

Achievable Rate Regions for Two-Way Relay Channel Using Nested Lattice Coding

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Abstract—This paper studies a Gaussian two-way relay channel where two communication nodes exchange messages with each other via a relay. It is assumed that all nodes operate in half-duplex mode without any direct link between the communication nodes. A compress-and-forward relaying strategy using nested lattice codes is first proposed. Then, the proposed scheme is improved by performing layered coding: A common layer is decoded by both receivers, and a refinement layer is recovered only by the receiver that has the best channel conditions. The achievable rates of the new scheme are characterized and are shown to be higher than those provided by the decode-and-forward strategy in some regions.

Index Terms—Compress-and-forward, Gaussian channel, lattice codes, physical-layer network coding, side information, two-way relay channel.

I. INTRODUCTION

CONSIDER the Two-Way Relay Channel (TWRC) depicted in Fig. 1. Two wireless terminals T_1 and T_2 , with no direct link between them, exchange individual messages via a relay. Recently, the capacity characterization of this channel has attracted a lot of interest since TWRC is encountered in various wireless communication scenarios, such as ad-hoc networks, range extension for cellular and local networks, or satellite links.

While network level routing is the standard option to solve this problem, it has been shown that network coding (NC) strategies provide better performance by leveraging the side information that is available at each node. In fact, NC [1] offers rate improvements by combining raw bits or packets at network layer. The rate performance of the system can be further improved if NC takes place at the physical layer. In this situation, the linear superposition property of the wireless channel is considered as a “code” and can be exploited appropriately to turn interference into a useful signal [2]. In this context, we consider a physical-layer network coding (PNC) architecture in

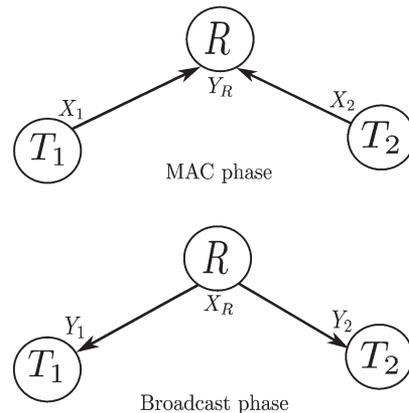


Fig. 1. The two-phase transmission of TWRC: MAC and Broadcast phases.

which the overall communication requires two phases, namely a Multiple Access (MAC) phase in which the terminals simultaneously send their messages to the relay and a Broadcast (BC) phase in which the relay transmits a message that is a function of the signals received in the MAC phase. An outer bound on the capacity region of this model is given in [3], [4].

Several coding strategies have been proposed for PNC by extending classical relaying strategies such as Amplify-and-Forward (AF), Decode-and-Forward (DF), and Compress-and-Forward (CF) to TWRC. AF strategy [5] is a linear relaying protocol where the relay only scales the received signal to meet its power constraints. This simple strategy suffers from noise amplification especially at low signal-to-noise ratios (SNRs). With DF strategy, the relay jointly decodes both messages, and then re-encodes them before broadcasting the resulting codeword. The authors in [5] derived an achievable rate region for TWRC by using DF strategy and superposition coding in the BC phase. This region has been improved in [6] where the authors propose that the relay sends a modulo sum of the decoded messages, thus mimicking the initial example of XOR NC. These DF relaying based schemes require full decoding of the incoming signals and thus suffer from a multiplexing loss due to the MAC phase limitation [3].

The authors in [2], [7] propose PNC schemes based on a partial DF (pDF) where the relay does not decode completely the incoming signals, but relies on the side information available at each terminal to decode a linear function of the transmitted codewords. The key strategy in these schemes is to design the codes at both transmitting terminals in the MAC phase so that the relay can compute a message which is decodable by both nodes during the BC phase. Nested lattice codes, which have the nice property to ensure that any integer-valued linear

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combination of codewords is a codeword, are used in [7], [8] to implement pDF for Gaussian channels. In [9], the lattice based-scheme proposed in [7] has been extended for TWRC with more than one relay. Although the advantage presented by these schemes in using lattice coding, the problem of pDF schemes is to guarantee phase coherence at the relay during the MAC phase [3].

Another strategy has been proposed where the relay compresses its, utilizing Wyner-Ziv binning. This strategy has attracted our particular attention since it offers a good trade-off between processing complexity at the relay and noise amplification compared to DF and AF strategies. This has motivated us to study CF based techniques. CF for TWRC [10] follows the same approach as CF schemes for the relay channel [11]. With CF scheme, the relay does not decode any message, but rather compresses the received signal and sends a new message that includes some useful information about the original messages. This technique does not impose decoding rates at the relay as in DF-based schemes. Performance bounds of CF scheme for TWRC have been investigated in [12]–[14]. It has been shown that for specific channel conditions, namely symmetric channels, CF outperforms the other relaying schemes at high SNR regimes. Random coding tools have been used in the aforementioned references to derive achievable rate regions of CF. Structured codes, on the other hand, have been found to be more advantageous in practical settings thanks to their reduced implementation complexity [15].

CF strategy using lattice coding for three node Gaussian relay channel has been considered in [16], [17]. In [18], we have proposed a CF scheme for TWRC that is based on nested lattice coding. In the MAC phase of this scheme, the communicating nodes simultaneously send their messages and the relay receives a mixture of the transmitted signals. The relay considers this mixture as an analog source which is compressed and transmitted during the BC phase. Taking into account that each terminal has a partial knowledge of this source (namely, its own signal that has been transmitted during the MAC phase, now considered as receiver side information), the BC phase is equivalent to a Wyner-Ziv compression setting with two decoders, each one having its own side information. Each user employs lattice decoding technique to retrieve its data based on the available side information. The proposed scheme can be seen as an extension of lattice quantization introduced in [19] to the TWRC model. In this paper, we first generalize this latter scheme and we apply the results to our transmission problem.

In the simplest situation, when a “single” layer of compression is performed, the relay broadcasts a common compressed message to both terminals. Therefore, it is easily understood that the achievable rates in both directions are constrained by the capacity of the worst channel. In this case, the user experiencing better channel conditions and side information is strongly constrained by this restriction on its transmission rate. To overcome this limitation, we propose an improved scheme where the relay also sends an individual description of its output that serves as an enhancement compression layer to be recovered only by the best receiver. Therefore, the new scheme employs three nested lattices. The common information is encoded using two nested lattices while the refinement

information is encoded with a finer lattice that contains the other two lattices. The channel codewords corresponding to the two layers are superimposed and sent during the BC phase. Through numerical analysis, we show that this layered scheme outperforms AF and CF strategies in all SNR regimes and DF strategy for specific SNR regions.

Layered coding for Wyner-Ziv problem has been addressed in [20] for lossy transmission over broadcast channel with degraded side information. In [14], the authors derive the achievable rate region of layered CF coding for TWRC, based on a random coding approach. The authors in [17], [21] and [22] proposed schemes for TWRC based on doubly nested lattice coding where different power constraints at all nodes are assumed. In these schemes, each of the two end terminals employs a different code (with carefully chosen rate) constructed from the lattice partition chain. The relay decodes a modulo-lattice sum of the transmitted codewords from the received signal. However, in [21] full-duplex nodes are considered and in [17] and [22], the direct link between both terminals is exploited and the transmission is performed in three phases. In these schemes, the relay follows a DF-lattice coding strategy since it decodes a function of the transmitted lattice codewords. On the other hand, in our proposed enhancement scheme, doubly nested lattice coding is only employed at the relay for CF strategy and half-duplex terminals are considered with no direct link between the two end terminals. Furthermore, the relay does not need to know either the other terminals’ codebooks or the precise value of the channel. It merely reconstructs its encoder from the channel module and the variances of the transmitted signals. This strategy ensures a small processing load at the relay. To our knowledge, our work is the first that proposes a doubly nested lattice coding for CF relaying in TWRC.

The remaining of the paper is organized as follows. Section II introduces the system model. Section III derives the achievable rate region when one layer lattice-based coding scheme is used and Section IV derives the achievable rate region with two layer lattice-based coding. Section V illustrates the performance of the proposed schemes through numerical results. Finally, Section VI concludes the paper.

Notations: Random variables (r.v.) are indicated by capital letters and their realizations are denoted by small letters. Vector of r.v. or a sequence of realizations are indicated by bold fonts.

II. SYSTEM MODEL

Consider a Gaussian TWRC in which two source nodes T_1 and T_2 exchange two individual messages m_1 and m_2 , with the help of a relay R as shown in Fig. 1. For this model, we have the following assumptions:

- a.1) The relay and the source nodes operate in half-duplex mode;
- a.2) The two users are assumed to be synchronized, and due to the half duplex mode, there is no direct link between T_1 and T_2 .
- a.3) The communication takes n channel uses that are split into two orthogonal phases: MAC phase and BC phase with lengths $n_1 = \alpha n$ and $n_2 = (1 - \alpha)n$, $\alpha \in [0, 1]$ respectively.

During the MAC phase, node T_1 draws uniformly a message m_1 from the set $\mathcal{M}_1 = \{1, 2, \dots, 2^{nR_{12}}\}$ and sends it to the other terminal T_2 where R_{12} denotes the message rate from node T_1 to node T_2 . Similarly, node T_2 draws uniformly a message m_2 from the set $\mathcal{M}_2 = \{1, 2, \dots, 2^{nR_{21}}\}$ and sends it to the other terminal T_1 where R_{21} denotes the message rate from node T_2 to node T_1 . Let $\mathbf{x}_i(m_i) \in \mathbb{R}^{n_1}$ be the channel codeword of length n_1 sent by node T_i , $i = 1, 2$ and P_i be the corresponding transmit power constraint that verify the following assumptions

a.4) $(1/n_1) \sum_{k=1}^{n_1} |x_{i,k}|^2 \leq P_i$.

The messages are transmitted via a memoryless Gaussian channel and the relay R receives a signal $\mathbf{y}_R \in \mathbb{R}^{n_1}$ given by

$$\mathbf{y}_R = h_1 \mathbf{x}_1 + h_2 \mathbf{x}_2 + \mathbf{z}_R \quad (1)$$

where h_i denotes the channel coefficient between T_i and R , $i = 1, 2$. We assume that:

a.5) The components of the random vector \mathbf{Z}_R are i.i.d Additive White Gaussian Noise (AWGN) at the relay with variance σ_R^2 , i.e., $\sim \mathcal{N}(0, \sigma_R^2)$ and they are independent from the channel inputs \mathbf{X}_i , $i = 1, 2$.

a.6) The channel coefficients follow a block fading model. Channel reciprocity between MAC and BC channels is assumed, i.e., $h_{i \rightarrow R} = h_{R \rightarrow i} = h_i$.

During the BC phase, the relay generates a codeword $\mathbf{x}_R(m_R) \in \mathbb{R}^{n_2}$ of dimension n_2 from the received sequence \mathbf{y}_R . The average power constraint at the relay P_R verifies

a.7) $(1/n_2) \sum_{k=1}^{n_2} |x_{R,k}|^2 \leq P_R$.

The signal \mathbf{x}_R is transmitted through a broadcast memoryless channel and the received signal at node T_i is $\mathbf{y}_i \in \mathbb{R}^{n_2}$, $i = 1, 2$

$$\mathbf{y}_i = h_i \mathbf{x}_R + \mathbf{z}_i, \quad (2)$$

a.8) The components of \mathbf{Z}_i are i.i.d AWGN at node T_i with variance σ_i^2 , $i = 1, 2$ and they are independent from the channel input \mathbf{X}_R .

Perfect CSI is assumed at all nodes. This assumption is further discussed in Remark 3. For the aforementioned TWRC, a rate pair (R_{12}, R_{21}) is said to be achievable if there exists a sequence of encoding and decoding functions such that the decoding error probability approaches zero for n sufficiently large.

For the sake of completeness, we hereafter outline some preliminaries on lattices [15], [23].

Fundamentals on Lattice Coding

A real n_1 -dimensional lattice Λ is a subgroup of the Euclidean space $(\mathbb{R}^{n_1}, +)$. $\forall \lambda_1, \lambda_2 \in \Lambda, \lambda_1 + \lambda_2 \in \Lambda$. We present below some fundamental properties associated with a lattice:

- The nearest neighbor lattice quantizer of Λ is defined as $Q_\Lambda(\mathbf{x}) = \arg \min_{\lambda \in \Lambda} \|\mathbf{x} - \lambda\|$ where $\mathbf{x} \in \mathbb{R}^{n_1}$ and $\|\cdot\|$ is the Euclidean norm.
- The basic Voronoi cell of Λ is the set of points in \mathbb{R}^{n_1} closer to the origin than to any other point of Λ , $\mathcal{V}(\Lambda) = \{\mathbf{x} | Q_\Lambda(\mathbf{x}) = \mathbf{0}\}$.

- The volume of a lattice $V := \text{Vol}(\mathcal{V}(\Lambda))$.
- The mod- Λ operation is defined as $\mathbf{x} \bmod \Lambda = \mathbf{x} - Q_\Lambda(\mathbf{x})$. It satisfies the distributive law: $(\mathbf{x} \bmod \Lambda + \mathbf{y}) \bmod \Lambda = (\mathbf{x} + \mathbf{y}) \bmod \Lambda$.
- The second moment per dimension of Λ is $\sigma^2(\Lambda) := (1/n_1) \cdot (1/V) \int_{\mathcal{V}(\Lambda)} \|\mathbf{x}\|^2 d\mathbf{x}$.
- The dimensionless normalized second moment is defined as $G(\Lambda) := (\sigma^2(\Lambda))/V^{2/n_1}$.
- A sequence of n_1 -dimensional lattices $\Lambda^{(n_1)}$ is said to be good for quantization if $G(\Lambda^{(n_1)}) \xrightarrow[n_1 \rightarrow \infty]{} (1/(2\pi e))$ [24].
- A sequence of n_1 -dimensional lattices $\Lambda^{(n_1)}$ is said to be good for AWGN channel coding if for n_1 -dimensional vector $\mathbf{Z} \sim \mathcal{N}(\mathbf{0}, \sigma^2 \mathbf{I}_{n_1})$, $P\{\mathbf{Z} \notin \mathcal{V}(\Lambda^{(n_1)})\}$ vanishes when n_1 goes to ∞ . In this case, $\text{Vol}(\Lambda^{(n_1)}) \xrightarrow[n_1 \rightarrow \infty]{} 2^{n_1 h(\mathbf{Z})}$, where $h(\mathbf{Z}) = (1/2) \log(2\pi e \sigma^2)$ is the differential entropy of \mathbf{Z} [25].
- There exist lattices which are simultaneously good for quantization and channel coding (see [26]).
- Lemma 1 Crypto Lemma [23]. For a dither vector \mathbf{T} independent of \mathbf{X} and uniformly distributed over $\mathcal{V}(\Lambda)$, then $\mathbf{Y} = (\mathbf{X} + \mathbf{T}) \bmod \Lambda$ is uniformly distributed over $\mathcal{V}(\Lambda)$ and is independent of \mathbf{X} .

Consider a pair of n_1 -dimensional nested lattices (Λ_1, Λ_2) such as $\Lambda_2 \subset \Lambda_1$. The fine lattice is Λ_1 with basic Voronoi region \mathcal{V}_1 of volume V_1 and second moment per dimension $\sigma^2(\Lambda_1)$. The coarse lattice is Λ_2 with basic Voronoi region \mathcal{V}_2 of volume V_2 and second moment $\sigma^2(\Lambda_2)$. The following properties of nested lattices hold:

- For $\Lambda_2 \subset \Lambda_1$, we have $Q_{\Lambda_2}(Q_{\Lambda_1}(x)) = Q_{\Lambda_1}(Q_{\Lambda_2}(x)) = Q_{\Lambda_2}(x)$.
- The points of the set $\Lambda_1 \cap \mathcal{V}_2 = \Lambda_1 \bmod \Lambda_2$ represent the coset leaders of Λ_2 relative to Λ_1 , where for each $\lambda \in \{\Lambda_1 \bmod \Lambda_2\}$, the shifted lattice $\Lambda_{2,\lambda} = \Lambda_2 + \lambda$ is called a coset of Λ_2 relative to Λ_1 . There are V_2/V_1 distinct cosets. It follows that the coding rate when using nested lattices is

$$\begin{aligned} R &= \frac{1}{n_1} \log_2 |\Lambda_1 \cap \mathcal{V}_2| \\ &= \frac{1}{n_1} \log_2 \frac{V_2}{V_1} \text{ (bits per dimension)}. \end{aligned} \quad (3)$$

III. ACHIEVABLE RATE REGION FOR TWRC

Theorem 1: For a Gaussian TWRC, under the assumptions a.1 to a.8, the convex hull of the following end-to-end rate-pairs (R_{12}, R_{21}) is achievable:

$$R_{12} \leq \frac{\alpha}{2} \log_2 \left(1 + \frac{|h_1|^2 P_1}{\sigma_R^2 + \frac{\max_{i \in \{1,2\}} |h_i|^2 P_i + \sigma_R^2}{\left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2}\right)^{\frac{1-\alpha}{\alpha}}} - 1}} \right) \quad (4)$$

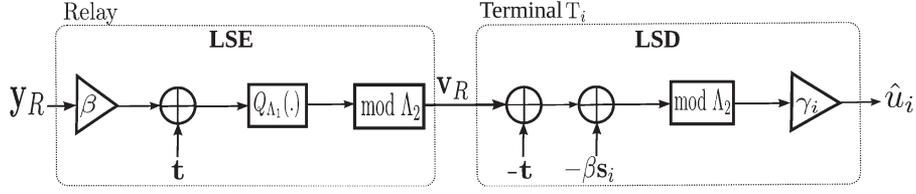


Fig. 2. Lattice encoding at the relay and decoding at T_i , $i = 1, 2$.

$$R_{21} \leq \frac{\alpha}{2} \log_2 \left(1 + \frac{|h_2|^2 P_2}{\sigma_R^2 + \frac{\max_{i \in \{1,2\}} |h_i|^2 P_i + \sigma_R^2}{\left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2}\right)^{\frac{1-\alpha}{\alpha}} - 1}} \right) \quad (5)$$

for $\alpha \in [0, 1]$.

The main idea of the proposed scheme is the following: during the BC phase, the relay sends a quantized version of the signal that was received during the MAC phase. The quantization procedure generates an index which is sent reliably to both users using appropriate channel codes. This index is decoded by both users and, based on their own information (sent during the MAC phase), each source recovers the transmitted message. The proof of *Theorem 1* is detailed in the next paragraphs: in Section III-A, the lattice coding scheme for the source coding is presented. The end-to-end achievable rates are derived in Section III-B and finally in Section III-C the achievable rate region is obtained by optimizing the lattice parameters.

A. Lattice Based Source Coding

We suppose that the elements of \mathbf{X}_i , $i = 1, 2$, are drawn from an independent identically distributed (i.i.d) Gaussian distribution with zero mean and variance P_i . Let $\mathbf{S}_i = h_i \mathbf{X}_i$ be the side information available at terminal T_i , $i = 1, 2$. The signal sent by the relay \mathbf{Y}_R can be written in two ways as the sum of two independent Gaussian r.v.: the side information \mathbf{S}_i and the unknown part $\mathbf{U}_i = \mathbf{Y}_R | \mathbf{S}_i = h_i \mathbf{X}_i + \mathbf{Z}_R$, $i \in \{1, 2\}$. From its received signal, each terminal T_i , $i \in \{1, 2\}$ decodes $\hat{\mathbf{U}}_i$ using \mathbf{S}_i . The variance per dimension of \mathbf{U}_i is $\sigma_{U_i}^2 = \text{VAR}(Y_R | S_i) = |h_i|^2 P_i + \sigma_R^2$.

In the following, we detail the proposed lattice source coding scheme.

1) *Encoding*: The lattice source encoding (LSE) operation is performed with four successive operations: first, the input signal \mathbf{y}_R is scaled with a factor β . Then, a random dither \mathbf{t} which is uniformly distributed over \mathcal{V}_1 is added. This dither is known by all nodes. The dithered scaled version of \mathbf{y}_R , $\beta \mathbf{y}_R + \mathbf{t}$ is quantized to the nearest point in Λ_1 . The outcome of this operation is processed with a modulo-lattice operation in order to generate a vector \mathbf{v}_R of size n_1 as shown in Fig. 2, and defined by:

$$\mathbf{v}_R = Q_{\Lambda_1}(\beta \mathbf{y}_R + \mathbf{t}) \quad \text{mod } \Lambda_2. \quad (6)$$

The relay sends the index of \mathbf{v}_R which identifies a coset of Λ_2 relative to Λ_1 that contains $Q_{\Lambda_1}(\beta \mathbf{y}_R + \mathbf{t})$. By construction,

the coset leader \mathbf{v}_R can be represented using $\log_2(V_2/V_1)$ bits. Thus, the rate of the source encoding scheme employed by the relay is R given by Eq. (3). We further assume that Λ_1 is good for quantization and Λ_2 is good for channel coding [19]. For high dimension n_1 and according to the properties of good lattices, we have $(1/n_1) \log_2(V_i) \approx (1/2) \log_2(2\pi e \sigma^2(\Lambda_i))$, $i \in \{1, 2\}$. Thus R reads

$$R = \frac{1}{2} \log_2 \left(\frac{\sigma^2(\Lambda_2)}{\sigma^2(\Lambda_1)} \right). \quad (7)$$

2) *Decoding*: For both users, \mathbf{v}_R is decoded first. Then $\hat{\mathbf{u}}_i$ is reconstructed with a lattice source decoder (LSD) using the side information \mathbf{s}_i as

$$\hat{\mathbf{u}}_i = \gamma_i ((\mathbf{v}_R - \mathbf{t} - \beta \mathbf{s}_i) \quad \text{mod } \Lambda_2), \quad i = 1, 2 \quad (8)$$

where γ_i , $i \in \{1, 2\}$ are the scaling factors at each decoder.

B. Rate Analysis

At the relay, message m_R corresponding to the index of \mathbf{v}_R is mapped to a codeword \mathbf{x}_R of size n_2 . We assume that the elements of the r.v. \mathbf{X}_R are drawn from an i.i.d Gaussian distribution with zero mean and variance P_R . We consider separate source-channel coding. The broadcast rate from the relay to both terminals is bounded by the capacity of the worst individual relay-terminal channel capacity $\min(I(X_R; Y_1), I(X_R; Y_2))$. From Shannon's source-channel separation theorem [27], we have

$$n_1 R \leq n_2 \min(I(X_R; Y_1), I(X_R; Y_2)). \quad (9)$$

Since real Gaussian codebooks are used for all transmissions, we have: $I(X_R; Y_i) = (1/2) \log_2(1 + (|h_i|^2 P_R)/\sigma_i^2)$, $i = 1, 2$. Finally, by combining Eqs. (7) and (9), we obtain the following constraint on the achievable rates

$$n_1 \log_2 \left(\frac{\sigma^2(\Lambda_2)}{\sigma^2(\Lambda_1)} \right) \leq n_2 \log_2 \left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2} \right). \quad (10)$$

This constraint ensures that index m_R is transmitted reliably to both terminals and \mathbf{v}_R is available at the input of the LSD of both receivers. At terminal T_i , $\hat{\mathbf{u}}_i$ in (8) can be written as:

$$\hat{\mathbf{u}}_i = \gamma_i ((\beta \mathbf{u}_i + \mathbf{e}_q) \quad \text{mod } \Lambda_2) \quad (11)$$

$$= \gamma_i (\beta \mathbf{u}_i + \mathbf{e}_q) \quad (12)$$

where $\mathbf{e}_q = Q_{\Lambda_1}(\beta \mathbf{y}_R + \mathbf{t}) - (\beta \mathbf{y}_R + \mathbf{t}) = -((\beta \mathbf{y}_R + \mathbf{t}) \quad \text{mod } \Lambda_1)$, is the quantization error. By *Lemma 1*, \mathbf{E}_q is independent from \mathbf{Y}_R , and thus from \mathbf{U}_i . Also \mathbf{E}_q is uniformly distributed over \mathcal{V}_1 thus the variance of \mathbf{E}_q per dimension is

$\sigma^2(\Lambda_1)$. Equation (12) is valid only if $\beta \mathbf{u}_i + \mathbf{e}_q \in \mathcal{V}_2$. According to [19], with good channel coding lattices, the probability $\Pr(\beta \mathbf{U}_i + \mathbf{E}_q \notin \mathcal{V}_2)$ vanishes asymptotically provided that:

$$\frac{1}{n_1} \mathbb{E} \|\beta \mathbf{U}_i + \mathbf{E}_q\|^2 = \beta^2 \sigma_{U_i}^2 + \sigma^2(\Lambda_1) \leq \sigma^2(\Lambda_2) \quad (13)$$

With respect to Eq. (13) which is considered as constraint, replacing \mathbf{U}_i by its value, we conclude that:

$$\hat{\mathbf{U}}_i = \gamma_i (\beta (h_1 \mathbf{X}_1 + \mathbf{Z}_R) + \mathbf{E}_q). \quad (14)$$

Let $\mathbf{Z}_{eq,i} = \gamma_i (\beta \mathbf{Z}_R + \mathbf{E}_q)$ be the effective additive noise at terminal T_i . Under high dimension assumption, $n_1 \rightarrow \infty$, we can approximate the uniform random variable \mathbf{E}_q over \mathcal{V}_1 by a Gaussian variable \mathbf{Z}_q with the same variance [24]. Therefore, the communication between terminals T_1 and T_2 (resp. T_2 and T_1) is equivalent to an AWGN channel where the Gaussian noise is given by $\mathbf{Z}_{eq,i}$. hence, the achievable rates of both links satisfy

$$nR_{12} \leq \frac{n_1}{2} \log_2 \left(1 + \frac{\beta^2 |h_1|^2 P_1}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_1)} \right) \quad (15)$$

$$nR_{21} \leq \frac{n_1}{2} \log_2 \left(1 + \frac{\beta^2 |h_2|^2 P_2}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_1)} \right) \quad (16)$$

C. Achievable Rate Region

The rate region that can be achieved by the proposed scheme is characterized by the constraints (15), (16), (10), and (13). Without loss of generality, we assume that $|h_2|^2 P_2 \leq |h_1|^2 P_1$. With this setting, T_2 is the terminal which experiences the weakest side information. Letting $\alpha = n_1/n$, from (10) and (13), the lower bound of $\sigma^2(\Lambda_1)$ is given by

$$\sigma^2(\Lambda_1) \geq \frac{\beta^2 \sigma_{U_2}^2}{\left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2} \right)^{\frac{1-\alpha}{\alpha}} - 1} \quad (17)$$

The rate region defined in (15) and (16) can be rewritten as

$$R_{12} \leq \frac{\alpha}{2} \log_2 (1 + \text{SNR}_{1 \rightarrow 2}) \quad (18)$$

$$R_{21} \leq \frac{\alpha}{2} \log_2 (1 + \text{SNR}_{2 \rightarrow 1}) \quad (19)$$

where $\text{SNR}_{1 \rightarrow 2}$ and $\text{SNR}_{2 \rightarrow 1}$ are the end-to-end SNRs, defined as follows:

$$\text{SNR}_{1 \rightarrow 2} = \frac{\beta^2 |h_1|^2 P_1}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_1)} \quad (20)$$

$$\text{SNR}_{2 \rightarrow 1} = \frac{\beta^2 |h_2|^2 P_2}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_1)} \quad (21)$$

Note that $\text{SNR}_{1 \rightarrow 2}$ and $\text{SNR}_{2 \rightarrow 1}$ are maximized when $\sigma^2(\Lambda_1)$ is minimal. Thus the optimal choice on the second moment of Λ_1 is

$$\sigma^2(\Lambda_1)_{\min} = \frac{\beta^2 \sigma_{U_2}^2}{\left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2} \right)^{\frac{1-\alpha}{\alpha}} - 1} \quad (22)$$

If $|h_1|^2 P_1 \leq |h_2|^2 P_2$, $\sigma_{U_2}^2$ is replaced with $\sigma_{U_1}^2$ in (22). Finally, replacing $\sigma^2(\Lambda_1)_{\min}$ in (20) and (21), Eqs. (4) and (5) are verified and the proof is concluded.

Remark 1: For the transmission problem of the TWRC, the achievable rate region is independent of the choice of the decoders scaling factors γ_i . It is also independent of the encoder scaling factor β provided that $\sigma^2(\Lambda_1)$ is set to its smallest value $\sigma^2(\Lambda_1)_{\min}$ in (22). In the next section, we show that these parameters are involved in the source coding problem that was addressed in [18]. Especially when considering analog signal transmission, the optimisation of these parameters allows to minimize the distortion.

D. Analog Signal Transmission

Since the relay quantizes an analog source, we can consider an end-to-end analog transmission. In this case, the distortion that affects the reconstructed signals at both terminals becomes the main performance metric. The second moment of this distortion is given by

$$\frac{1}{n_1} \mathbb{E} \|\mathbf{Y}_R - \hat{\mathbf{Y}}_{Ri}\|^2 = D_i; \quad i \in \{1, 2\} \quad (23)$$

where $\mathbf{Y}_R = \mathbf{U}_i + \mathbf{S}_i$ and $\hat{\mathbf{Y}}_{Ri} = \hat{\mathbf{U}}_i + \mathbf{S}_i$. By replacing $\hat{\mathbf{U}}_i$ by its value in (12), (23) becomes

$$D_i = (1 - \gamma_i \beta)^2 \sigma_{U_i}^2 + \gamma_i^2 \sigma^2(\Lambda_1); \quad i \in \{1, 2\}. \quad (24)$$

For the analog signal transmission, this distortion has to be minimized to obtain the optimal source coding scheme. For fixed β , the distortion at T_i depends only on two parameters namely γ_i and $\sigma^2(\Lambda_1)$. The optimal distortion can be obtained by calculating the following derivatives:

$$\frac{\partial D_i}{\partial \gamma_i} = 0 \Rightarrow \gamma_i^* = \frac{\beta \sigma_{U_i}^2}{\beta^2 \sigma_{U_i}^2 + \sigma_{\Lambda_1}^2} \quad (25)$$

where γ_i^* , $i \in \{1, 2\}$ are the optimal decoder scaling factors. Since $\gamma_i > 0$, then $(\partial D_i / \partial \sigma^2(\Lambda_1)) > 0$. Thus, the function D_i is increasing with $\sigma^2(\Lambda_1)$ and $\sigma^2(\Lambda_1)_{\min}$ in (22) is the optimal choice that minimizes the distortion at each terminal. Therefore,

$$\gamma_i^* = \frac{\beta \sigma_{U_i}^2}{\beta^2 \sigma_{U_i}^2 + \sigma^2(\Lambda_1)_{\min}}, \quad i \in \{1, 2\}. \quad (26)$$

By replacing $\sigma^2(\Lambda_1)$ and γ_i by their optimal values, we obtain the minimal value of D_i^{\min} , $i \in \{1, 2\}$, given by

$$D_i^{\min} = \frac{\sigma^2(\Lambda_1)_{\min} \sigma_{U_i}^2}{\beta^2 \sigma_{U_i}^2 + \sigma^2(\Lambda_1)_{\min}} \quad (27)$$

$$= \frac{\sigma_{U_2}^2 \sigma_{U_i}^2}{\left(\left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2} \right)^{\frac{1-\alpha}{\alpha}} - 1 \right) \sigma_{U_i}^2 + \sigma_{U_2}^2}. \quad (28)$$

D_i^{\min} , $i \in \{1, 2\}$, just like the achievable rates, are independent of β . However, for a fixed β , the optimal lattice parameters and receivers scaling factors depend on that choice.

Comments on the Distortions: At terminal T_2 , the distortion writes:

$$D_2^{\min} = \frac{\sigma_{U_2}^2 \sigma_{U_2}^2}{(A-1)\sigma_{U_2}^2 + \sigma_{U_2}^2} = \frac{\sigma_{U_2}^2}{A}$$

where $A = (1 + \min_{i \in \{1,2\}} (|h_i|^2 P_R / \sigma_i^2))^{1-\alpha/\alpha}$. It can be reformulated as

$$\frac{\sigma_{U_2}^2}{D_2^{\min}} = \left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2}\right)^{\frac{1-\alpha}{\alpha}}$$

$$\alpha \log_2 \left(\frac{\sigma_{U_2}^2}{D_2^{\min}}\right) = (1-\alpha) \log_2 \left(1 + \min_{i \in \{1,2\}} \frac{|h_i|^2 P_R}{\sigma_i^2}\right). \quad (29)$$

We find, in the left hand side of (29), the Wyner-Ziv rate distortion function of the Gaussian source \mathbf{Y}_R with side information \mathbf{S}_2 at the decoder T_2 [28]. It is defined as the minimum rate needed to achieve D_2^{\min} and it is given by:

$$R_{WZ}(D_2^{\min}) = \frac{1}{2} \log_2 \left(\frac{\sigma_{U_2}^2}{D_2^{\min}}\right) \quad (30)$$

Note that the source coding rate is no larger than the channel coding rate to the relay. Also, according to (26) the optimal value of γ_2 is given by

$$\gamma_2^* = \frac{\beta \sigma_{U_2}^2}{\beta^2 \sigma_{U_2}^2 + \sigma^2(\Lambda_1)_{\min}}.$$

With the choice $\beta = \gamma_2^*$, we get $\beta = \sqrt{1 - (D_2^{\min}/\sigma_{U_2}^2)}$. This is in accordance with the optimal scaling factor reported in [18], [28] for the optimum Gaussian forward test channel. For this choice of β , $\sigma^2(\Lambda_1)_{\min} = D_2^{\min}$ which is consistent with the source coding parameters choices in [18].

At terminal T_1 , the reconstruction distortion is smaller than D_2^{\min} of terminal T_2 . This is compatible with the fact that T_1 has the best side information quality and the proposed achievable scheme is optimal for the worst user.

IV. IMPROVED ACHIEVABLE RATE REGION FOR TWRC

In the previous section, we presented a PNC scheme in which a common information is sent from the relay to both users. The rates that are achievable by this scheme depend only on the ratio $(\sigma^2(\Lambda_1)_{\min})/\beta^2$. This ratio is determined, as shown by (22), essentially by the variance $\sigma_{U_i}^2$ of the unknown part of the source at the terminal T_i and the lowest channel coefficient amplitude $\min_{i \in \{1,2\}} (|h_i|^2/\sigma_i^2)$. Thus, the achievable rates are limited by the user which has the weakest side information and also the worst channel condition. In this case, the best user suffers from this limitation on its achievable rate. In order to improve its rate, an additional refinement information can be sent from the relay, that can be only decoded by the best user.

Without loss of generality, assume that terminal T_1 has better channel condition and side information than T_2 i.e., $|h_1| \geq |h_2|$ and $|h_2|^2 P_1 \geq |h_2|^2 P_2$. The following theorem provides an achievable rate region for the TWRC, obtained using the refinement scheme.

Theorem 2: For a Gaussian TWRC, under the assumptions a.1 to a.8, the convex hull of the following end-to-end rate-pairs (R_{12}, R_{21}) is achievable, for $\alpha, \nu \in [0, 1]$, as given in (31) and (32), shown at the bottom of the page.

As mentioned previously, the main idea of the coding scheme employed for *Theorem 2* is that the relay should be sending two descriptions of its received signal, a common layer that is intended to be recovered by both users and an individual or refinement layer that is intended to be recovered only by the best user, i.e., terminal T_1 .

The proof of *Theorem 2* is detailed below.

A. Doubly Nested Lattices for Source Coding

We use a doubly nested lattice chain $(\Lambda_0, \Lambda_1, \Lambda_2)$ such as $\Lambda_2 \subset \Lambda_1 \subset \Lambda_0$. We require that Λ_2 is good for channel coding, Λ_1 is simultaneously good for channel and source coding and Λ_0 is good for source coding.

From these lattices, we form three codebooks

$$\mathcal{C}_c = \Lambda_1 \cap \mathcal{V}_2$$

$$\mathcal{C}_r = \Lambda_0 \cap \mathcal{V}_1$$

$$\mathcal{C}_1 = \Lambda_0 \cap \mathcal{V}_2$$

$$R_{12} \leq \frac{\alpha}{2} \log_2 \left(1 + \frac{|h_1|^2 P_1}{\sigma_R^2 + \frac{|h_1|^2 P_1 + \sigma_R^2}{\left(1 + \frac{\nu |h_2|^2 P_R}{(1-\nu) |h_2|^2 P_R + \sigma_2^2}\right)^{\frac{1-\alpha}{\alpha}} - 1}} \right) \quad (31)$$

$$R_{21} \leq \frac{\alpha}{2} \log_2 \left(1 + \frac{|h_2|^2 P_2}{\sigma_R^2 + \frac{|h_1|^2 P_1 + \sigma_R^2}{\left(1 + \frac{(1-\nu) |h_1|^2 P_R}{\sigma_1^2}\right)^{\frac{1-\alpha}{\alpha}} \left[\left(1 + \frac{\nu |h_2|^2 P_R}{(1-\nu) |h_2|^2 P_R + \sigma_2^2}\right)^{\frac{1-\alpha}{\alpha}} - 1 \right]}} \right) \quad (32)$$

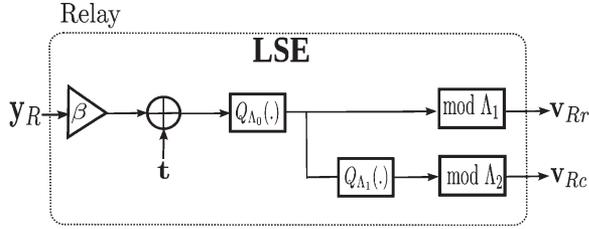


Fig. 3. Layered Lattice encoding at the relay.

with the following coding rates:

$$R_c = \frac{1}{n_1} \log_2 \left(\frac{V_2}{V_1} \right) \xrightarrow{n_1 \rightarrow \infty} \frac{1}{2} \log_2 \left(\frac{\sigma^2(\Lambda_2)}{\sigma^2(\Lambda_1)} \right) \quad (33)$$

$$R_r = \frac{1}{n_1} \log_2 \left(\frac{V_1}{V_0} \right) \xrightarrow{n_1 \rightarrow \infty} \frac{1}{2} \log_2 \left(\frac{\sigma^2(\Lambda_1)}{\sigma^2(\Lambda_0)} \right) \quad (34)$$

$$\begin{aligned} R_1 &= R_c + R_r \\ &= \frac{1}{n_1} \log_2 \left(\frac{V_2}{V_0} \right) \xrightarrow{n_1 \rightarrow \infty} \frac{1}{2} \log_2 \left(\frac{\sigma^2(\Lambda_2)}{\sigma^2(\Lambda_0)} \right) \end{aligned} \quad (35)$$

where R_c is the common source rate, R_r is the refinement source rate and R_1 is the total source rate at terminal T_1 .

1) *Encoding*: Fig. 3 shows the LSE operation. The input signal \mathbf{y}_R is scaled with a factor β . Then, a random dither \mathbf{t} which is uniformly distributed over \mathcal{V}_1 is added. This dither is known by all nodes. The dithered scaled version of \mathbf{y}_R , $\beta\mathbf{y}_R + \mathbf{t}$, is quantized to the nearest point in Λ_0 . The outcome of this operation is then processed to generate two messages. First, the coset leader of Λ_1 relative to Λ_0 , \mathbf{v}_{Rr} , is generated by a modulo-lattice operation. The index of \mathbf{v}_{Rr} identifies the refinement message. Then, another quantization to the nearest point in Λ_1 is performed and processed with another modulo-lattice operation to generate the coset leader of Λ_2 relative to Λ_1 , \mathbf{v}_{Rc} . The index of \mathbf{v}_{Rc} identifies the common message. Both messages are defined as:

$$\mathbf{v}_{Rr} = Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t}) \quad \text{mod } \Lambda_1 \quad (36)$$

$$\begin{aligned} \mathbf{v}_{Rc} &= Q_{\Lambda_1}(Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t})) \quad \text{mod } \Lambda_2 \\ &= Q_{\Lambda_1}(\beta\mathbf{y}_R + \mathbf{t}) \quad \text{mod } \Lambda_2. \end{aligned} \quad (37)$$

It can easily be seen that $\mathbf{v}_{Rr} \in \mathcal{C}_r$ and $\mathbf{v}_{Rc} \in \mathcal{C}_c$. We obtain the same common information generated in (6). Thus, the (total) information that is intended to terminal T_1 is such that

$$\mathbf{v}_{R1} = \mathbf{v}_{Rr} + \mathbf{v}_{Rc} \quad (38a)$$

$$\begin{aligned} &= Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t}) \quad \text{mod } \Lambda_1 + Q_{\Lambda_1}(\beta\mathbf{y}_R + \mathbf{t}) \quad \text{mod } \Lambda_2 \\ &= Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t}) - Q_{\Lambda_1}(Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t})) \\ &\quad + Q_{\Lambda_1}(\beta\mathbf{y}_R + \mathbf{t}) - Q_{\Lambda_2}(Q_{\Lambda_1}(\beta\mathbf{y}_R + \mathbf{t})) \end{aligned} \quad (38b)$$

$$\begin{aligned} &= Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t}) - Q_{\Lambda_2}(Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t})) \\ &\quad + Q_{\Lambda_1}(\beta\mathbf{y}_R + \mathbf{t}) - Q_{\Lambda_2}(Q_{\Lambda_1}(\beta\mathbf{y}_R + \mathbf{t})) \end{aligned} \quad (38c)$$

$$= Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t}) - Q_{\Lambda_2}(\beta\mathbf{y}_R + \mathbf{t}) \quad (38d)$$

$$= Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t}) - Q_{\Lambda_2}(Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t})) \quad (38e)$$

$$= Q_{\Lambda_0}(\beta\mathbf{y}_R + \mathbf{t}) \quad \text{mod } \Lambda_2 \quad (38f)$$

where (38c), (38d), and (38e) follow using the properties of the modulo operation as given in Section II.

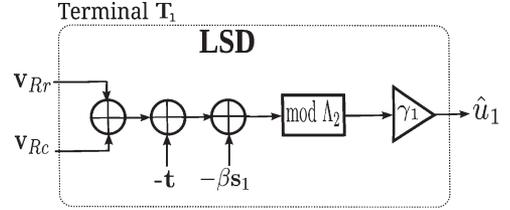


Fig. 4. Lattice source decoding at the Terminal 1.

2) *Decoding*: \mathbf{v}_{Rc} is decoded at terminal T_2 . Then, $\hat{\mathbf{u}}_2$ is reconstructed with an LSD using the side information \mathbf{s}_2 as

$$\hat{\mathbf{u}}_2 = \gamma_2((\mathbf{v}_{Rc} - \mathbf{t} - \beta\mathbf{s}_2) \quad \text{mod } \Lambda_2). \quad (39)$$

At terminal T_1 , \mathbf{v}_{Rc} and \mathbf{v}_{Rr} are both decoded correctly. These coset leaders are used to recalculate the total information \mathbf{v}_{R1} from (38a). Finally, the decoder reconstructs $\hat{\mathbf{u}}_1$ as defined by (40) and shown in Fig. 4, as

$$\hat{\mathbf{u}}_1 = \gamma_1((\mathbf{v}_{R1} - \mathbf{t} - \beta\mathbf{s}_1) \quad \text{mod } \Lambda_2) \quad (40)$$

B. Rate Analysis

The relay generates the indices of \mathbf{v}_{Rc} and \mathbf{v}_{Rr} . Then they are mapped to the channel codewords \mathbf{x}_{Rc} and \mathbf{x}_{Rr} . The relay sends $\mathbf{x}_R(m_R)$ which is the superposition of \mathbf{x}_{Rc} and \mathbf{x}_{Rr} with transmit power νP_R and $(1 - \nu)P_R$, $\nu \in \{0, 1\}$, respectively. The refinement codeword \mathbf{x}_{Rr} is encoded on top of the common codeword \mathbf{x}_{Rc} and it is treated as an interference while decoding the common message. Thus, $\mathbf{X}_{Rc} \rightarrow \mathbf{X}_r \rightarrow (\mathbf{Y}_1, \mathbf{Y}_2)$ forms a Markov chain. As described in previous single layer PNC scheme, the broadcast rate is bounded by the worst relay-terminal channel capacity for the common message, and by the relay- T_1 channel for the refinement message. In addition, the source-channel separation ensures that the codewords \mathbf{x}_{Rc} and \mathbf{x}_{Rr} are transmitted reliably to the terminals and that \mathbf{v}_{Rc} and \mathbf{v}_{Rr} are available at the LSD input of corresponding receivers. Therefore, the rates are such that

$$n_1 R_c \leq n_2 \min\{I(X_{Rc}; Y_1), I(X_{Rc}; Y_2)\} \quad (41)$$

$$n_1 R_r \leq n_2 I(X_{Rr}; Y_1 | X_{Rc}) \quad (42)$$

For real Gaussian codebooks, we have

$$I(X_{Rc}; Y_1) = \frac{1}{2} \log_2 \left(1 + \frac{\nu |h_1|^2 P_R}{(1 - \nu) |h_1|^2 P_R + \sigma_1^2} \right)$$

$$I(X_{Rc}; Y_2) = \frac{1}{2} \log_2 \left(1 + \frac{\nu |h_2|^2 P_R}{(1 - \nu) |h_2|^2 P_R + \sigma_2^2} \right)$$

$$I(X_{Rr}; Y_1 | X_{Rc}) = \frac{1}{2} \log_2 \left(1 + \frac{(1 - \nu) |h_1|^2 P_R}{\sigma_1^2} \right).$$

Since $|h_2| \leq |h_1|$, $\min\{I(X_{Rc}; Y_1), I(X_{Rc}; Y_2)\} = I(X_{Rc}; Y_2)$. Using (33), (34), (41), and (42), the rates' conditions become

$$n_1 \log_2 \left(\frac{\sigma^2(\Lambda_2)}{\sigma^2(\Lambda_1)} \right) \leq n_2 \log_2 \left(1 + \frac{\nu |h_2|^2 P_R}{(1 - \nu) |h_2|^2 P_R + \sigma_2^2} \right) \quad (43)$$

$$n_1 \log_2 \left(\frac{\sigma^2(\Lambda_1)}{\sigma^2(\Lambda_0)} \right) \leq n_2 \log_2 \left(1 + \frac{(1-\nu)|h_1|^2 P_R}{\sigma_1^2} \right). \quad (44)$$

Now, $\hat{\mathbf{u}}_1$ and $\hat{\mathbf{u}}_2$ can be obtained using (40) and (39), respectively. At terminal T_2 , $\hat{\mathbf{u}}_2$ can be written as:

$$\hat{\mathbf{u}}_2 = \gamma_2 ((\beta \mathbf{u}_2 + \mathbf{e}_{q,1}) \bmod \Lambda_2) \quad (45)$$

$$= \gamma_2 (\beta \mathbf{u}_2 + \mathbf{e}_{q,1}) \quad (46)$$

where $\mathbf{e}_{q,1}$ is the quantization error at lattice Λ_1 given by

$$\begin{aligned} \mathbf{e}_{q,1} &= Q_{\Lambda_1}(\beta \mathbf{y}_R + \mathbf{t}) - (\beta \mathbf{y}_R + \mathbf{t}) \\ &= -((\beta \mathbf{y}_R + \mathbf{t}) \bmod \Lambda_1) \end{aligned}$$

and (46) can be obtained by proceeding as in Section III-B. Note that $\Pr(\beta \mathbf{U}_2 + \mathbf{E}_{q,1} \notin \mathcal{V}_2)$ vanishes asymptotically provided that:

$$\frac{1}{n_1} \mathbb{E} \|\beta \mathbf{U}_2 + \mathbf{E}_{q,1}\|^2 = \beta^2 \sigma_{U_2}^2 + \sigma^2(\Lambda_1) \leq \sigma^2(\Lambda_2). \quad (47)$$

In this case, the rate achievable at terminal T_2 is such that

$$nR_{12} \leq \frac{n_1}{2} \log_2 \left(1 + \frac{\beta^2 |h_1|^2 P_1}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_1)} \right). \quad (48)$$

At terminal T_1 , $\hat{\mathbf{u}}_1$ can be obtained as

$$\hat{\mathbf{u}}_1 = \gamma_1 ((\beta \mathbf{u}_1 + \mathbf{e}_{q,0}) \bmod \Lambda_2) \quad (49)$$

$$\equiv \gamma_1 (\beta \mathbf{u}_1 + \mathbf{e}_{q,0}) \quad (50)$$

where $\mathbf{e}_{q,0}$ is the modulo- Λ_0 quantization error given by

$$\begin{aligned} \mathbf{e}_{q,0} &= Q_{\Lambda_0}(\beta \mathbf{y}_R + \mathbf{t}) - (\beta \mathbf{y}_R + \mathbf{t}) \\ &= -((\beta \mathbf{y}_R + \mathbf{t}) \bmod \Lambda_0) \end{aligned}$$

and (50) holds if $\beta \mathbf{u}_1 + \mathbf{e}_{q,0} \in \mathcal{V}_2$. Note that, by using Lemma 1, $\mathbf{E}_{q,0}$ is independent from \mathbf{Y}_R , and thus from \mathbf{U}_1 . Also this quantization error is uniformly distributed over \mathcal{V}_0 . Therefore, $\text{VAR}(\mathbf{E}_{q,0}) = \sigma^2(\Lambda_0)$. The probability $\Pr(\beta \mathbf{U}_1 + \mathbf{E}_{q,0} \notin \mathcal{V}_2)$ vanishes asymptotically provided that:

$$\frac{1}{n_1} \mathbb{E} \|\beta \mathbf{U}_1 + \mathbf{E}_{q,0}\|^2 = \beta^2 \sigma_{U_1}^2 + \sigma^2(\Lambda_0) \leq \sigma^2(\Lambda_2). \quad (51)$$

Thus,

$$\hat{\mathbf{U}}_1 = \gamma_1 (\beta h_1 \mathbf{X}_2 + \beta \mathbf{Z}_R + \mathbf{E}_{q,0}).$$

Communication from terminal T_2 to terminal T_1 is equivalent to that over an AWGN channel with noise $\gamma_1 (\beta \mathbf{Z}_R + \mathbf{E}_{q,0})$. Hence the achievable rate of this link satisfies:

$$nR_{21} \leq \frac{n_1}{2} \log_2 \left(1 + \frac{\beta^2 |h_2|^2 P_2}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_0)} \right). \quad (52)$$

C. Achievable Rate Region

The rate region that is achievable using the coding scheme that we described so far can be obtained using (43), (44), (47),

and (51). Letting $n_1/n = \alpha$, we get

$$\begin{cases} \frac{\sigma^2(\Lambda_2)}{\sigma^2(\Lambda_1)} \leq \left(1 + \frac{\nu |h_2|^2 P_R}{(1-\nu) |h_2|^2 P_R + \sigma_2^2} \right)^{\frac{1-\alpha}{\alpha}} \\ \frac{\sigma^2(\Lambda_1)}{\sigma^2(\Lambda_0)} \leq \left(1 + \frac{(1-\nu) |h_1|^2 P_R}{\sigma_1^2} \right)^{\frac{1-\alpha}{\alpha}} \\ \sigma^2(\Lambda_1) \leq \sigma^2(\Lambda_2) - \beta^2 \sigma_{U_2}^2 \\ \sigma^2(\Lambda_0) \leq \sigma^2(\Lambda_2) - \beta^2 \sigma_{U_1}^2 \end{cases}$$

Since $\sigma^2(\Lambda_2) \geq \sigma^2(\Lambda_1) \geq \sigma^2(\Lambda_0)$, the last constraint in the system is not active. Thus we obtain the following bounds on the second moment of the lattices

$$\sigma^2(\Lambda_1) \geq \frac{\beta^2 \sigma_{U_2}^2}{\left(1 + \frac{\nu |h_2|^2 P_R}{(1-\nu) |h_2|^2 P_R + \sigma_2^2} \right)^{\frac{1-\alpha}{\alpha}} - 1} \quad (53)$$

$$\sigma^2(\Lambda_0) \geq \frac{\sigma_{\Lambda_1}^2}{\left(1 + \frac{(1-\nu) |h_1|^2 P_R}{\sigma_1^2} \right)^{\frac{1-\alpha}{\alpha}}}. \quad (54)$$

The rate region defined by (48) and (52) can then be rewritten equivalently as

$$R_{12} \leq \frac{\alpha}{2} \log_2(1 + \text{SNR}_{1 \rightarrow 2}) \quad (55)$$

$$R_{21} \leq \frac{\alpha}{2} \log_2(1 + \text{SNR}_{2 \rightarrow 1}) \quad (56)$$

where the end-to-end SNRs are given by

$$\text{SNR}_{1 \rightarrow 2} = \frac{\beta^2 |h_1|^2 P_1}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_1)} \quad (57)$$

$$\text{SNR}_{2 \rightarrow 1} = \frac{\beta^2 |h_2|^2 P_2}{\beta^2 \sigma_R^2 + \sigma^2(\Lambda_0)}. \quad (58)$$

It is easily seen that one obtains larger rates if the inequalities in (57) and (58) hold with equality, i.e., the optimal choice on the second moment of Λ_1 is

$$\sigma^2(\Lambda_1)_{\min} = \frac{\beta^2 \sigma_{U_2}^2}{\left(1 + \frac{\nu |h_2|^2 P_R}{(1-\nu) |h_2|^2 P_R + \sigma_2^2} \right)^{\frac{1-\alpha}{\alpha}} - 1} \quad (59)$$

and the optimal choice on the second moment of Λ_0 is

$$\sigma^2(\Lambda_0)_{\min} = \frac{\sigma^2(\Lambda_1)_{\min}}{\left(1 + \frac{(1-\nu) |h_1|^2 P_R}{\sigma_1^2} \right)^{\frac{1-\alpha}{\alpha}}} \quad (60)$$

Finally, by substituting $\sigma^2(\Lambda_1)_{\min}$ and $\sigma^2(\Lambda_0)_{\min}$ in (57) and (58), we get (31) and (32). This completes the proof of *Theorem 2*.

Remark 2: The obtained achievable rates are independent of the choice of the scaling factors β and γ_i . The optimal choice of these parameters is explained when considering the source coding problem as explained in the next section.

D. Analog Signal Transmission

Proceeding as in the analysis in III-D, it can be easily obtained that the optimal scaling factors γ_i that minimize the distortion at each terminal are given by

$$\gamma_1^* = \frac{\beta\sigma^2(\Lambda_1)}{\beta^2\sigma_{U_2}^2 + \sigma^2(\Lambda_1)} \quad (61)$$

$$\gamma_2^* = \frac{\beta\sigma^2(\Lambda_0)}{\beta^2\sigma_{U_1}^2 + \sigma^2(\Lambda_0)}. \quad (62)$$

Thus, the minimal distortion at terminal T_2 is

$$D_2^{\min} = \frac{\sigma_{U_2}^2}{\left(1 + \frac{\nu|h_2|^2 P_R}{(1-\nu)|h_2|^2 P_R + \sigma_2^2}\right)^{\frac{1-\alpha}{\alpha}}} \quad (63)$$

and the minimal distortion at terminal T_1 is

$$\begin{aligned} D_1^{\min} &= \frac{\sigma_{U_1}^2 \sigma^2(\Lambda_0)_{\min}}{\beta^2 \sigma_{U_1}^2 + \sigma^2(\Lambda_0)_{\min}} \quad (64) \\ &= \sigma_{U_2}^2 \sigma_{U_1}^2 \left[\left(1 + \frac{(1-\nu)|h_1|^2 P_R}{\sigma_1^2}\right)^{\frac{1-\alpha}{\alpha}} \right. \\ &\quad \left. \times \left(\left(1 + \frac{\nu|h_2|^2 P_R}{(1-\nu)|h_2|^2 P_R + \sigma_2^2}\right)^{\frac{1-\alpha}{\alpha}} - 1 \right) \sigma_{U_1}^2 + \sigma_{U_2}^2 \right]^{(-1)}. \quad (65) \end{aligned}$$

Observe that the distortion D_1^{\min} that is allowed by the layered coding scheme described so far is, as expected, smaller than that of the coding scheme of Section III given by (27).

To summarize, if we are interested in the distortion problem in addition to the transmission problem addressed in this paper, the choice of β can be left to the designer. On one hand, the optimal lattice parameters and the receivers' scaling factors that depend on the chosen value of β are given by (22) and (26) for the first scheme and (59), (60), (61), and (62) for the second scheme. On the other hand, the choice of any value of β does not affect the optimal end-to-end achievable rates and distortions that depend only on the system parameters.

V. NUMERICAL RESULTS

This section presents numerical results of the achievable rates of our proposed schemes compared to AF and DF protocols and the outer-bound capacity given in [3], [13].

We select the time-division parameter $\alpha \in [0, 1]$ that permits to trade among the multiaccess and broadcast phases in a manner that maximizes the users rates. The bounds are determined by maximizing the weighted sum of the rates R_{12} and R_{21} for each protocol. For example, for the scheme of Section IV, we solve the following problem for all values of $\eta \in [0, 1]$

$$\max \quad \eta R_{12} + (1 - \eta) R_{21} \quad (66a)$$

$$\text{s.t.} \quad (R_{12}, R_{21}) \text{ satisfy (31) and (32)} \quad (66b)$$

$$\text{for } \alpha \text{ and } \nu \in [0, 1] \quad (66c)$$

It is worth noting that the time division α with AF relaying scheme is set optimally to $1/2$.

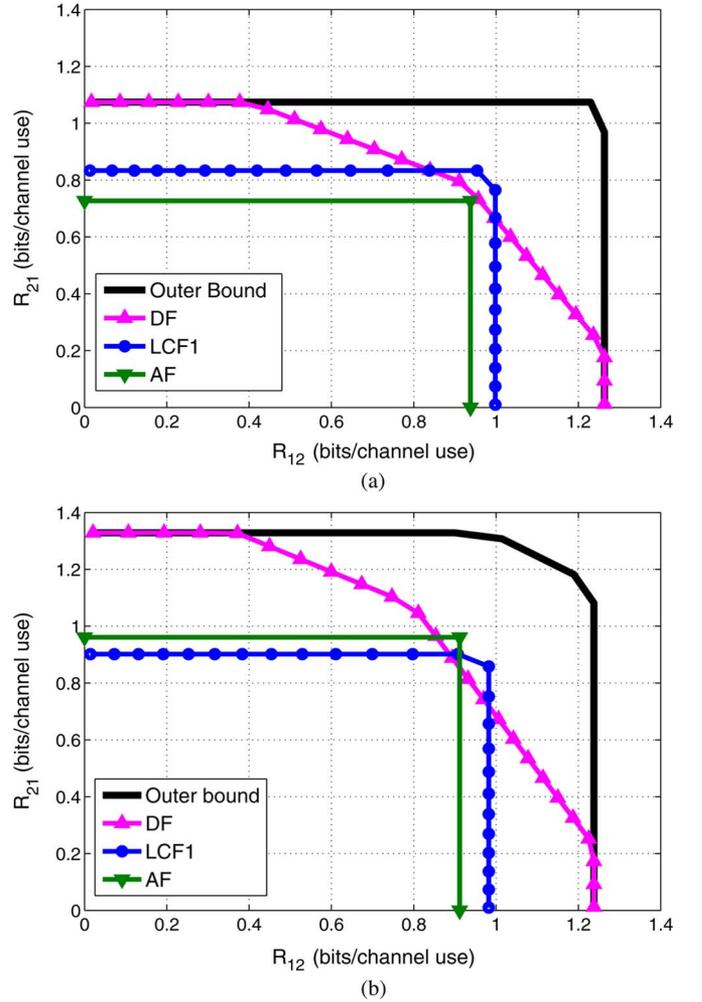


Fig. 5. Achievable rate regions and the outer bound capacity of the Gaussian TWRC. In the left, T_1 has the best transmit power and the worst channel. In the right, T_2 has the best transmit power and the worst channel. (a) $P_1 = 15$ dB, $P_2 = 10$ dB, $P_R = 20$ dB, $|h_1|^2 = 0.5$, $|h_2|^2 = 1$; (b) $P_1 = 10$ dB, $P_2 = 15$ dB, $P_R = 20$ dB, $|h_1|^2 = 2$, $|h_2|^2 = 0.5$.

We consider equal noise variances $\sigma_1^2 = \sigma_2^2 = \sigma_R^2 = 1$, different transmit powers and asymmetric channels with $|h_1|^2 P_1 \geq |h_2|^2 P_2$. For convenience, we refer to the achievable rate regions of Theorems 1 and 2 respectively as LCF1 and LCF2.

Fig. 5 shows the rates allowed by AF, DF and our proposed scheme LCF1 for two different setups: i) terminal T_2 experiencing better channel conditions and having less power than terminal T_1 in Fig. 5(a), and ii) terminal T_1 experiencing better channel conditions and having less power than terminal T_2 in Fig. 5(b).

Note that our scheme LCF1 is, in essence, a CF relaying strategy that is tailored appropriately for the TWRC. Being based on linear (lattice) coding, this strategy has been shown in [18] to achieve the same rates as those allowed by random coding [13], [14]. It has been shown in [13], that CF strategy achieves rates that are larger than those by AF for symmetric power and channel configurations. However, this result is not verified for asymmetric channels. This is shown in Fig. 5 where the difference between the rate regions of AF and LCF1 is negligible for moderate SNR values and asymmetric channels.

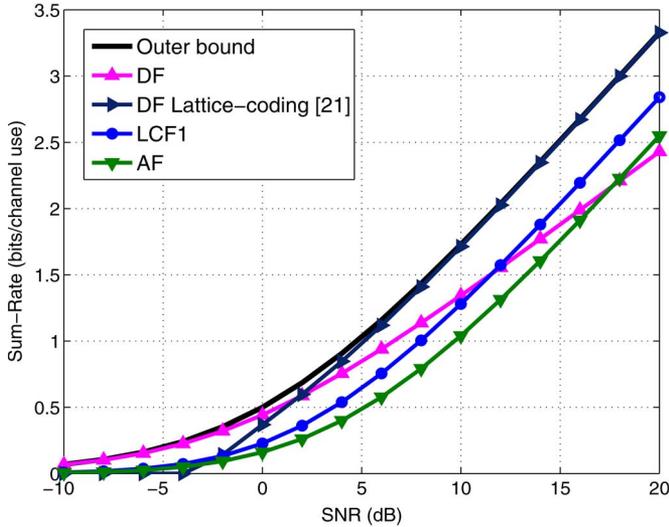


Fig. 6. Maximum sum-rate for symmetric channels: $\text{SNR} = \text{SNR}_{1R} = \text{SNR}_{R1} = \text{SNR}_{2R} = \text{SNR}_{R2}$. LCF1 outperforms AF and DF for $\text{SNR} > 11$ dB.

Fig. 6 illustrates the performance of all schemes in the symmetric power and channel conditions case. We consider also for comparison the DF lattice-coding scheme proposed in [21] and extended for asymmetric channels in [22]. End-to-end maximum sum-rate $R_{12} + R_{21}$ is drawn as a function of SNR for equal channel and power conditions for all nodes. Define $\text{SNR}_{ij} = (|h_{ij}|^2 P_i) / \sigma_j^2$. It is clearly seen that LCF1 outperforms DF for $\text{SNR}_s \geq 12$ dB. This result can be interpreted analytically. In fact, it can be seen easily that for small SNR values, DF rate approaches

$$R_{\text{DF}} \rightarrow \max_{\alpha} \min \{ \alpha \text{SNR}, (1 - \alpha) \text{SNR} \} = \frac{1}{4} \text{SNR}.$$

Also, the rate offered by LCF1 approaches

$$R_{\text{LCF1}} \rightarrow \frac{\text{SNR}^2}{2} \left(1 + \frac{\left(1 + 2 \left(\sqrt{\text{SNR}} - 1 \right)^2 \right) \sqrt{\text{SNR}}}{2 \left(\sqrt{\text{SNR}} + 1 - 1 \right) + \sqrt{\text{SNR}}} \right)$$

Thus, in such small SNR regime, we have $R_{\text{LCF1}} \leq R_{\text{DF}}$. On the other hand, for high SNR, DF rate can be approximated by

$$R_{\text{DF}} \rightarrow \frac{1}{6} \log_2(\text{SNR})$$

and LCF1 rate approaches

$$R_{\text{LCF1}} \rightarrow \frac{1}{4} (\log_2(\text{SNR}) - 1).$$

Therefore, for large SNRs, $R_{\text{LCF1}} \geq R_{\text{DF}}$ that reflects the result in Fig. 6. More generally, for equal channel conditions and transmit power at both terminals, i.e., $\text{SNR} = \text{SNR}_{1R} = \text{SNR}_{2R}$, when $\text{SNR}_{Ri} \gg \text{SNR}$, $R_{\text{LCF1}} > R_{\text{DF}}$ and when $\text{SNR}_{Ri} \ll \text{SNR}$, $R_{\text{LCF1}} < R_{\text{DF}}$ $i \in 1, 2$. In other words, when the relay power is too high compared to the terminals power, CF is better than DF and vice versa. This result is consistent with the previous comparison in Fig. 6, where we show that LCF1 is better than DF for high SNR regime. More

generally, the SNR threshold required for LCF1 to outperform DF decreases as the relay power increases. In particular, in middle to high SNR regime at both terminals, if the relay has limited transmit power, LCF1 can be better than DF. For example, for $\text{SNR} = 25$ dB and $P_R = 0$ dB, CF is better than DF.

The achievable rates of DF-Lattice coding scheme [21], [22] when considering time division optimization are given by (67) and (68). This scheme achieves rates within 1/2 bit of the upper bound. This bound becomes tight for high SNR as depicted in Fig. 6. However, for very low SNR, the minimum in the right hand side in both equations (67) and (68) is equal to zero. In this case, the achievable rates of this scheme are equal to zero. All the other schemes, in this case, outperforms DF-Lattice coding

$$R_{12} \leq \min \left\{ \left[\frac{\alpha}{2} \log \left(\frac{|h_1|^2 P_1}{|h_1|^2 P_1 + |h_2|^2 P_2} + \frac{|h_1|^2 P_1}{\sigma_R^2} \right) \right]^+, \frac{(1 - \alpha)}{2} \log \left(1 + \frac{|h_2|^2 P_R}{\sigma_2^2} \right) \right\} \quad (67)$$

$$R_{21} \leq \min \left\{ \left[\frac{\alpha}{2} \log \left(\frac{|h_2|^2 P_2}{|h_1|^2 P_1 + |h_2|^2 P_2} + \frac{|h_2|^2 P_2}{\sigma_R^2} \right) \right]^+, \frac{(1 - \alpha)}{2} \log \left(1 + \frac{|h_1|^2 P_R}{\sigma_1^2} \right) \right\} \quad (68)$$

where $[x]^+ = \max(0, x)$.

In what follows, we consider channel parameters combinations such that $|h_1|^2 P_1 \geq |h_2|^2 P_2$ and $|h_1|^2 \geq |h_2|^2$. Fig. 7 draws the achievable rate regions of LCF1 and LCF2. One can see that the two-layer based scheme LCF2 enlarges the rate region compared to the basic scheme since the relay sends additional information to the best terminal T_1 . For the setting presented in Fig. 7(a), the achievable rate R_{21} increases by 60% due to the additional refinement individual description. Fig. 7(b) illustrates this aspect for a different choice of the channel parameters where R_{21} increases by more than 100%.

Fig. 8(a) shows the maximum sum-rate as a function of the transmit SNR for asymmetric channel condition and equal power constraints. At lower SNRs, LCF2 outperforms DF-lattice coding, while at higher SNRs ($\text{SNR} \geq 11$ dB), DF-lattice coding is better. It is important to stress here that with LCF2 and LCF1, the relay uses less information than DF-based schemes to reconstruct its encoder. These schemes have also less complexity since the relay does not have to decode any message. Fig. 8(b) shows ν , the fraction of relay power P_R allocated to the common message. $(1 - \nu)P_R$ represents the power allocated to the refinement message. For the considered channel settings, although the common message gets more than 90% of the relay power, the remaining power is sufficient to ameliorate the performance of LCF2 compared to LCF1 by 10% at high SNR.

Fig. 9 illustrates the achievable rate regions of all the schemes for various SNR settings. At low SNR regime, the scheme LCF2 outperforms the scheme LCF1; but they both fall short of attaining the same performance as that offered by DF which is nearly optimal. In fact, in this SNR regime, the rate region obtained

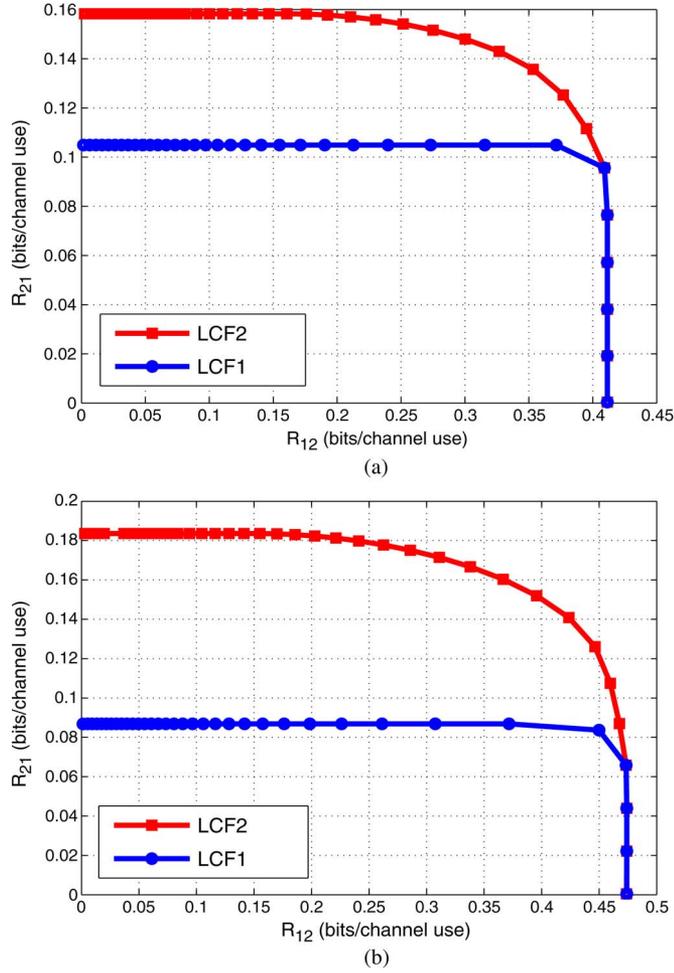


Fig. 7. Achievable rate regions of LCF1 and LCF2. LCF2 achieves better end-to-end rates at T_1 . (a) $P_1 = 10$ dB, $P_2 = P_R = 5$ dB, $|h_1|^2 = 2$, $|h_2|^2 = 0.5$; (b) $P_1 = 10$ dB, $P_2 = P_R = 5$ dB, $|h_1|^2 = 6$, $|h_2|^2 = 0.5$.

with DF relaying approaches relatively closely the outer bound as can be seen in Fig. 9(d). Note that our observation here is consistent with the results in [13], [29] that showed that DF scheme is better than the other relaying schemes for low SNR region.

At very large SNRs, LCF1 and LCF2 achieve better sum-rates than DF as shown in Fig. 9(a). In this case, DF-lattice coding is optimal since it coincides with the outer bound. This scheme approaches the capacity asymptotically as the uplink SNRs increase, i.e., SNR_{1R} and SNR_{2R} . However, as these SNRs decrease, the achievable rates of this scheme approach zero as depicted in Fig. 9(e).

At moderate to large SNRs, LCF2 scheme can achieve sum-rates higher than classic DF. For low to moderate SNRs, It can achieve sum-rates higher than DF-lattice coding. Finally, simulations show that LCF2 scheme outperforms AF in all SNR regimes for symmetric and asymmetric configurations. The proposed schemes present a trade-off between performance and complexity compared to the other schemes.

Fig. 10 draws the fraction of power allocated by the relay to the common message in LCF2. As the relay transmit power increases, the power allocated to the refinement message decreases. At high SNR regime, with favourable relay

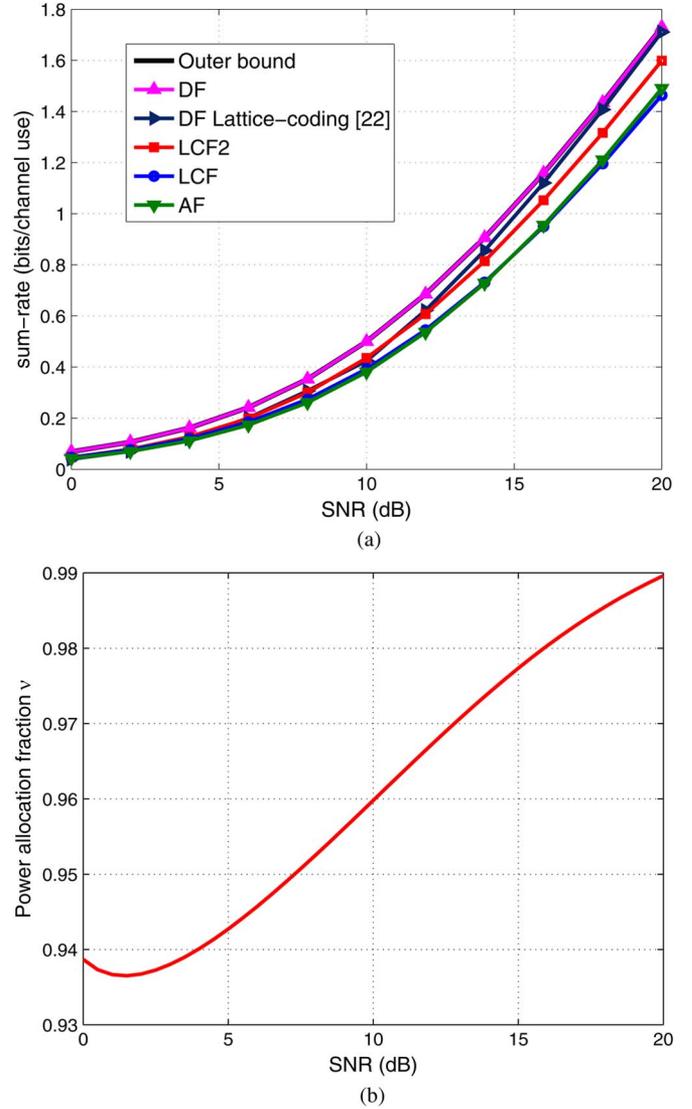


Fig. 8. Maximum sum-rate and relay power fraction allocated to the common message in LCF2 for asymmetric channels and equal powers. Here $\text{SNR} = P_1/\sigma_R^2 = P_2/\sigma_R^2 = P_R/\sigma_1^2$ and $|h_1|^2 = 4$, $|h_2|^2 = 0.1$; LCF2 outperforms DF-lattice coding for $\text{SNR} < 11$ dB. (a) Maximum sum-rate for different schemes; (b) Relay power fraction ν in LCF2.

channel conditions, a small power fraction is sufficient for the refinement message to ameliorate the performance of LCF2 compared to the basic scheme as depicted in Fig. 9(a).

Remark 3: We have assumed in our system model perfect CSI at all nodes. However, in the proposed two lattice-based coding schemes (LCF1 and LCF2), this perfect knowledge of the channel state can be relaxed. In fact, in order to compress its received signal, the relay needs only the module of the channel gains to reconstruct its encoding scheme. For each terminal, the decoder uses the available side information $S_i = h_i X_i$ that depends on its terminal-relay channel. Appropriate training sequences can be employed to estimate the channel of the relay. Furthermore, each decoder estimates only its unknown part of the relay received signals. It is shown in Sections III-B and IV-B that that the link between both terminals is equivalent to a Gaussian channel for both proposed schemes for both proposed schemes. Thus, a training sequence can also be used in order to estimate at each decoder, the channel on the other link.

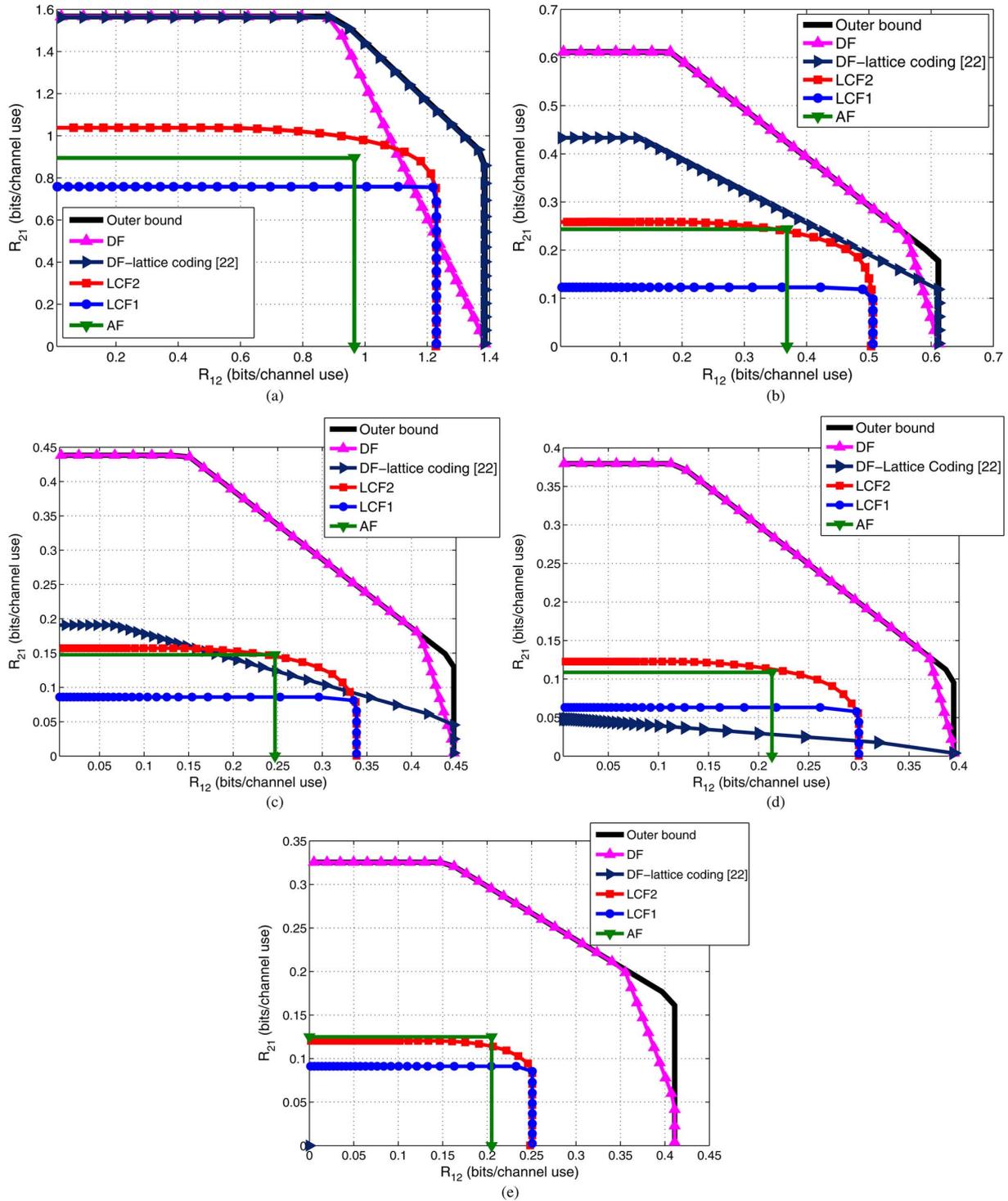


Fig. 9. Achievable rate regions for different channel and power settings. (a) $P_1 = 20$ dB, $P_2 = 20$ dB, $P_R = 15$ dB, $|h_1|^2 = 4$, $|h_2|^2 = 0.5$; (b) $P_1 = 10$ dB, $P_2 = 10$ dB, $P_R = 10$ dB, $|h_1|^2 = 4$, $|h_2|^2 = 0.2$; (c) $P_1 = 5$ dB, $P_2 = 4$ dB, $P_R = 4$ dB, $|h_1|^2 = 4$ and $|h_2|^2 = 0.5$; (d) $P_1 = 5$ dB, $P_2 = 3$ dB, $P_R = 3$ dB, $|h_1|^2 = 4$ and $|h_2|^2 = 0.5$; (e) $P_1 = 2$ dB, $P_2 = 2$ dB, $P_R = 5$ dB, $|h_1|^2 = 2$ and $|h_2|^2 = 0.5$.

VI. CONCLUSION

In this paper, we studied the problem of exchanging messages over a Gaussian two-way relay channel. We derived two achievable rate regions based on compress and forward lattice coding.

In the proposed schemes, the relay uses a lattice based Wyner-Ziv encoding by taking into account the presence of the

side information at each node. (i.e., the signal broadcasted by the relay includes also the signal that has been transmitted by each user to the relay during the first MAC transmission phase).

First, we developed a coding scheme in which the relay broadcasts the same signal to both terminals. We showed that this scheme offers the same performance as random coding based compress-and-forward protocol [18]. Then, we

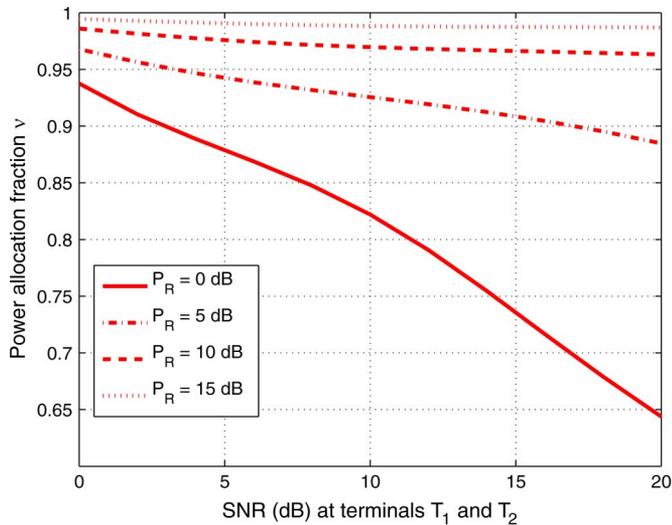


Fig. 10. Relay power allocation fraction ν in LCF2 for different fixed relay transmit power P_R and equal SNR $= P_1/\sigma_R^2 = P_2/\sigma_R^2$ at both terminals, $|h_1|^2 = 4$ and $|h_2|^2 = 0.5$.

proposed, and analyzed the performance of, an improved coding scheme in which the relay sends not only a common description of its output, but also an individual description that is destined to be recovered by only the user who experiences better channel conditions and better side information. We showed that this results in substantial gains in rates. Numerical results demonstrate an enhancement of the achievable rate region over the basic scheme up to 100% for moderate SNR regime and asymmetric channel conditions. Also, the improved scheme outperforms classic amplify-and-forward at all SNR values, and classic and lattice coding decode-and-forward for certain SNR regimes. This scheme can achieve higher performance than DF strategies with less complexity at the relay without use of full CSI.

Finally, it is worth mentioning that our schemes are based on structured codes that have low complexity compared to random coding from practical viewpoints. However, in these schemes, lattices codewords are used only at the relay while Gaussian codewords are used at the terminals' nodes. Considering lattice codes at all the nodes can be even more appropriate for practical systems.

REFERENCES

- [1] R. Ahlswede, N. Cai, S.-Y. R. Li, and R. W. Yeung, "Network information flow," *IEEE Trans. Inform. Theory*, vol. 46, no. 4, pp. 1204–1216, Jul. 2000.
- [2] S. Zhang, S. Liew, and P. Lam, "Physical layer network coding," in *Proc. ACM MOBICOM*, Los Angeles, CA, USA, 2006, pp. 358–365.
- [3] R. Knopp, "Two-way wireless communication via a relay station," in *GDR-ISIS Meet.*, Paris, France, Mar. 2007, pp. 1–1–21–1.
- [4] S. J. Kim, P. Mitran, and V. Tarokh, "Performance bounds for bi-directional coded cooperation protocols," *IEEE Trans. Inform. Theory*, vol. 54, no. 11, pp. 5235–5241, Nov. 2008.
- [5] B. Rankov and A. Wittneben, "Spectral efficient signaling for half-duplex relay channels," in *Proc. ACSSC*, Pacific Grove, CA, USA, Nov. 2005, pp. 1066–1071.
- [6] R. Knopp, "Two-way radio networks with a star topology," in *Proc. Int. Zurich Semin. Commun.*, Zurich, Switzerland, Feb. 2006.
- [7] M. P. Wilson, K. Narayanan, H. D. Pfister, and A. Sprintson, "Joint physical layer coding and network coding for bidirectional relaying," *IEEE Trans. Inform. Theory*, vol. 56, no. 11, pp. 5641–5654, Nov. 2010.

- [8] B. Nazer and M. Gastpar, "Compute-and-forward: Harnessing interference through structured codes," *IEEE Trans. Inform. Theory*, vol. 57, no. 10, pp. 6463–6486, Oct. 2011.
- [9] Y. Song, N. Devroye, H.-R. Shao, and C. Ngo, "Lattice coding for the two-way two-relay channel," in *Proc. IEEE ISIT*, 2013, pp. 1312–1316.
- [10] B. Rankov and A. Wittneben, "Achievable rate regions for the two-way relay channel," in *Proc. IEEE Int. Symp. Inf. Theory*, Seattle, WA, USA, Jul. 2006, pp. 1668–1672.
- [11] T. M. Cover and A. E. Gamal, "Capacity theorems for the relay channel," *IEEE Trans. Inform. Theory*, vol. 25, no. 5, pp. 572–584, Sep. 1979.
- [12] C. Schnurr, T. J. Oechtering, and S. Stanczak, "Achievable rates for the restricted half-duplex two-way relay channel," in *Proc. 41st ACSSC*, Pacific Grove, CA, USA, Nov. 2007.
- [13] S. J. Kim, N. Devroye, P. Mitran, and V. Tarokh, "Comparison of bi-directional relaying protocols," in *IEEE Sarnoff Symp.*, Princeton, NJ, USA, Apr. 2008, pp. 1–5.
- [14] D. Gunduz, E. Tuncel, and J. Nayak, "Rate regions for the separated two-way relay channel," in *Proc. 46th Annu. Allerton Conf. Commun., Control, Comput.*, Urbana-Champaign, IL, USA, Sep. 2008, pp. 1333–1340.
- [15] J. H. Conway and N. J. Sloane, *Sphere Packings, Lattices and Groups*, 3rd ed. New York, NY, USA: Springer-Verlag, 1998.
- [16] Y. Song and N. Devroye, "A lattice compress-and-forward scheme," in *Proc. IEEE ITW*, 2011, pp. 110–114.
- [17] Y. Song and N. Devroye, "Lattice codes for the Gaussian relay channel: Decode-and-forward and compress-and-forward," *IEEE Trans. Inform. Theory*, vol. 59, no. 8, pp. 4927–4948, Aug. 2013.
- [18] S. Smirani, M. Kamoun, M. Sarkiss, A. Zaidi, and P. Duhamel, "Wyner-Ziv lattice coding for two-way relay channel," in *Proc. Int. Conf. ATC*, Hanoi, Vietnam, Oct. 2012.
- [19] R. Zamir, S. Shamai, and U. Erez, "Nested linear/lattice codes for structured multiterminal binning," *IEEE Trans. Inform. Theory*, vol. 48, no. 6, pp. 1250–1276, Jun. 2002.
- [20] J. Nayak, E. Tuncel, and D. Gunduz, "Wyner-Ziv coding over broadcast channels: Digital schemes," *IEEE Trans. Inform. Theory*, vol. 56, no. 4, pp. 1782–1799, Apr. 2010.
- [21] W. Nam, S.-Y. Chung, and Y. H. Lee, "Capacity of the Gaussian two-way relay channel to within 1/2 bit," *IEEE Trans. Inform. Theory*, vol. 56, no. 11, pp. 5488–5494, Nov. 2010.
- [22] Y. Tian, D. Wu, C. Yang, and A. F. Molisch, "Asymmetric two-way relay with doubly nested lattice codes," *IEEE Trans. Wireless Commun.*, vol. 11, no. 2, pp. 694–702, Feb. 2012.
- [23] U. Erez and R. Zamir, "Achieving $(1/2) \log(1 + \text{SNR})$ on the AWGN channel with lattice encoding and decoding," *IEEE Trans. Inform. Theory*, vol. 50, no. 10, pp. 2293–2314, Oct. 2004.
- [24] R. Zamir and M. Feder, "On lattice quantization noise," *IEEE Trans. Inform. Theory*, vol. 42, no. 4, pp. 1152–1159, Jul. 1996.
- [25] G. Poltyrev, "On coding without restrictions for the AWGN channel," *IEEE Trans. Inform. Theory*, vol. 40, no. 52, pp. 409–417, Mar. 1994.
- [26] U. Erez, S. Litsyn, and R. Zamir, "Lattices which are good for (almost) everything," *IEEE Trans. Inform. Theory*, vol. 51, no. 10, pp. 3401–3416, Oct. 2005.
- [27] T. Cover and J. Thomas, *Elements of Information Theory*. New York, NY, USA: Wiley, 1991.
- [28] A. Wyner, "The rate-distortion function for source coding with side information at the decoder—II: General sources," *Inf. Control*, vol. 38, no. 1, pp. 60–80, Jul. 1978.
- [29] S. J. Kim, N. Devroye, P. Mitran, and V. Tarokh, "Achievable rate regions and performance comparison of half duplex bi-directional relaying protocols," *IEEE Trans. Inform. Theory*, vol. 57, no. 10, pp. 6405–6418, Oct. 2011.



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