

TRAINING DESIGN IN SINGLE RELAY AF COOPERATIVE SYSTEMS WITH CORRELATED CHANNELS

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ABSTRACT

In this paper, training design is studied for a single relay amplify-and-forward cooperative network. The taps of the frequency selective channels are assumed to be correlated and OFDM modulation is used for transmission. Based on the least squares (LS) criterion, conditions for pilot tone positioning and a number of power allocation schemes are described. Analytical closed-form power allocation expressions are provided for both the source and the relay. The theoretical results are fully corroborated by simulations.

Index Terms— Training design, LS criterion, correlated channels, cooperative communications

1. INTRODUCTION

Cooperative communication systems have attracted a lot of attention recently [1], due to their ability to exploit spatial diversity by utilizing relays to assist transmission between a source and a destination. One of the most popular transmission protocols in cooperative systems is the so-called amplify-and-forward (AF) scheme, in which the relay amplifies and forwards the received signal to the destination. In this paper, we are dealing with the problem of training design for channel estimation in a single relay AF system. All channels in the network are assumed frequency selective, while orthogonal frequency division multiplexing (OFDM) modulation is employed for transmission.

The design of the optimal training in relay based systems has gained increasing interest lately. More specifically, in [2, 3, 4], optimal training design conditions were derived, while a single fixed gain amplification factor per relay was applied independently of the optimization procedure. In [5], different amplification factors per pilot tone (of the OFDM symbol) were incorporated in the design problem. However, in the suggested solutions, all these factors are proved to be

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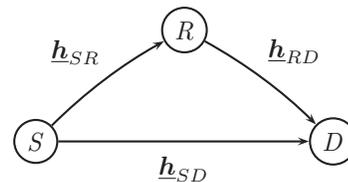


Fig. 1. The cooperative system

equal. All the previous works assume that the channels to be estimated, are uncorrelated. Space channel correlation was assumed in [6], where by using an LMMSE estimator, the optimal training design along with the amplification factors per pilot tone were derived through a convex optimization problem. However, since no closed-form expressions could be extracted, the correlation assumption was relaxed and a suboptimal solution was described. By experiments, it was shown that performance degradation could be considered negligible.

In this paper, the least-squares (LS) criterion is used and the minimum required number of pilot tones are employed. Under a general tap correlation model, conditions are derived for power allocation for the training symbols and their positioning at the source side, as well as power allocation for forwarding the pilot tones at the relay side. It is shown that the well-known conditions for pilot positioning (e.g. [7]) are independent of channel correlation. Thus, the general design problem is cast to a power allocation one, in which allocation of equal power to all pilots is not optimal. Although, the optimal solution to this problem can not be obtained in closed-form, some interesting closed-form suboptimal power allocation solutions are provided, along with their corresponding mean squared error (MSE) performance, which are fully supported by experiments. From these closed-form expressions, the case of uncorrelated channels comes out naturally as a special case.

The outline of the paper is as follows. In Section 2, the system model is presented. Different training design schemes are presented and evaluated in Section 3. In Section 4, a discussion on the theoretical results along with simulations are presented to conclude the paper.

In the following, bold underlined small and capital letters

denote vectors at the time and frequency domain, respectively. Also, bold capital letters are used for matrices. \mathbf{F} denotes the $N \times N$ Fourier matrix whose (p, q) -th element is given by $[\mathbf{F}]_{p,q} = \frac{1}{\sqrt{N}} e^{-j2\pi pq/N}$. \mathbf{A}^T and \mathbf{A}^H denote transposition and conjugate transposition of \mathbf{A} . Also, $\text{diag}\{\underline{\mathbf{a}}\}$ produces a diagonal matrix with $\underline{\mathbf{a}}$ on its main diagonal and $\text{vect}\{\mathbf{A}\}$ produces a vector whose elements are the diagonal elements of \mathbf{A} . \mathbf{I}_N is the identity matrix of size N , $\text{Tr}\{\mathbf{A}\}$ is the trace of \mathbf{A} and $\mathcal{E}\{\cdot\}$ denotes statistical expectation. Finally, $\underline{\mathbf{x}} \sim \text{CN}(\underline{\boldsymbol{\mu}}, \boldsymbol{\Sigma})$ denotes a complex Gaussian random vector with mean $\underline{\boldsymbol{\mu}}$ and covariance matrix $\boldsymbol{\Sigma}$.

2. SYSTEM MODEL

In Fig. 1, the frequency selective channels $\underline{\mathbf{h}}_{SD}$, $\underline{\mathbf{h}}_{SR}$ and $\underline{\mathbf{h}}_{RD}$ are modeled as vectors of lengths L_{SD} , L_{SR} and L_{RD} , respectively. Also, it is assumed that $\underline{\mathbf{h}}_i \sim \text{CN}\{\mathbf{0}, \mathbf{C}_i\}$ and \mathbf{C}_i is the correlation matrix of channel i , where $i \in \{SD, SR, RD\}$. Transmission in the network is performed in blocks of N symbols by utilizing OFDM and “S”, “R” are assumed to be synchronized. To avoid interblock interference, a cyclic prefix (CP) of appropriate length is appended to the transmitted signals ([8]), however, this issue will not be elaborated any further.

A two-phase non-orthogonal transmission protocol [9] is used by “S” in order to transmit information to “D”. At the first phase, “S” sends the OFDM symbol $\underline{\mathbf{x}}_1 = \mathbf{F}^H \underline{\mathbf{X}}_1$ at both “R” and “D” in which the received signals are

$$\underline{\mathbf{y}}_R = \mathbf{H}_{SR} \underline{\mathbf{x}}_1 + \underline{\mathbf{w}}_R, \quad (1)$$

$$\underline{\mathbf{y}}_1 = \mathbf{H}_{SD} \underline{\mathbf{x}}_1 + \underline{\mathbf{w}}_1, \quad (2)$$

respectively. At the second phase, “S” transmits a new symbol $\underline{\mathbf{x}}_2 = \mathbf{F}^H \underline{\mathbf{X}}_2$ and, concurrently, “R” amplifies and forwards to “D” the symbol received at the previous phase. The received signal in this case is given by

$$\underline{\mathbf{y}}_2 = \mathbf{H}_{SD} \underline{\mathbf{x}}_2 + \mathbf{H}_{RD} \mathbf{K} \underline{\mathbf{y}}_R + \underline{\mathbf{w}}_2. \quad (3)$$

In (1)-(3), $\underline{\mathbf{w}}_R \sim \text{CN}(\mathbf{0}, \sigma_R^2 \mathbf{I}_N)$ and $\underline{\mathbf{w}}_1, \underline{\mathbf{w}}_2 \sim \text{CN}(\mathbf{0}, \sigma_D^2 \mathbf{I}_N)$ are the involved noise vectors. The $N \times N$ matrices \mathbf{H}_i are circulant, having as first columns the vectors $[\underline{\mathbf{h}}_i^T \mathbf{0} \dots \mathbf{0}]^T$ and can be expressed as $\mathbf{H}_i = \mathbf{F}^H \boldsymbol{\Lambda}_i \mathbf{F}$, where the diagonal matrix $\boldsymbol{\Lambda}_i$ contains the frequency response (i.e. DFT) of the channel $\underline{\mathbf{h}}_i$. We assume that the relay amplifies each OFDM tone with a different amplification factor by employing a diagonal matrix \mathbf{A} , to be defined shortly. This operation corresponds to pre-multiplication of $\underline{\mathbf{y}}_R$ with a circulant matrix $\mathbf{K} = \mathbf{F}^H \mathbf{A} \mathbf{F}$ in the time domain, as in (3).

In the following, we assume, without loss of generality, that $L_{SD} = L_{SR} + L_{RD} - 1 = L$ and a minimum number of L tones in each $\underline{\mathbf{X}}_i$ are devoted to training. Also, let $\underline{\mathbf{h}} = [\underline{\mathbf{h}}_{SD}^T \underline{\mathbf{h}}_{RD}^T]^T$ be the vector of channels to be estimated by “D”, where $\underline{\mathbf{h}}_R = \underline{\mathbf{h}}_{SR} * \underline{\mathbf{h}}_{RD}$ and $*$ denotes convolution.

After selecting L out of N pilot rows, from (2), (3), indexed by the set $\{i_k\}$, where $k = 1, 2 \dots L$, the frequency domain vectors $\underline{\mathbf{Y}}_i = \mathbf{F} \mathbf{y}_i$, $i = 1, 2$, can be expressed as [4]

$$\underline{\mathbf{Y}} = \mathbf{B} \underline{\mathbf{h}} + \underline{\mathbf{W}} \quad (4)$$

where $\underline{\mathbf{Y}} = [\mathbf{Y}_{1,L}^T \mathbf{Y}_{2,L}^T]^T$, $\underline{\mathbf{W}} = [\mathbf{W}_{1,L}^T \tilde{\mathbf{W}}_{2,L}^T]^T$ and

$$\mathbf{B} = \sqrt{N} \begin{bmatrix} \mathbf{X}_{1,L} & \mathbf{0} \\ \mathbf{X}_{2,L} & \mathbf{A}_L \mathbf{X}_{1,L} \end{bmatrix} \begin{bmatrix} \mathbf{F}_L & \mathbf{0} \\ \mathbf{0} & \mathbf{F}_L \end{bmatrix} = \sqrt{N} \mathbf{X} \mathbf{F}_d. \quad (5)$$

The meaning of subscript L in the above expressions is that L elements or rows ($\{i_k\}$) have been retained from the respective vectors and matrices. It can be verified that $\mathbf{W}_{1,L} \sim \text{CN}\{\mathbf{0}, \sigma_D^2 \mathbf{I}_L\}$ and $\tilde{\mathbf{W}}_{2,L}$ is zero mean with covariance matrix $\mathbf{C}_L = \sigma_R^2 \mathbf{A}_L \mathbf{A}_L^H \boldsymbol{\Theta}_{RD,L} + \sigma_D^2 \mathbf{I}_L$ and $\boldsymbol{\Theta}_{RD,L} = \mathcal{E}\{\boldsymbol{\Lambda}_{RD,L} \boldsymbol{\Lambda}_{RD,L}^H\}$ is a diagonal matrix where the k -th diagonal entry is denoted by $\theta_{k,RD}^2$. It is noted that $\boldsymbol{\Theta}_{RD,L}$ is not the correlation matrix of the OFDM frequency tones and that, in general, has different diagonal elements, i.e. $\theta_{k,RD}^2 = \theta_{j,RD}^2$, $k \neq j$ ([10]). Also, $\mathbf{X}_{i,L} = \text{diag}\{\mathbf{X}_{i,L}\}$ ($i = 1, 2$) and the $L \times L$ matrix \mathbf{F}_L is produced by the corresponding L rows of \mathbf{F} and its first L columns. In the above derivation, the relation $\underline{\mathbf{A}}_{i,L} = \sqrt{N} \mathbf{F}_L \underline{\mathbf{h}}_i$, $i \in \{SD, R\}$, was used ([7]), where $\underline{\mathbf{A}}_{i,L} = \text{vect}\{\boldsymbol{\Lambda}_{i,L}\}$. Finally, the diagonal matrix \mathbf{A}_L contains the L amplification factors corresponding only to the pilot tones. Its k -th element is expressed as

$$\alpha_k = \sqrt{\frac{e_k}{\theta_{k,SR}^2 p_k + \sigma_R^2}}, \quad (6)$$

where $\theta_{k,SR}^2$ is the k -th diagonal element of $\boldsymbol{\Theta}_{SR,L}$, which is defined similarly to $\boldsymbol{\Theta}_{RD,L}$. e_k is the mean energy assigned by the relay to the k -th pilot tone and p_k is the energy assigned by “S” to the k -th pilot symbol residing in position i_k of $\underline{\mathbf{X}}_1$. For future use, we also define with q_k the corresponding energy for the k -th pilot symbol in $\underline{\mathbf{X}}_2$.

It is assumed that the mean total energy available at the relay for forwarding pilot tones is E_R . It can be shown that this is achieved if $\sum_{k=1}^L e_k = E_R$. Also, “S” assigns energies $\sum_{k=1}^L p_k = E_1$ and $\sum_{k=1}^L q_k = E_2$ to the first and second phase, respectively, and $E_1 + E_2 = E_S$ in total.

3. TRAINING DESIGN FOR LS CHANNEL ESTIMATION

The LS estimator of $\underline{\mathbf{h}}$ in (4) and the error covariance matrix $\mathbf{C}_e = \mathcal{E}\{(\hat{\underline{\mathbf{h}}} - \underline{\mathbf{h}})(\hat{\underline{\mathbf{h}}} - \underline{\mathbf{h}})^H\}$ that describes its performance, are given by [11]

$$\hat{\underline{\mathbf{h}}} = \mathbf{B}^{-1} \underline{\mathbf{Y}}, \quad \mathbf{C}_e = \mathbf{B}^{-1} \mathbf{C}_W \mathbf{B}^{-H}, \quad (7)$$

where

$$\mathbf{C}_W = \begin{bmatrix} \sigma_D^2 \mathbf{I}_L & \mathbf{0} \\ \mathbf{0} & \mathbf{C}_L \end{bmatrix}. \quad (8)$$

In this section, the conditions for pilot tone positioning and power allocation will be determined. As it will be shown, the equispaced positioning and the zero power allocation in the second phase (i.e. $E_2 = 0$) are similar to results reported for the uncorrelated case, as in [4]. The design problem, then, will be formulated as a power allocation scheme with respect to p_k 's and e_k 's only.

The optimal training design can be determined by solving the following minimization problem

$$\min_{\underline{p}, \underline{q}, \underline{e}, E_1, E_2, \{i_k\}} \frac{1}{2L} \text{Tr}\{\mathbf{C}_e\} \quad (9)$$

$$\begin{aligned} \text{s.t.} \quad & \sum_k e_k = E_R \\ & \sum_k p_k = E_1 \\ & \sum_k q_k = E_2 \\ & E_1 + E_2 = E_S, \end{aligned}$$

where the $L \times 1$ vectors \underline{p} , \underline{q} and \underline{e} contain the parameters p_k , q_k and e_k , respectively.

The minimization of (9) can be conducted based on the following lower bound expression.

$$\begin{aligned} \text{MSE} &= \frac{1}{2L} \text{Tr}\{\mathbf{B}^{-1} \mathbf{C}_W \mathbf{B}^{-H}\} = \\ &= \frac{1}{2LN} \text{Tr}\{(\mathbf{X}^H \mathbf{C}_W^{-1} \mathbf{X})^{-1} (\mathbf{F}_d \mathbf{F}_d^H)^{-1}\} = \\ &= \frac{1}{2LN} \text{Tr}\{\mathbf{M} \mathbf{Q}\} \geq \\ &\geq \frac{1}{2LN} \sum_{i=1}^{2L} \mu_i \nu_i, \end{aligned} \quad (10)$$

where μ_i and ν_i are the eigenvalues of $\mathbf{M} = (\mathbf{X}^H \mathbf{C}_W^{-1} \mathbf{X})^{-1}$ and $\mathbf{Q} = (\mathbf{F}_d \mathbf{F}_d^H)^{-1}$ in non-decreasing and non-increasing order, respectively. The lower bound in (10) is achievable when matrices \mathbf{M} , \mathbf{Q} are diagonal ([12, p. 249]).

By simple inspection ((4)), \mathbf{M} becomes diagonal when $\mathbf{X}_{2,L} = \mathbf{0}$ or else $\underline{q} = \underline{0}$ and $E_1 = E_S$, $E_2 = 0$. Moreover, $\mathbf{Q} = (\mathbf{F}_d \mathbf{F}_d^H)^{-1}$ is diagonal when pilot tones are equispaced, i.e., $i_k = (k-1)N/L$, $k = 1, 2, \dots, L$ and N/L assumed to be integer ([7]). In this case, \mathbf{Q} is expressed as $\mathbf{Q} = \frac{N}{L} \mathbf{I}_L$.

Based on these results, it turns out that (10) holds with equality, which after some straightforward manipulations is rewritten as

$$f(\underline{p}, \underline{e}) = \frac{1}{2L^2} \left(\sum_{k=1}^L \frac{\gamma_k}{p_k} + \sum_{k=1}^L \frac{\delta_k}{e_k} + \sum_{k=1}^L \frac{\beta}{p_k e_k} \right) \quad (11)$$

where $\gamma_k = \sigma_D^2 + \sigma_R^2 \theta_{k, RD}^2$, $\delta_k = \sigma_D^2 \theta_{k, SR}^2$ and $\beta = \sigma_D^2 \sigma_R^2$.

From (11), the minimization problem (9) is written as

$$\min_{\underline{p}, \underline{e}} f(\underline{p}, \underline{e}), \quad \text{s.t.} \quad \sum_k e_k = E_R, \quad \sum_k p_k = E_S. \quad (12)$$

To the best of our knowledge there is no closed-form analytical solution to this minimization problem. In the following some interesting suboptimal schemes are presented, proceeding from the simpler to the more elaborated one.

3.1. Case I

When the pilot tones are equipowered at both the source and the relay, i.e., $p_k^I = E_S/L$ and $e_k^I = E_R/L$, the MSE in (11) is expressed as $\text{MSE}_{\text{both}} = f(\underline{p}^I, \underline{e}^I)$. In this case, no further information is required by the source and the relay for training power allocation. This scheme will be used as a reference for the subsequent allocation schemes.

3.2. Case II

When only the source uses equipowered training symbols, i.e., $p_k^{II} = E_S/L$, the $f(\underline{p}^{II}, \underline{e})$ is a function of \underline{e} . By substituting \underline{p}^{II} in (12) and using Lagrange multipliers, the optimal power allocation at the relay turns out to be

$$e_k^{II} = \frac{\sqrt{\delta_k + \frac{L\beta}{E_S}}}{\sum_j \sqrt{\delta_j + \frac{L\beta}{E_S}}} E_R = \frac{\sqrt{\theta_{k, SR}^2 + \frac{L\sigma_R^2}{E_S}}}{\sum_j \sqrt{\theta_{j, SR}^2 + \frac{L\sigma_R^2}{E_S}}} E_R \quad (13)$$

and the corresponding performance is given by $\text{MSE}_{\text{source}} = f(\underline{p}^{II}, \underline{e}^{II})$. In this case, the relay requires knowledge of $\theta_{k, SR}^2$'s and σ_R^2 , which are second order statistical terms that are related only to the source-to-relay channel. As we will see shortly, this is an important difference with *Case III* requirements.

3.3. Case III

Similarly, if only the relay uses equipowered tones with $e_k^{III} = E_R/L$, then the optimal values at the source are

$$p_k^{III} = \frac{\sqrt{\gamma_k + \frac{L\beta}{E_R}}}{\sum_j \sqrt{\gamma_j + \frac{L\beta}{E_R}}} E_S \quad (14)$$

and the MSE is given by $\text{MSE}_{\text{relay}} = f(\underline{p}^{III}, \underline{e}^{III})$. In this case, the source requires $\theta_{k, RD}^2$'s, σ_R^2 and σ_D^2 . Although, all these quantities are also second order statistical terms, they are more difficult to be acquired by the source, as they are mostly related to the relay-to-destination channel.

3.4. Case IV

Finally, an approach is described to minimize (12) if neither the source nor the relay use equipowered tones. First, (12) is minimized only with respect to the p_k 's using Lagrange multipliers leading to

$$p_k^{IVa} = \frac{\sqrt{\gamma_k + \frac{\beta}{e_k}}}{\sum_j \sqrt{\gamma_j + \frac{\beta}{e_j}}} E_S. \quad (15)$$

By substituting (15) in (12), the optimization problem is expressed with respect to the e_k 's as follows

$$\min_{\underline{e}: \sum_k e_k = E_R} \frac{1}{2L^2 E_S} \left(\sum_k \sqrt{\gamma_k + \frac{\beta}{e_k}} \right)^2 + \frac{1}{2L^2} \sum_k \frac{\delta_k}{e_k}. \quad (16)$$

To get a closed form solution from (16), we suggest to minimize an upper bound by utilizing Jensen's inequality. According to Jensen's inequality if a function $f(\cdot)$ is convex, then $f(\frac{\sum_i x_i}{L}) \leq \frac{\sum_i f(x_i)}{L}$ with equality if either all x_i 's are equal or $f(\cdot)$ is linear. Applying this inequality to the first summation term in (16) and using Lagrange multipliers in the transformed problem yields

$$e_k^{IVa} = \frac{\sqrt{\delta_k + \frac{L\beta}{E_S}}}{\sum_j \sqrt{\delta_j + \frac{L\beta}{E_S}}} E_R \quad (17)$$

and the performance is $MSE_{nonea} = f(\underline{p}^{IVa}, \underline{e}^{IVa})$.

The same procedure can be followed if we start by minimizing first for e_k 's and then for p_k 's. The final results, in this case, are

$$e_k^{IVb} = \frac{\sqrt{\delta_k + \frac{\beta}{p_k}}}{\sum_j \sqrt{\delta_j + \frac{\beta}{p_j}}} E_R, \quad p_k^{IVb} = \frac{\sqrt{\gamma_k + \frac{L\beta}{E_R}}}{\sum_j \sqrt{\gamma_j + \frac{L\beta}{E_R}}} E_S \quad (18)$$

and the performance is $MSE_{noneb} = f(\underline{p}^{IVb}, \underline{e}^{IVb})$.

4. DISCUSSION

In the previous section, power allocation schemes for the training symbols were presented. The variety of the proposed schemes is due to the *channel taps correlation* assumption made in this work. Such an assumption is reasonable and leads to channels, whose $\theta_{k,SR}^2$ and $\theta_{k,RD}^2$ parameters take, in general, distinct values. It is thus clear that optimal power allocation for training should take into consideration the characteristics of the channels involved, and adapt the design according to each scenario. It should be noted that, all schemes fall into scheme *I* when the channels are uncorrelated (in this case $\theta_{k,SR}^2 = \theta_{SR}^2$, $\theta_{k,RD}^2 = \theta_{RD}^2$ for all k). Moreover, using Jensen's inequality on the MSE results, it can be proven that

$$\begin{aligned} MSE_{source} &\leq MSE_{both}, & MSE_{relay} &\leq MSE_{both}, \\ MSE_{nonea} &\leq MSE_{source}, & MSE_{noneb} &\leq MSE_{relay}. \end{aligned}$$

These inequalities state that scheme *IV* is always better than scheme *II* or *III* which, is superior than scheme *I*. Again, equalities are valid when the channels are uncorrelated.

Finally, we provide some simulations to verify our theoretical analysis. We assume that C_i 's are Toeplitz with first rows $[1, \rho^1, \dots, \rho^{L_i-1}]$, $0 \leq \rho < 1$ and $i \in \{SD, SR, RD\}$. We set $\rho = 0.9$, $L_{SD} = 16$, $L_{SR} = 8$, $L_{RD} = 9$ and $E_S = E_R = 1$. In Fig. 2, the MSE_{both} , MSE_{source} and MSE_{nonea} versus the $SNR = 10 \log_{10} \frac{E_S}{2L\sigma_D^2}$ are plotted. It can be seen that the theoretical results coincide with the experimental ones. Moreover, although not shown here, MSE_{both} is independent of ρ . Finally, the performance of the schemes *II*, *III* and *IV* tends very fast to that of scheme *Case I*, as ρ goes to 0 (a similar observation was also made in [6]).

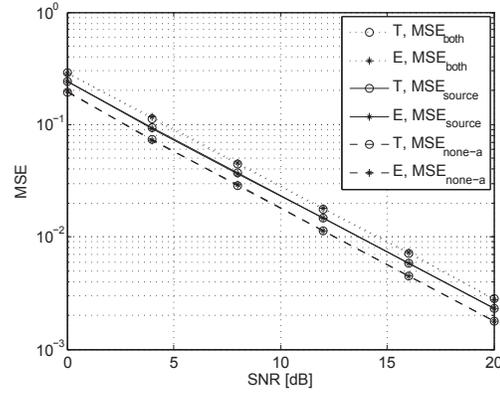


Fig. 2. (T)heoretical and (E)xperimental evaluation of three training power allocation schemes

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