(v_1(1)) = \sum_{n=0}^{a_1} (a_1 - a_1) b_i \]
\[
L(v_1(1)) = -\sum_{n=0}^{a_1} \left( \frac{a_1 - a_1 b_i}{(a_1 - a_1) v_i + [I_n]b_i} \right)^2
\]

2. Update value:
\[
v_i(i+1) = v_i(i) - \frac{L(v_i(i))}{L'(v_i(i))}
\]

3. Find \( v_{opt}^v = v_i(i) \) so that \( |v_i(i+1)| \leq \varepsilon \) or \( |L(v_i(i+1))| \leq \varepsilon \)

Find
\[
x_i = a_2 \left[ (a_i - a_1 b_j) v_{opt}^v + [I_n]b_j \right]^{-1} - \sigma_{M,i}^{-1}
\]

V. SIMULATION

We assume that the model consists of 2 relays, each with four transmit and four receive antennas. Likewise, the transmitter and receiver have an antenna count of 4. We assume channel matrices with random fading coefficients like \( H_{s,k}, k = 1 : 2 \), as even the two relay precoders \( F_{s,k}, k = 1 : 2 \) are considered arbitrary. Based on (4), we can find the source beam eigenvectors of the matrix by matrix. Specifically, we consider the case of using algorithm [6] to increase the capacity; then we are assuming that the beam eigenvectors (number of 4) \( F_s \) are equal to the conjugate transpose of the correlated matrix \( M_{F_s} \)'s eigenvectors, we have two cases: Transmitted power values \( x_i, i = 1: 4 \) are taken in absolute value (marked "o," solid- Case A), and when negative values of \( x_i, i = 1 : 4 \) return to 0 are the second case (marked "o", dashed- Case B).

Meanwhile, with the proposed Newton-Raphson algorithm applied to find \( x_i, i = 1: 4 \), we have a line marked "x" dashed (Case B) which characterizes values \( x_i, i = 1: 4 \) as 0 when they are negative, the remaining line "x," solid (Case A) is the case of taking the absolute value of \( x_i, i = 1: 4 \).

Similarly, the solid or dashed line, marked "v" is characteristic of the algorithm [6]; however, the generator has random eigenvectors, which are 0 when it is negative (Case B) or absolute value (Case A). Solid or dashed line, marked with "x" is typical in the case of Newton-Raphson's proposed algorithm; however, the generator has random eigenvectors.
(at the moment the receiver only send to the source eigenvalue, corresponding to the gains of expected channels), which is 0 when it is negative (Case B) or its absolute value (Case A).

When we randomize the channel matrices like $H_{rr,k} H_{ak,k} k = 1:2$, and also the two relay precoders for 5 times, we get the diagram as Fig. 2. Here, in all 5 cases, the capacity of the proposed algorithm (Newton-Raphson) is the most significant and equal, marked "x" (in the case of taking the absolute value – Case A and taking the value of 0 when negative power – Case B). In the case of the first, fourth and fifth randomization, the capacity of the proposed algorithm has the most significant value when SNR has a high value. The capacity of other cases, including algorithm [6], marked "o" when the source eigenvectors are known (Case A and Case B) or marked "v" randomly taken at the source (Case A and Case B), proposed algorithm marked "+" the source eigenvectors known randomly (Case A and Case B), is confused to be better among algorithms. This is also because when randomly taking the source eigenvectors, it is possible that in some cases, they coincide with the ideal source eigenvectors, thus giving a high capacity.

Fig. 2. Channel capacities for proposed algorithm and algorithm [6].
VI. CONCLUSION

MIMO systems using relays are receiving much attention due to their ability to increase capacity and reduce transmission errors. However, complexity increases as we need to find the optimal source and relay precoders to get the highest capacity. The paper focuses on the simpler algorithm to find the optimum power allocation at the source precoder, assuming that the optimal source beam eigenvectors are perfectly known. This algorithm is more uncomplicated than the algorithm [6], while in many testing attempts, it gives higher capacity than the algorithm [6].

REFERENCES


Hoai Trung Tran was born in 1976. He got Bachelor degree in University of Transport and Communications (UTC) in 1997 and hold the post of lecturer at the University. He then got a Master degree from Hanoi University of Science and Technology (HUST) in 2000. In the period 2003 to 2008, he had concentrated on researching in the field of Telecommunication engineering and got his PhD at University of Technology, Sydney (UTS) in Australia. He is currently lecturer at the UTC. His main research interests are digital signal processing (DSP), applied information theory, radio propagation, MIMO antenna techniques and advanced wireless receiver design.